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The clustering of galaxies in the completed SDSS-III Baryon Oscillation Spectroscopic Survey: On the measurement of growth rate using galaxy correlation functions

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ABSTRACT

We present a measurement of the linear growth rate of structure, f from the Sloan Digital Sky Survey III (SDSS III) Baryon Oscillation Spectroscopic Survey (BOSS) Data Release 12 (DR12) using Convolution Lagrangian Perturbation Theory (CLPT) with Gaussian Streaming Redshift-Space Distortions (GSRSD) to model the two point statistics of BOSS galaxies in DR12. The BOSS-DR12 dataset includes 1,198,006 massive galaxies spread over the redshift range $0.2 < z < 0.75$. These galaxy samples are categorized in three redshift bins. Using CLPT-GSRSD in our analysis of the combined sample of the three redshift bins, we report measurements of $f\sigma_8$ for the three redshift bins. We find $f\sigma_8 = 0.430 \pm 0.054$ at $z_{\text{eff}} = 0.38$, $f\sigma_8 = 0.452 \pm 0.057$ at $z_{\text{eff}} = 0.51$ and $f\sigma_8 = 0.457 \pm 0.052$ at $z_{\text{eff}} = 0.61$. Our results are consistent with the predictions of Planck Λ CDM-GR. Our constraints on the growth rates of structure in the Universe at different redshifts serve as a useful probe, which can help distinguish between a model of the Universe based on dark energy and models based on modified theories of gravity. This paper is part of a set that analyses the final galaxy clustering dataset from BOSS. The measurements and likelihoods presented here are combined with others in Alam et al. (2016) to produce the final cosmological constraints from BOSS.

Key words: Cosmology: dark energy - Cosmology: large scale structure of Universe - Galaxies: statistics - Cosmology: cosmological parameters.

1 INTRODUCTION

The theory of General Relativity (GR) gives us a relation between the expansion rate of the Universe and its matter and energy content (Einstein 1915, 1916). At the same time, cosmological observations have also given us a glimpse into the Universe’s dark sector. The observation of the accelerated expansion of the Universe is a landmark discovery in cosmology (Riess et al. 1998; Perlmutter et al. 1999). The acceleration of the expansion of the Universe is most commonly explained by a framework which suggests that our Universe is dominated by a ‘dark energy’ field with negative pressure. The dark energy is similar to the cosmological constant (Λ) in Einstein’s theory of General Relativity (Padmanabhan 2007). The Λ CDM-GR model which proposed the accelerated expansion of the Universe is in consonance with probes such as the Cosmic Microwave Background (CMB) (Bennett et al. 2013; Planck Collaboration et al. 2014a) and Baryon Acoustic Oscillations (BAO) (Eisenstein et al. 2005; Cole et al. 2005; Hütsi 2006; Percival et al. 2007; Kazin et al. 2010; Percival et al. 2010; Reid et al. 2010). The observation that the expansion of the Universe is accelerating can also pertain to the possibility of ‘dark gravity’ (Henry-Couannier 2005a,b; Bludman 2007; Durrer & Maartens 2007; Henry-Couannier et al. 2007; Heavens 2009; Lobo 2011, 2012), which suggests that General Relativity is incorrect on the largest scales and is a limit of a more complete theory of gravity. Such a possibility gives scope to the explanation of the accelerated expansion of the Universe by frameworks which try to reproduce cosmological observations by modifying the form of the equations of GR. The cause for the acceleration in the expansion of the Universe remains a mystery and one cannot settle on a preferred candidate to explain the measurement of the expansion history $H(z)$ from spectroscopic surveys and probes like Type Ia supernova and BAO since modified theories of gravity (Carroll et al. 2004; Kolb et al. 2006; Carroll et al. 2006; Cardone et al. 2012) and theories based on dark energy explain the observations equally well. In other words, the measurement of the expansion rate alone will not be able to distinguish between a model based on dark energy and modified theories of gravity.

One way of resolving this conundrum lies in the investigation of the growth rate of structure inside the Universe (Peacock et al. 2006; Albrecht et al. 2007; Pouri et al. 2013, 2014; Alam et al. 2015a; Mohammad et al. 2015). The growth rate of structure in the Universe is decided by the competing effects of the gravitational collapse of density fluctuations, which accelerate their growth and the expansion rate, which inhibits it. Since the theory of General Relativity gives us a relation between the growth rate of cosmological structure and the expansion history of the Universe, measurements of the growth rate give us a handle on the underlying theory of gravity. In cosmological observations, the positions of galaxies are mapped by redshift, which correspond to the true distance according to the Hubble Law. Probes which look at the growth of structure in the Universe also include the peculiar velocities of galaxies. The observed redshift (z) is, in fact, a sum of the Hubble recession velocity and the peculiar velocity caused by gravitational dynamics. Components from the peculiar velocities, which are deviations of galaxies’ velocity from pure Hubble flow, combine with components from the Hubble flow to give rise to dis-

tortions in the reconstructed spatial distribution of the observed objects. The ensuing distortions manifest themselves as anisotropy in the distribution of objects and are caused in the radial direction in the redshift space (Kaiser 1987; Hamilton 1992; Cole et al. 1995; Guzzo et al. 1997; Peacock et al. 2001; Scoccimarro 2004; Tegmark et al. 2006; White et al. 2009; Percival & White 2009; Yoo 2009; McDonald 2009; McDonald & Seljak 2009; Percival et al. 2011; Reid & White 2011; Yoo et al. 2012; Samushia et al. 2012; McQuinn & White 2013; Beutler et al. 2014; White et al. 2015; Simpson et al. 2016). These distortions are referred to as ‘redshift space distortions’ (RSD).

Redshift space distortions can be used to reveal information about the motion of galaxies and underlying matter distribution in the Universe. The distinctive features of RSD are revealed in the two-point correlation statistics of galaxy distributions which are obtained as functions of variables representing distances parallel and perpendicular to the line of sight (s_{\parallel} and s_{\perp} respectively). In small spatial scales where galaxies with high velocities are dominant, RSD is manifested as elongation in redshift space maps with an axis of elongation pointing towards the observer (i.e., along s_{\parallel}). This phenomenon is referred to as the ‘‘Fingers of God’’ effect (Jackson 1972; Tegmark et al. 2004). On larger scales, one observes the ‘‘Kaiser effect’’ where coherent peculiar velocities cause an apparent contraction of structure along the line of sight in redshift space. As a result, we see two distinct effects, *viz.* the non-linear and the linear effects due to small-scale elongation and large scale flattening in redshift scale maps. Measurement of the growth rate of structure from RSD is intricate. In 1987, Nick Kaiser (Kaiser 1987) tried to tackle this problem by introducing a prescription which discusses the redshift space power spectrum by modifying the linear theory of large scale structure. In his landmark work, Kaiser was able to relate the power spectrum in redshift space $P_s(\mathbf{k})$ and its counterpart in real space $P_r(\mathbf{k})$ by the following relation:

$$P_s(\mathbf{k}) = (1 + \beta\mu_k^2)^2 P_r(\mathbf{k}) \quad (1)$$

where μ_k corresponds to the cosine of the angle between \mathbf{k} and the line of sight and $\beta = \Omega_m^{0.55}/b$ is the linear distortion parameter. Here, Ω_m is the mass density parameter and b denotes the linear bias parameter. Exploration of the concept of peculiar velocities in non-linear scales using ideas of the ‘‘streaming model’’ was presented in Peebles (1980); Davis & Peebles (1983); Fisher (1995). Peebles (1980) showed that the factor $\Omega_m^{0.55}$ relates peculiar velocities to density fluctuations. The real space counterpart of the Fourier space formalism given by Kaiser was introduced by Hamilton (1992). Extensions of the linear model, called the ‘‘dispersion model’’ have been used to determine the growth rate from the two point galaxy correlation function $\xi(s_{\parallel}, s_{\perp})$ (Peacock et al. 2001; Hawkins et al. 2003). However, measurements of the growth rate parameter from the dispersion model have been found to introduce systematic errors in the results (Taruya et al. 2010; Bianchi et al. 2012). An important breakthrough in the dispersion model was effected by Taruya in 2010 (Taruya et al. 2010) when he proposed a new model of redshift space distortion which studied correction factors arising from the non-linear coupling between velocity and density fields. RSD analyses from the data releases 9 (DR09; Ahn et al. 2012) and 10 (DR10; Ahn et al. 2014) of Sloan Digital Sky Sur-

vey III (SDSS III; Eisenstein et al. 2011) which include the Baryon Oscillation Spectroscopic Survey (BOSS; Dawson et al. 2013), employed the Lagrangian Perturbation Theory discussed by Matsubara (2008a,b) and the Gaussian Streaming Model to measure the linear growth rate of structure in the Universe (Reid & White 2011; Reid et al. 2012; Samushia et al. 2013). Other measurements of the linear growth rate of the Universe ($f\sigma_8$) include Cooray et al. (2004); Percival et al. (2004); Narikawa & Yamamoto (2010); Blake et al. (2011); Giovannini (2011); Okumura & Jing (2012); Beutler et al. (2012); Gupta et al. (2012); Hirano et al. (2012); Hudson & Turnbull (2012); Nusser et al. (2012); Samushia et al. (2012); Shi et al. (2012); Contreras et al. (2013); de la Torre et al. (2013); Macaulay et al. (2013); Sánchez et al. (2013, 2014); Reid et al. (2014); Avsajanishvili et al. (2014); Alam et al. (2015a); Feix et al. (2015); Marulli et al. (2015); Hamaus et al. (2016).

In the work presented here, we follow Alam et al. (2015a) to study galaxies in the final data release (DR12; Alam et al. 2015b). We use the MultiDark Patchy mock catalogs (Kitaura et al. 2014, 2015) and the BOSS DR12 galaxy dataset (Reid et al. 2015) in our analysis to recover information about the growth rate of the Universe at different redshifts. Our paper is a support paper and the final cosmological analysis is discussed in Alam et al. (2016). In addition to our paper there are other companion papers of Alam et al. (2016) which analyze the full shape of the anisotropic two-point correlation function. Alam et al. (2016) use the methodology presented in Sánchez et al. (2016b) to combine the results of all these companion papers into a final set of BOSS consensus constraints and explore their cosmological implications. In section 3.1, we provide a brief summary of three different full shape analyses of galaxy clustering for the BOSS DR12 sample using different models (Beutler et al. 2016a; Grieb et al. 2016; Sánchez et al. 2016a), which are support papers to Alam et al. (2016). Other companion papers where the BAO scale is measured using the anisotropic two-point correlation function include Beutler et al. (2016b); Ross et al. (2016); Vargas Magaña et al. (2016).

Our paper is organized as follows. In section 2, we review the Convolution Lagrangian Perturbation Theory and the Gaussian Streaming Model which we use as the theoretical basis of our investigation. In section 3, we sketch the details of the BOSS DR12 galaxy dataset and the mock galaxy catalogs that we use in our analysis. We discuss details of the approach adopted in our analysis in section 4. Our results from the mocks and the galaxy data are discussed in section 5. We conduct a critical analysis and present a summary of the obtained results for cosmological parameters in sections 6 and 7.

2 THE CLPT-GSRSD MODEL

2.1 The Convolution Lagrangian Perturbation Theory

We choose the theoretical framework for the modeling of correlation functions proposed in the Convolution Lagrangian Perturbation Theory (CLPT) as the theoretical basis of our analysis. The framework of CLPT was proposed by Carlson et al. (2013). CLPT seeks to use a non-perturbative resum-

mation of the Lagrangian Perturbation Theory to make predictions of correlation functions in real and redshift-space. In CLPT, terms in the expansion of the two point correlation function $\langle \delta_X(\mathbf{x}_1)\delta_X(\mathbf{x}_2) \rangle$ for a tracer X , which become constants in the limit of large scale, are identified and kept from being expanded in the resummation. We discuss the concept of two point galaxy correlation functions in section 4.1. The lowest order in the expansion of the matter correlation function, $\xi(s)$ returns the Zel'dovich approximation. We introduce details of the concept of multipoles, i.e. the monopole ($\xi_0(s)$) and the quadrupole ($\xi_2(s)$) in equation 9. The monopole and the quadrupole from CLPT agree well with the results of N -body simulations till small scales ($20 h^{-1}\text{Mpc}$). Also, the results from CLPT tally with the results from the Lagrangian resummation theory (LRT) by Matsubara (Matsubara 2008a,b) on large scales. Although CLPT has been shown to perform better than LRT and the linear theory in the modeling of multipoles, the desire to model even smaller scales effectively necessitates better performance from the theory, especially with regard to the performance of the quadrupole on smaller scales.

2.2 The Gaussian streaming model

To overcome the deficiencies in the predictions of the linear theory one needs to consider the non-linear mapping between real space and redshift space and the non-linearity of the halo pairwise velocities. The Gaussian streaming model (GSM) attempts to non-perturbatively model the non-linear mapping between real space and redshift space positions in halo redshift-space correlation functions on quasilinear scales ($\sim 30 - 80 h^{-1}\text{Mpc}$) (Reid & White 2011). In the Gaussian streaming model, the pairwise velocity is assumed to have a Gaussian probability distribution function. The mean and the dispersion of the pairwise velocities depend on the pair separation vector and the angle of the pair separation vector with the line of sight. Reid & White (2011) presented the first calculation of next-to-leading-order corrections to pairwise mean infall velocities and dispersions for linearly biased haloes in their work. The scale dependent Gaussian streaming ansatz predicts the monopole and quadrupole estimators to accuracies of $\sim 0.5\%$ above $10 h^{-1}\text{Mpc}$ and $\sim 2\%$ above $25 h^{-1}\text{Mpc}$ respectively for halo correlation functions. However, the Fingers of God effect is expected to affect the observations at scales below $25 h^{-1}\text{Mpc}$. The multipoles $\xi_{2,4}$ obtained from GSM are significantly more enhanced on quasi-linear scales when compared to the multipoles obtained from the linear theory. Among the more detailed simulation based models which describe the velocity distribution of galaxies are Zu & Weinberg (2013) which analyzes the velocity distribution of galaxies around galaxy groups and Bianchi et al. (2015) which develops a prescription for galaxy pairwise velocities at large scales.

The level of accuracy of the results obtained from GSM leads to it being a desirable candidate for the analysis of two point correlation functions. In order to further improve the estimation of the multipoles from CLPT, the velocity statistics and the correlation function from CLPT are combined with the formalism of GSM to enhance the accuracy of the monopoles and quadrupoles predicted by the theory (Wang et al. 2014).

There are examples of papers in literature (Hawkins et

al. 2003) where it was found that pairwise velocities have non zero higher order moments and hence, they deviate from perfectly Gaussian distributions. Recently, people have conducted even more detailed studies of pairwise velocity moments, e.g., Bianchi et al. (2015, 2016) (in simulations) and Uhlemann et al. (2015) (in theory and simulations) and have found that the ensuing profiles of pairwise velocity distributions are not close to Gaussian distributions. Nevertheless, all the aforementioned papers have found that the effect of the non-Gaussian features of velocity moments in correlation functions are pronounced at non-linear scales and negligible at linear scales. Including higher order velocity moments will probably improve the modeling at smaller scales and lead to better signal to noise ratios in RSD measurements. But the evaluation of these components will involve higher order integrals (which will need to be tested using methods similar to those proposed in Uhlemann et al. (2015)). We consider such sophisticated tests and modeling beyond the scope of the current paper, and hence, we use just a Gaussian streaming model. This is one of the reasons why we fail to use scales smaller than $25 h^{-1}\text{Mpc}$ in our analysis.

2.3 CLPT-GSRSD

The idea to use predictions from CLPT along side GSM is motivated by the need to model multipoles more effectively on smaller scales. The use of CLPT to predict the components of GSM was suggested by Carlson et al. (2013) and was implemented by Wang et al. (2014).

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

In equation 2, y and s_{\parallel} represent the line of sight separation in real and redshift space respectively. The symbol s_{\perp} denotes the perpendicular separations in both redshift and real space. The parameter $r = \sqrt{y^2 + s_{\perp}^2}$ is indicative of the pair separation in real space, while $\mu = y/r$ depicts the cosine of the angle between the line of sight separation in real space y and the pair separation vector r . Reid et al. (2012) presented an extension of the GSM where the parameter σ_{FOG} was introduced and tested for the first time as a single extra parameter which deals with the incorporation of the Fingers of God effect in the GSM. The parameter σ_{FOG} serves to describe an isotropic dispersion that modifies the scale dependence of quadrupole moments on small distance scales. The parameters $v_{12}(\mathbf{r})$ and $\sigma_{12}(\mathbf{r})$ represent the mean infall velocity between pairs of matter tracers and the velocity dispersion along the line of sight respectively. Brief details of the calculation of the parameters $v_{12}(\mathbf{r})$ and $\sigma_{12}(\mathbf{r})$ are as follows:

$$v_{12,n}(\mathbf{r}) = [1 + \xi(r)]^{-1} \int M_{1,n}(\mathbf{r}, \mathbf{q}) d^3q$$

$$\sigma_{12,nm}^2(\mathbf{r}) = [1 + \xi(r)]^{-1} \int M_{2,nm}(\mathbf{r}, \mathbf{q}) d^3q \quad (3)$$

Here, $M_{1,n}(\mathbf{r}, \mathbf{q})$ and $M_{2,nm}(\mathbf{r}, \mathbf{q})$ are specific integrals in CLPT that draw on the linear matter power spectrum

$P_{1\text{in}}(\mathbf{k})$. The indices n and m are representative of directions along the galaxy position vectors. Details of the form of the integrals $M_{1,n}(\mathbf{r}, \mathbf{q})$ and $M_{2,nm}(\mathbf{r}, \mathbf{q})$ are expounded in Wang et al. (2014). The projection of the mean infall velocity between pairs of matter tracers, $v_{12,n}(\mathbf{r})$ along the direction of pair separation vector, \mathbf{r} gives the radial component of v_{12} which is obtained as $v_{12} = v_{12,n}\hat{r}_n$.

From the projection of the pairwise velocity dispersion in directions along and perpendicular to the pairwise separation unit vector \hat{r} , one obtains the parameters σ_{\parallel}^2 and σ_{\perp}^2 respectively. Equation 4 shows the scheme in which $\sigma_{12,nm}^2$ is contracted to obtain σ_{\parallel}^2 and σ_{\perp}^2 .

$$\sigma_{\parallel}^2 = \sigma_{12,nm}^2 \hat{r}_n \hat{r}_m$$

$$\sigma_{\perp}^2 = \left(\sigma_{12,nm}^2 \delta_{nm}^K - \sigma_{\parallel}^2 \right) / 2 \quad (4)$$

Using the knowledge of the components of the pairwise velocity dispersion (i.e. σ_{\parallel}^2 and σ_{\perp}^2), one can obtain σ_{12}^2 as functions of the components σ_{\parallel}^2 and σ_{\perp}^2 .

$$\sigma_{12}^2(r, \mu) = \mu^2 \sigma_{\parallel}^2(r) + (1 - \mu^2) \sigma_{\perp}^2(r) \quad (5)$$

The combination of CLPT and GSM is shown to improve the accuracy for $\xi_0(s)$ and $\xi_2(s)$ to 2 and 4 percent respectively till scales of $s > 25 h^{-1}\text{Mpc}$. The ensuing model is referred to as the Convolution Lagrangian Perturbation Theory with Gaussian Streaming Redshift Space Distortions (CLPT-GSRSD). We will use the CLPT-GSRSD model to derive deductions about constraints on cosmological parameters including the growth rate.

3 DATA

In this section we sketch the nuances of the galaxy and the mock datasets that we use in our analysis. We briefly touch upon the particulars of the BOSS DR12 galaxy dataset and give necessary details of the Multi-Dark Patchy mock catalogs.

3.1 The BOSS DR12 Galaxy Dataset

We use the Data Release 12 (DR12; Alam et al. 2015b) of the Sloan Digital Sky Survey III (Eisenstein et al. 2011) Baryon Oscillation Spectroscopic Survey (Dawson et al. 2013) dataset in our analysis. The dataset includes an ensemble of galaxies obtained from the SDSS survey using a wide-field, drift-scanning CCD camera (Gunn et al. 1998, 2006) and selected using multicolor imaging photometry in five color bands (u, g, r, i, z ; Fukugita et al. 1996). This dataset was used by BOSS multi-fibre spectrographs (Bolton et al. 2012; Smee 2013) to observe the spectra of 1,198,006 galaxies. These galaxies were observed along with approximately 300,000 quasars, 200,000 stars and 400,000 ancillary objects. The observations were made in a sequence of 15-minute exposures and integrated until a minimum signal-to-noise ratio is reached for the faint galaxy targets. This procedure ascertains homogeneity in the dataset with a redshift completeness of over 97% for the entire survey footprint. Bolton et al. (2012) describe the method for the determination of redshift from the classified spectra.

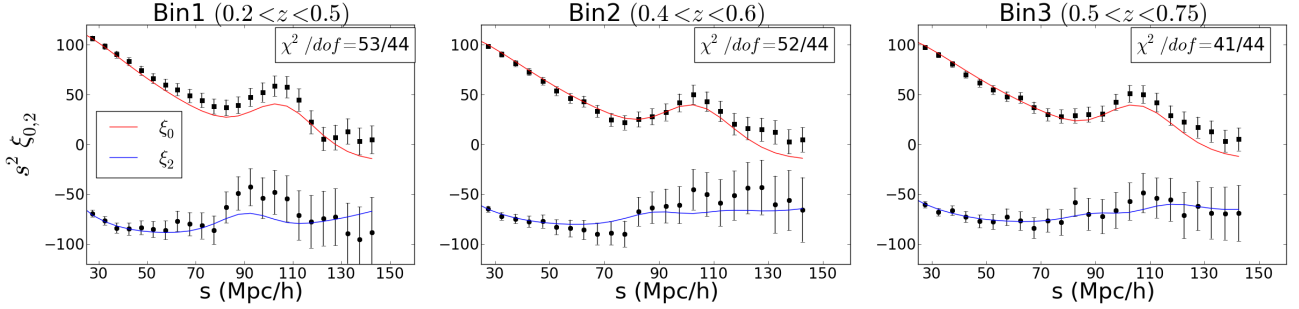


Figure 1. The black dots in the plots of this figure represent monopole (\blacksquare) and quadrupole (\bullet) for BOSS DR12 galaxy sample evaluated at different values of s . The error bars are obtained from the diagonal elements of the covariance matrices corresponding to mocks in the three redshift bins. The red and the blue lines denote the best fit models of monopole and quadrupole of the galaxy data. The analysis assumes a fitting range $25 h^{-1}\text{Mpc} \leq s \leq 150 h^{-1}\text{Mpc}$ with a bin size of $5 h^{-1}\text{Mpc}$.

The selected sample of 1,198,006 galaxies encompasses a redshift range of 0.2 to 0.75 and covers 9,329 square degrees. For the purpose of this paper, we divide this redshift range into three overlapping redshift bins of roughly equal volume (Alam et al. 2016), viz. $0.2 < z < 0.5$ (*bin1*), $0.4 < z < 0.6$ (*bin2*) and $0.5 < z < 0.75$ (*bin3*). These bins have effective redshifts of $z_{\text{eff}} = 0.38, 0.51$ and 0.61 respectively. For the three bins, we work with a $\Lambda\text{CDM-GR}$ cosmological model with a fiducial cosmology of $\Omega_m = 0.31$, $H_0 = 0.676$, $\Omega_\Lambda = 0.69$, $\Omega_b h^2 = 0.022$ and $\sigma_8 = 0.80$.

We follow the methods outlined in Ross et al. (2012) and Anderson et al. (2014) to give weights to each galaxy under study to compensate for the effects of redshift failures and fiber collisions. We introduce the weight factor w_{zf} to account for redshift failure of the nearest neighbor of a galaxy. Similarly, the weight factor w_{cp} is intended to account for a scenario where the redshift of a neighbor was not obtained because it was in a close pair. The weight factor w_{star} serves to correct for the non-cosmological fluctuations which arise due to the dependence of target identification on the local star density. The weight factor w_{see} corrects the effect of the seeing conditions (during photometric observations) on the target density. The factor w_{FKP} is included because of the need to minimize the variance in the weighted number of galaxy counts. Following is the weighting scheme that we use: $w_{\text{tot}} = (w_{\text{cp}} + w_{\text{zf}} - 1) w_{\text{star}} w_{\text{see}} w_{\text{FKP}}$.

Four companion papers (including this paper) present different approaches for the full-shape analysis of the BOSS DR12 combined galaxy sample (Alam et al. 2016):

(i) In this work, we use multipoles obtained from anisotropic two-point galaxy correlation functions to analyze the DR12 data. Details of the approach used for our analysis are given in section 4.

(ii) Sánchez et al. (2016a) present an analysis of the BOSS DR12 combined galaxy sample using wedges obtained from anisotropic two-point correlation functions.

(iii) The methodology presented in Beutler et al. (2016a) for the analysis of DR12 galaxy data employs multipoles obtained from anisotropic power spectrum.

(iv) Grieb et al. (2016) use an analysis based on wedges from anisotropic power spectrum in their investigation of the BOSS DR12 galaxy data.

3.2 Mock Galaxy Catalogs

We use the Multi-Dark Patchy (MD-P) mock catalogs (Kitauro et al. 2014, 2015) as an essential statistical tool and as a precursor to the analysis of the SDSS III DR12 combined galaxy dataset. These mock catalogs require the generation of accurate reference catalogs. For these MD-P mock catalogs, the reference catalogs are extracted from one of the BigMultiDark cosmological N-body simulations (Klypin et al. 2014) which uses gadget-2 (Springel 2005) with 3840^3 particles in a volume of $(2.5 h^{-1}\text{Mpc})^3$. These simulations are based on a ΛCDM cosmology of $H_0 = 67.77 \text{ km.s}^{-1}.\text{Mpc}^{-1}$, $\Omega_m = 0.307115$, $\Omega_b = 0.048206$, $n_s = 0.9611$ and $\sigma_8 = 0.8288$.

In a manner akin to the division of the BOSS DR12 data into redshift bins, the MD-P mock catalogs that we use are segregated into three redshift bins with effective redshifts of $z_{\text{eff}} = 0.38$ (*bin1*), 0.51 (*bin2*) and 0.61 (*bin3*). In each redshift bin we use 997 mocks in our analysis. The primary purpose of the use of the MD-P mocks is to assist in the formulation of covariance matrices for the different bins of the galaxy dataset and to mimic the statistics of the same. We discuss more about the use of the MD-P mocks to obtain covariance matrices in sections 4.1 and 4.2.

4 ANALYSIS

In this section, we outline the methodology that we have used in our analysis of the MultiDark-Patchy (MD-P) mock catalogs and the BOSS DR12 dataset in the three redshift bins. We sketch the steps that we employ in analyzing the positions of galaxies to obtain multipoles ($\xi_0(r)$ and $\xi_2(r)$) from two-point correlation functions ($\xi(r)$). We also discuss the computation of covariance matrices from MD-P mock catalogs. We conclude by shedding light on the use of Markov Chain Monte Carlo (MCMC) in chosen parameter spaces for the MD-P mocks and the SDSS III galaxy dataset to obtain a handle on the variation of different RSD and BAO parameters.

4.1 The two-point galaxy correlation function

In the fiducial cosmology $\Omega_m = 0.31$, $H_0 = 0.676$, $\Omega_\Lambda = 0.69$, $\Omega_b h^2 = 0.022$ and $\sigma_8 = 0.80$, we map redshift and

celestial coordinates (α, δ) to the position of a galaxy in three dimensional space. We use the Landy-Szalay estimator (Landy & Szalay 1993) to obtain the two-point correlation function $\hat{\xi}(s)$ for a given galaxy sample.

$$\hat{\xi}_{\text{LS}}(s, \mu) = \frac{DD(s, \mu) - 2DR(s, \mu) + RR(s, \mu)}{RR(s, \mu)} \quad (6)$$

Here $\mu = \cos \theta$ (where θ is the angle between the line of sight and the radial distances), $DD(s, \mu)$ is the pair count of galaxies with separation s and orientation μ , $DR(s, \mu)$ is the cross-pair counts between the galaxies and a random distribution, and $RR(s, \mu)$ is the number of pairs for a random distribution. The Landy-Szalay estimator has only a second order bias caused by finite sample effects. The performance of the Landy-Szalay estimator has been proved to be better than other comparable two point correlation functions at large scales (Pons-Bordería et al. 1999; Kerscher et al. 2000). In the measurement of the two-point correlation function, each galaxy pair is weighted by w_{tot} . More details of the weighting scheme that we use can be found in section 3.1.

The use of two dimensional two-point correlation functions ($\hat{\xi}(s, \mu)$) will lead to a large number of bins (due to the presence of two dimensions). To fit a two-dimensional two-point correlation function directly, we will need to construct a large covariance matrix which will in turn necessitate the use of a very large number of mocks. Creating such a large number of mocks will be computationally intensive. This motivates a need to reduce the number of bins. We use different kernels to obtain averages of the two dimensional two-point correlation function ($\hat{\xi}(s, \mu)$) to condense the information contained in it.

$$\tilde{\xi}_f(s) = \int f(s, \mu) \hat{\xi}(s, \mu) d\mu \quad (7)$$

Here $f(s, \mu)$ is an appropriately selected kernel function. We outline two different ways of defining these kernels; one, where we take recourse to the use of clustering wedges, $\hat{\xi}_{\Delta\mu}(s)$, and another, where we use angle averaged correlation functions, $\tilde{\xi}(s)$.

In the first approach, one obtains a clustering wedge by averaging the two dimensional two-point correlation function ($\hat{\xi}(s, \mu)$) over a chosen interval $\Delta\mu = \mu_{\text{max}} - \mu_{\text{min}}$ (Kazin et al. 2012; Sánchez et al. 2013, 2014), i.e.

$$\begin{aligned} \tilde{\xi}_{\Delta\mu}(s) &\equiv \frac{1}{(\mu_{\text{max}} - \mu_{\text{min}})} \int_{\mu_{\text{min}}}^{\mu_{\text{max}}} \hat{\xi}_{\text{LS}}(s, \mu) d\mu \\ &\approx \frac{1}{(\mu_{\text{max}} - \mu_{\text{min}})} \sum_j \Delta\mu_j \hat{\xi}_{\text{LS}}(s, \mu_j) \end{aligned} \quad (8)$$

In the second approach, we obtain isotropized correlation functions by finding projections of the anisotropic two dimensional two-point correlation functions in the basis of Legendre polynomials. Hamilton (1992) demonstrated one of the first uses of spherical harmonics as kernels when he used Legendre polynomials to obtain averages of the anisotropic $\xi(s, \mu)$. In our analysis, we also use the orthonormal basis of Legendre polynomials $P_\ell(\mu)$ as our kernel functions to obtain ‘multipoles’ of different orders, ℓ ($\ell = 0, 1, 2 \dots$).

$$\begin{aligned} \tilde{\xi}_\ell(s) &= \frac{2\ell + 1}{2} \int_{-1}^1 \hat{\xi}_{\text{LS}}(s, \mu) P_\ell(\mu) d\mu \\ &\approx \frac{2\ell + 1}{2} \sum_j \Delta\mu_j \hat{\xi}_{\text{LS}}(s, \mu_j) P_\ell(\mu_j) \end{aligned} \quad (9)$$

Due to the anti-symmetry of Legendre polynomials of odd orders, the angle averaged 2-point correlation function of odd orders ($\tilde{\xi}_{2\ell+1}(s)$) vanish. We use even order multipoles i.e., monopole ($\tilde{\xi}_0(s)$) and the quadrupole ($\tilde{\xi}_2(s)$) for our analysis. One observes huge errors for multipoles of the fourth order, viz. hexadecapoles ($\tilde{\xi}_4(s)$). Hexadecapoles from the theoretical model are suspected to be more susceptible to survey systematics and the correction schemes adopted in the model than $\tilde{\xi}_0(s)$ and $\tilde{\xi}_2(s)$. Also, for hexadecapoles, one sees disagreement between the results of CLPT and N -body simulations (Carlson et al. 2013). Hence, we base our analysis on multipoles of orders no higher than 2.

The datasets that we use have 120 bins in μ . We use equation 9 to compute multipoles corresponding to the MD-P mocks and the BOSS DR12 dataset in evenly spaced bins of width $5 h^{-1} \text{Mpc}$ in s . Our analysis has been found to be relatively insensitive to bin sizes for bin sizes between $2 - 8 h^{-1} \text{Mpc}$ (Alam et al. 2015a). Since these multipoles are computed from data, we will use the notation ξ^{data} to refer to these multipoles. We bin the correlation function multipoles obtained from theory into the same bins as that of data for the purpose of comparison of the multipoles obtained from data with those obtained from theory. In the first step, we calculate $\xi_0(s)$ and $\xi_2(s)$ for the MD-P mocks in each redshift bin using the information about galaxy correlation functions $\xi(s, \mu)$ at different positions. Figure 1 illustrates the monopole and the quadrupole for BOSS DR12 galaxy data and the best fit models of the data for the three bins. More details about the techniques used to obtain the error bars in Figure 1 are given in section 4.2 while sections 4.3 and 4.4 describe the methods used to obtain the best fit models. In section 5.3 we give a comprehensive description of the details of Figure 1.

4.2 The covariance matrix

We use the multipoles corresponding to the MD-P mocks in a given bin to get an estimate of the covariance matrix of that bin. We follow the prescription outlined in Vargas Magaña et al. (2013); Percival (2013) to calculate the covariance matrix from the monopole and the quadrupole corresponding to the mocks.

$$\hat{\Sigma}_{ij} = \left[\sum_{n=1}^{N_{\text{mock}}} (\tilde{\xi}_{i,n} - \bar{\xi})(\tilde{\xi}_{j,n} - \bar{\xi}) \right] / [N_{\text{mock}} - 1] \quad (10)$$

In equation 10, $\hat{\Sigma}_{ij}$ is the (i, j) th entry of the calculated covariance matrix, where the indices ‘ i, j ’ correspond to the index of the binned value of the radial position, s in the two point correlation function. N_{mock} is the total number of mocks and the index ‘ n ’ denotes the number (index) of a given mock. $\bar{\xi}$ corresponds to the mean of the mocks. The

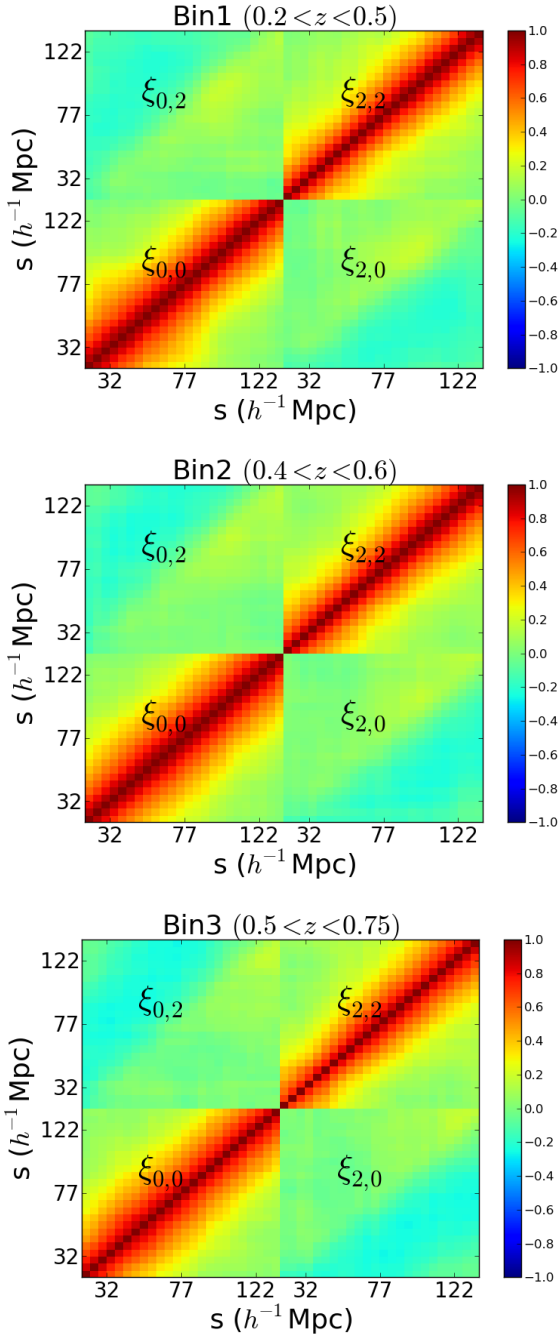


Figure 2. The plot on the top of the figure shows the correlation matrices obtained from MD-P mocks with $z_{\text{eff}} = 0.38$. The plot in the center denotes the correlation matrix gotten from MD-P mocks with $z_{\text{eff}} = 0.51$. The plot on the bottom of the figure shows the correlation matrix obtained from MD-P mocks with $z_{\text{eff}} = 0.61$.

correlation matrix, $\hat{\mathbf{r}}$ is obtained from the covariance matrix $\hat{\Sigma}$ as

$$\hat{r}_{ij} = \frac{\hat{\Sigma}_{ij}}{\sqrt{\hat{\Sigma}_{ii}\hat{\Sigma}_{jj}}} \quad (11)$$

The inverse of the covariance matrix obtained from the

mocks is used to estimate the likelihood as shown in equation 17.

Figure 2 shows the correlation matrices obtained from the three redshift bins of the MD-P mock catalogs, i.e. $z_{\text{eff}} = 0.38, 0.51$ and 0.61 .

4.3 The CosmoMC code and parameterizations

We compare the results for multipoles that we obtain for data (or mock catalogs) using equation 9 with the correlation functions that we obtain from theory using equation 2 in a Markov Chain Monte Carlo (MCMC) analysis which operates in a chosen parameter space. We use COSMOMC (Lewis & Bridle 2002) to effectuate the MCMC algorithm to explore the chosen cosmological parameter space. For the MD-P mocks we use a six dimensional parameter space with $\{F_1, F_2, f, \sigma_{\text{FOG}}, \alpha, \epsilon\}$ being the parameters in the MCMC analysis. For the three bins in the BOSS DR12 galaxy data, we explore a ten dimensional parameter space: $\{\Omega_b h^2, \Omega_c h^2, n_s, \ln(10^{10} A_s), F_1, F_2, f, \sigma_{\text{FOG}}, \alpha, \epsilon\}$ in the MCMC analysis. Here Ω_b, Ω_c represent the baryon density and the dark matter density respectively and the symbol h denotes $H_0/100$ where H_0 is the Hubble constant. The parameter A_s is the amplitude of the primordial power spectrum. The quantities F_1, F_2 are the first and the second order Lagrangian biases while the quantity f depicts the logarithm derivative of the growth factor, i.e. $f = d \ln(D)/d \ln(a)$. For Λ CDM cosmology, the growth factor f can be related to the matter density Ω_m by the approximation $f(a) \approx \Omega^{0.55}(a)$. It is common to report measurements of the linear growth rate in terms of $f\sigma_8$. Here σ_8 corresponds to the rms fluctuation of mass within a top-hat sphere of $8 h^{-1}\text{Mpc}$ radius and it is treated as a normalization constant. The symbol σ_{FOG} accounts for an isotropic velocity dispersion due to the Fingers of God effect.

The comoving galaxy power spectrum contains cosmological information about galaxy clustering. In practice, we measure galaxy redshifts and angles and deduce cosmological distances from these instead of measuring clustering directly in comoving space. This approach relies on the use of the relevant cosmological model to convert redshift to cosmological distance. The use of erroneous or inaccurate cosmological models leads to distortions in the comoving clustering and incorrect measurement of distances. These cosmological distortions are manifested in the radial and the angular directions of the clustering signal. Such distortions are sensitive to the Hubble parameter ($1/H(z)$) and the angular diameter distance ($D_A(z)$) in the radial and the angular directions respectively. The Alcock-Paczyński (AP) test (Alcock & Paczyński 1979; López-Corredoira 2014) presents an approach where one tries to mitigate these distortions by adjusting the cosmological model. The primary advantage of the AP approach and the use of AP parameters is that they depend only on the geometry of the Universe. The parameter α is representative of the assessment of BAO from spherically averaged clustering measurements while the parameter ϵ illustrates the importance of the BAO feature in ξ_2 for a spherically symmetric sample. When a fiducial cosmological model is used to infer distances from the redshift, the separation between the observed and the expected positions of the BAO pertains to two dilation scales. The dilation scale which is parallel to the line of sight is denoted by α_{\parallel} while

the dilation scale perpendicular to the line of sight is represented by α_{\perp} . The parameters α_{\parallel} and α_{\perp} can be obtained from knowledge of the Hubble expansion rate, $H(z)$ and the angular diameter distance, $D_A(z)$ for the fiducial and model cosmologies.

$$\alpha_{\parallel} = \frac{H^{\text{fid}}(z)r_s^{\text{fid}}(z_d)}{H(z)r_s(z_d)}, \quad \alpha_{\perp} = \frac{D_A(z)r_s^{\text{fid}}(z_d)}{D_A^{\text{fid}}(z)r_s(z_d)} \quad (12)$$

Here, the superscript ‘fid’ refers to the value of a quantity in the fiducial cosmology and the parameter $r_s(z_d)$ denotes the fiducial sound horizon assumed in the power spectrum template. The parameter $r_s(z_d)$ sets the comoving BAO scale. It is also common to obtain the α_{\parallel} and α_{\perp} as derived parameters from α and ϵ using the following relationships:

$$\alpha = \alpha_{\parallel}^{1/3} \alpha_{\perp}^{2/3}, \quad 1 + \epsilon = \left(\frac{\alpha_{\parallel}}{\alpha_{\perp}} \right)^{1/3} \quad (13)$$

One can also use the volume averaged distance D_v and the AP-parameter F_{AP} to report measurements of the angular and radial projected distance scales. The volume averaged distance D_v is related to the redshift z , the speed of light c , the angular diameter distance D_A and the Hubble constant H by the following relation:

$$D_v = \left[(1+z)^2 cz \frac{D_A^2}{H} \right]^{1/3} \quad (14)$$

The AP-parameter F_{AP} is defined as:

$$F_{\text{AP}} = \frac{1+z}{c} D_A H \quad (15)$$

In the work presented in this paper, we allow the AP effect to change due to cosmology through the free parameters α and ϵ . This essentially accounts for error in cosmology, which is small, and hence, slightly inflates our distance measurement error bars. But, this is largely ignored in other analyses where the cosmology is kept fixed while allowing the AP parameters to be free. Also, we allow the shape of the power spectrum to vary using A_s and n_s which gives σ_8 indirectly. This is in contrast with the methods used in other analyses which fix A_s and n_s and vary σ_8 directly at lower redshifts. This could affect the measurement of growth rate (f) and σ_8 but $f\sigma_8$ is independent of this choice in the parameter space.

Once the choice of parameters is fixed, COSMOMC executes a MCMC algorithm to sample the parameter space and compute the linear power spectrum $P_{1\text{in}}(\mathbf{k})$ for the sampled points in the parameter space using the Code for Anisotropies in the Microwave Background (CAMB) (Lewis et al. 2000). CAMB takes the cosmological parameters of the model that we are working with as input. At each sampled point, CLPT uses the linear power spectrum to compute the velocity statistics and the two-point correlation function, $\xi(r)$. From this two-point correlation function, the Gaussian streaming model GSRSD calculates the redshift space two-point correlation function. This is followed by the rescaling of the two-point correlation function in relation to the difference in the fiducial and the current MCMC cosmologies

to obtain the correlation function corresponding to the theory, ξ^{model} . We follow the approach adopted in Marulli et al. (2012); Xu et al. (2012); Samushia et al. (2014); Alam et al. (2015a) to rescale the model redshift space correlation function to account for the cosmological distortions in the clustering signal.

$$\xi^{\text{RSD}}(s_{\parallel}, s_{\perp}) = \xi^{\text{model}}(\alpha_{\parallel} s_{\parallel}, \alpha_{\perp} s_{\perp}) \quad (16)$$

Here, s_{\parallel} represents the separation between two galaxies along the line of sight while s_{\perp} represents the separation between two galaxies perpendicular to the line of sight. The rescaling discussed in equation 16 is a result of the Alcock-Paczyński effect and it involves the use of the Alcock-Paczyński parameters which are discussed in equation 12. Corresponding to each set of cosmological parameters in the sampling process we get theoretical models for monopole and quadrupole using the aforementioned prescription. More details of the evaluation of multipoles from two point correlation functions using CLPT are given in section 4.4. A comparison of the multipoles obtained from the theory and the data yields the likelihood. Optimization of the obtained likelihood helps us arrive at the best fit model corresponding to a given data (section 4.5).

4.4 Correlation functions from CLPT-GSRSD

A comprehensive analysis of the chosen cosmological parameters will necessitate the comparison of the two point correlation functions obtained from data 4.1 with those obtained from theory. We use CLPT-GSRSD to evaluate the required theoretical two point correlation functions (ξ^{model}). We discuss the Convolution Lagrangian Perturbation Theory in section 2. Details of the incorporation of the Gaussian streaming model into this formalism is described in 2.2. We present a brief sketch of the CLPT-GSRSD model in section 2.3. Equation 2 encapsulates the details of the technique used to calculate the theoretical correlation functions. We rely heavily on the work presented in Wang et al. (2014) for this section.

4.5 Likelihood analysis

Concatanated combinations of the linearly independent constructs of monopoles (ξ_0) and quadrupoles (ξ_2) can be thought of as vectors. We assume the correlation function multipoles to be Gaussian distributed. Also, we ignore the parameter dependence of the covariance matrices in our analysis. Consequently, the check of the extent of correspondence between the data and the model vectors and the determination of the likelihood of the parameter values reduces to the computation of the χ^2 . Given a data vector ξ^{data} , a model vector ξ^{model} and the covariance matrix Σ , the χ^2 is obtained as

$$\chi^2 = \left(\xi^{\text{data}} - \xi^{\text{model}} \right) \Sigma^{-1} \left(\xi^{\text{data}} - \xi^{\text{model}} \right)^T \quad (17)$$

In our analysis $\xi^{\text{data}} = \left[\tilde{\xi}_0^{\text{data}}, \tilde{\xi}_2^{\text{data}} \right]$ and $\xi^{\text{model}} = \left[\tilde{\xi}_0^{\text{model}}, \tilde{\xi}_2^{\text{model}} \right]$ and Σ^{-1} is the inverse of the covariance matrix obtained from the MD-P mocks.

The RSD likelihood, \mathcal{L}_{RSD} of a parameter p is now given by: $\mathcal{L}_{\text{RSD}}(p) \propto e^{-\chi^2(p)/2}$

Cosmic Microwave Background (CMB) data from the Planck satellite puts constraints on the cosmological parameters. The Planck temperature anisotropy data from [Planck Collaboration et al. \(2014b\)](#) is used to compute the Planck likelihood. We use the covariance obtained from [Planck Collaboration et al. \(2014c\)](#). To get the full Planck likelihood, we use a multi-variate Gaussian approximation. This Gaussian approximation is close to the likelihood in the parameter space that we are doing our analysis in. The four cosmological parameters mentioned in equation given below, i.e. $\{\Omega_b h^2, \Omega_c h^2, n_s, \ln(10^{10} A_s)\}$ represent the likelihood really well and provide an excellent speed gain.

$$\Omega_b h^2 = 0.02207, \quad \Omega_c h^2 = 0.1196, \quad n_s = 0.9616,$$

$$\ln(10^{10} A_s) = 3.098,$$

$$\Sigma_{\text{Planck}} =$$

$$\begin{pmatrix} 1.089 \times 10^{-7} & -4.501 \times 10^{-7} & 1.365 \times 10^{-6} & 3.564 \times 10^{-6} \\ -4.501 \times 10^{-7} & 9.610 \times 10^{-6} & -2.215 \times 10^{-5} & 1.562 \times 10^{-5} \\ 1.365 \times 10^{-6} & -2.215 \times 10^{-5} & 8.836 \times 10^{-5} & 2.030 \times 10^{-5} \\ 3.564 \times 10^{-6} & 1.562 \times 10^{-5} & 2.030 \times 10^{-5} & 5.184 \times 10^{-3} \end{pmatrix}$$

We multiply the Planck likelihood with the RSD likelihood to get a joint constraint on the parameters in our analysis: $\mathcal{L}_{\text{total}}(p) = \mathcal{L}_{\text{Planck}}(p) \times \mathcal{L}_{\text{RSD}}(p)$.

For each set of cosmological parameters (p) we obtain $\chi^2(p)$. Choice of parameters corresponding to the minimum $\chi^2(p)$ leads to maximization of the likelihood $\mathcal{L}(p)$ and selection of the best fit model. This analysis is performed independently for data in all effective redshift bins. The results obtained from likelihood analysis of the data and the theory multipoles help us determine the optimized χ^2 and the best fit model at the chosen minimum fitting scale. This analysis forms an essential part of the research that we present in this work.

For all the calculations in our analysis we use $25 h^{-1}\text{Mpc}$ as the minimum fitting scale and $150 h^{-1}\text{Mpc}$ as the maximum fitting scale. The choice of the fitting scales is inspired by [Alam et al. \(2015a\)](#) and follows closely the results presented in there. We use evenly spaced bins of r with a bin size of $5 h^{-1}\text{Mpc}$. It's instructive to measure the deviation of the data vectors from the model vectors by measuring the χ^2/dof for each calculation, where dof pertains to the degree of freedom in the analysis. The degree of freedom is obtained from the knowledge of the bin size, minimum (r_{min}) and maximum (r_{max}) fitting scales of each multipole and the number of cosmological parameters (N_C) that are sampled by COSMOMC. For each instance of data (and theory) we use binned values of r from both the monopole and the quadrupole. Hence, the computation of the dof takes into account the numbers of bins in r in both the monopole and the quadrupole. We use the following equation to compute the degree of freedom: $\text{dof} = [(r_{\text{max}} - r_{\text{min}})/\text{bin size}] \times 2 - N_C$.

The obtained values of χ^2/dof give us an idea of the goodness of the fit of the best fit model with data.

5 RESULTS

In this section we discuss the results from mocks and the BOSS DR12 galaxy data.

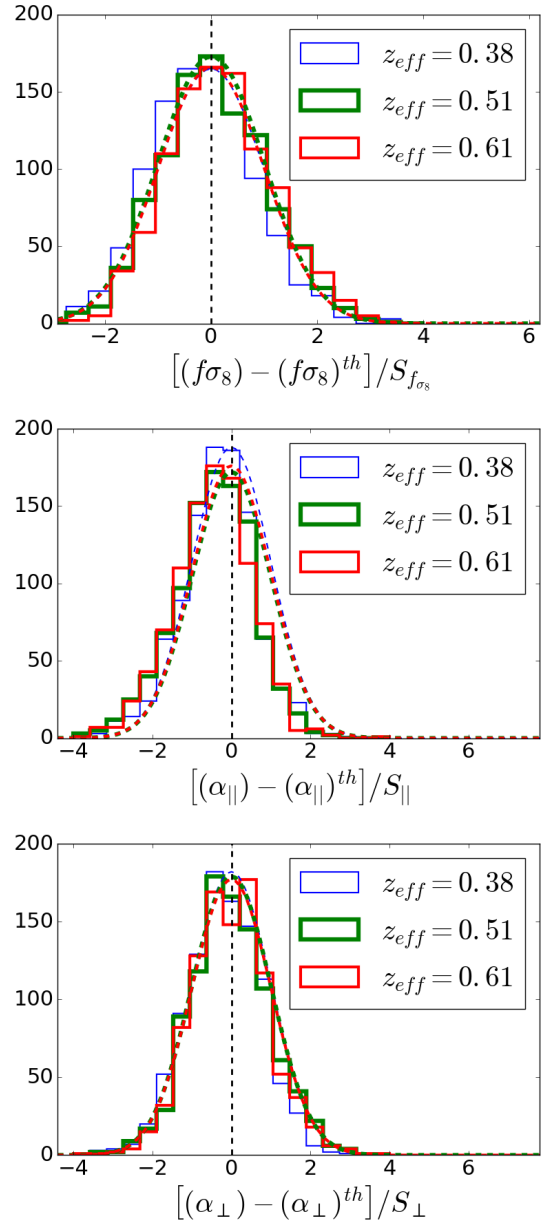


Figure 3. Histograms of the distributions of parameters obtained from the analysis of the MD-P mocks for the three bins. We independently fit the theory to the correlation function multipoles for each of the 997 mocks (in each bin) using the covariance matrices obtained from the mocks to obtain the statistics shown in these figures. The solid blue line represents the bin corresponding to $0.2 < z < 0.5$ whereas the solid green line depicts the bin for the redshift range $0.4 < z < 0.6$. The solid red line denotes the bin with the redshift range $0.5 < z < 0.75$. The dashed blue, green and red lines depict Gaussian functions with zero means, unit variances and heights equal to the heights of the histograms corresponding to $z_{\text{eff}} = 0.35, 0.51$ and 0.61 respectively. The symbols $S_{f\sigma_8}$, $S_{||}$ and S_{\perp} represent the standard deviations of the parameters $f\sigma_8$, $\alpha_{||}$ and α_{\perp} respectively. The x-axes of the three plots denote the ratio of the differences between the obtained parameters $f\sigma_8$, $\alpha_{||}$, α_{\perp} and their respective theoretical values $(f\sigma_8)^{\text{th}}$, $(\alpha_{||})^{\text{th}}$, $(\alpha_{\perp})^{\text{th}}$ with their respective standard deviations $S_{f\sigma_8}$, $S_{||}$, S_{\perp} . For *bin1*, we have $(f\sigma_8)^{\text{th}} = 0.484$, $(\alpha_{||})^{\text{th}} = 1.0031$, $(\alpha_{\perp})^{\text{th}} = 1.0008$. For *bin2*, we find $(f\sigma_8)^{\text{th}} = 0.483$, $(\alpha_{||})^{\text{th}} = 1.0040$, $(\alpha_{\perp})^{\text{th}} = 1.0010$. For *bin3*, we have $(f\sigma_8)^{\text{th}} = 0.477$, $(\alpha_{||})^{\text{th}} = 1.0046$, $(\alpha_{\perp})^{\text{th}} = 1.0012$.

Table 1. Statistics of the CLPT-GSRSD fits of MD-P mocks in the three redshift bins. Sampling parameters are the parameters which are sampled in the MCMC algorithm. Derived parameters represent the parameters which are obtained as functions of the sampling parameters. In each bin, we analyze different mocks using the prescription outlined in section 4 to obtain best fit values and errors for chosen cosmological parameters for each mock. From the statistics obtained for all the mocks, we compute the mean (denoted by $\langle \cdot \rangle$) and the standard deviations (S) of the parameters.

Redshift (z_{eff})	Sampling Parameters											
	$\langle \alpha \rangle$	S_α	$\langle \epsilon \rangle$	S_ϵ	$\langle f \rangle$	S_f	$\langle F_1 \rangle$	S_{F_1}	$\langle F_2 \rangle$	S_{F_2}	$\langle \sigma_{\text{FOG}} \rangle$	$S_{\sigma_{\text{FOG}}}$
0.38	0.990	0.020	-0.0046	0.0252	0.720	0.102	0.962	0.074	0.77	0.78	3.26	0.90
0.51	0.993	0.017	-0.0069	0.0222	0.793	0.098	1.020	0.072	0.60	0.79	3.56	0.89
0.61	0.993	0.016	-0.0071	0.0220	0.833	0.099	1.112	0.069	0.49	0.84	3.87	0.98
	Derived Parameters											
	$\langle \alpha_{\parallel} \rangle$	S_{\parallel}	$\langle \alpha_{\perp} \rangle$	S_{\perp}	$\langle f\sigma_8 \rangle$	$S_{f\sigma_8}$						
0.38	0.984	0.058	0.995	0.028	0.472	0.067						
0.51	0.981	0.050	1.000	0.024	0.487	0.060						
0.61	0.981	0.048	1.001	0.025	0.487	0.058						

5.1 Correlation function results on the challenge mocks

In order to check the correspondence between the different modeling and measurement techniques which have been used in the analysis of the BOSS DR12 combined sample galaxy dataset (including the methods that are combined to obtain the final consensus constrains in Alam et al. (2016)), the different techniques were tried on large-volume synthetic catalogs in a RSD-fit challenge and compared with each other (Tinker et al. 2016). In addition to checking the agreement of the cosmological information extracted from the different full-shape approaches, this exercise also served to check for possible systematics. We refer the readers Tinker et al. (2016) to see more details of this RSD data challenge.

Seven different HOD galaxy samples which are built from large-volume N -body simulations are investigated in the first part of the challenge. The aforesaid simulations are in tune with Λ CDM cosmology with moderately different density parameters. The difference between the recovered cosmological parameters and true cosmological parameters were smaller than the statistical error bars for all the methods. The fact that methods based on both the configuration and the momentum spaces show great accuracy and agreement in the constraints that were obtained in the challenge catalogs is very encouraging.

The next part of the challenge involved a set of 84 synthetic catalogs which mimic the North Galactic Coordinate (NGC) part of the DR12 CMASS galaxies (‘cut-sky’ mocks). These are called the ‘N series’ samples. All of these mocks are obtained using a standard HOD model and are generated from N -body simulations which assume the same set of cosmological parameters. As a result, these mocks serve to check for systematic biases in the obtained parameter constraints. In our analysis for the N series mocks, the biases in our recovered parameters were much smaller than the statistical error bars.

5.2 Results from MD-P mocks

In Figure 2, we present the correlation matrices obtained from the three redshift bins of the MD-P Patchy mocks. In each plot in Figure 2, the upper right corner shows the corre-

lation between the bins in the monopole and the quadrupole, the upper left corner shows the correlation between the bins in the quadrupole, the lower right corner represents the correlation between bins in the monopole while the lower left corner illustrates the correlation between the bins in the quadrupole and the monopole.

We independently fit the theory to the correlation function monopoles and quadrupoles for each of the 997 mocks using the covariance matrices computed from the MD-P mocks (for the three redshift bins). The best-fit results are obtained from the comparison of the correlation function multipoles from the mocks with the corresponding multipoles obtained from CLPT-GSRSD theory. The multipoles are binned in evenly spaced values of s with a bin size of $5 h^{-1}\text{Mpc}$. The tally of data versus theory is done over a fitting range $25 h^{-1}\text{Mpc} \leq s \leq 150 h^{-1}\text{Mpc}$. The statistics for the parameters are obtained from the analysis of 997 mocks in each redshift bin. Table 1 encapsulates the mean and the standard deviations of the best-fit results of the six parameters that have been sampled by COSMOMC, *viz.* $\{F_1, F_2, f, \sigma_{\text{FOG}}, \alpha, \epsilon\}$. We also catalog the statistics of various derived parameters ($f\sigma_8, b\sigma_8, D_A$ and H) which are obtained as functions of the sampling parameters in Table 1.

The values that we get for $f\sigma_8, \alpha_{\parallel}$ and α_{\perp} from our analysis of the MD-P mocks (which are given in Table 1) tally satisfactorily with the corresponding theoretical values of the parameters (which are given in the caption of Figure 3). Figure 3 illustrates the statistical spread of the derived parameters $\alpha_{\parallel}, \alpha_{\perp}$ and $f\sigma_8$ for the three bins in the form of histograms. In Figure 3 we show the distribution of the ratios of the differences between the derived parameters $\alpha_{\parallel}, \alpha_{\perp}, f\sigma_8$ and their respective theoretical values $(f\sigma_8)^{th}, (\alpha_{\parallel})^{th}, (\alpha_{\perp})^{th}$ with their respective standard deviations $S_{f\sigma_8}, S_{\parallel}, S_{\perp}$. The statistics shown in Figure 3 are obtained from the fitting of the multipoles of each of the 997 mocks (in each redshift bin) with multipoles from the theory.

5.3 Results for BOSS DR12 Galaxy Data

We follow the steps outlined in section 4 to obtain optimized values of cosmological parameters for the three redshift bins

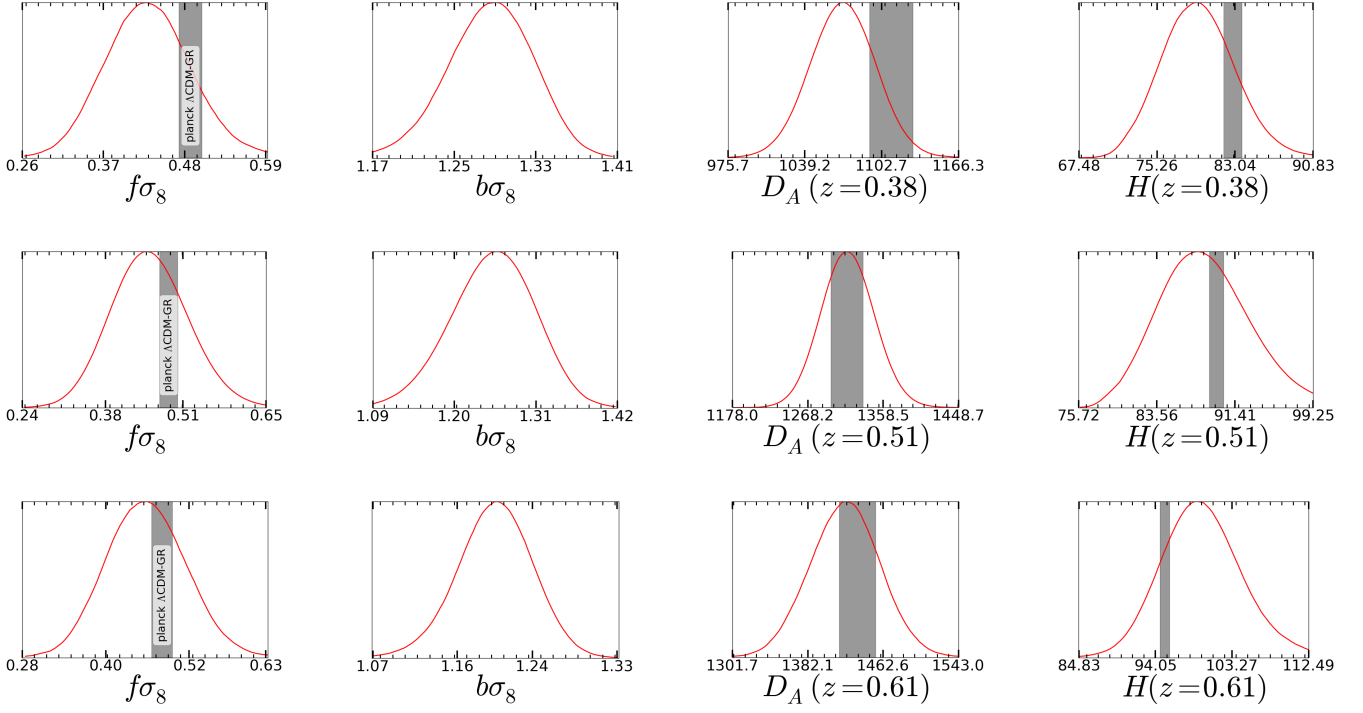


Figure 4. One dimensional marginalized likelihoods for the parameters $\{f\sigma_8, b\sigma_8, D_A, H\}$ for the three redshift bins. The top row depicts results of one dimensional marginalized likelihoods of parameters for the redshift bin with $0.2 < z < 0.5$, the middle row represents results of one dimensional marginalized likelihoods of parameters for the redshift bin with $0.4 < z < 0.6$ while the bottom row represents results of one dimensional marginalized likelihoods of parameters for the redshift bin with $0.5 < z < 0.75$. The grey shaded regions represent 1σ spreads of the Planck Λ CDM predictions of the parameters.

Table 2. These are the results of the constraints for parameters in our analysis obtained from the BOSS DR12 dataset analysis. In our analysis we have used the fitting range $25 h^{-1}\text{Mpc} \leq s \leq 150 h^{-1}\text{Mpc}$ with a bin size of $5 h^{-1}\text{Mpc}$. For each parameter, we report the prior range, mean and 1σ error. Sampling parameters are parameters which are a part of the parameter space which is sampled by COSMOMC. Derived parameters are parameters which are obtained as functions of the sampling parameters.

Parameter	prior range	<i>Bin1</i> ($z_{\text{eff}} = 0.38$)	<i>Bin2</i> ($z_{\text{eff}} = 0.51$)	<i>Bin3</i> ($z_{\text{eff}} = 0.61$)
Sampling Parameters				
$\Omega_b h^2$	[0.02042, 0.02372]	0.02207 ± 0.00026	0.02206 ± 0.00026	0.02208 ± 0.00026
$\Omega_c h^2$	[0.1041, 0.1351]	0.11947 ± 0.00089	0.11953 ± 0.00086	0.11963 ± 0.00086
n_s	[0.9146, 1.009]	0.9605 ± 0.0058	0.9616 ± 0.0059	0.9621 ± 0.0058
$\ln(10^{10} A_s)$	[2.670, 3.535]	3.076 ± 0.073	3.078 ± 0.070	3.089 ± 0.068
α	[0.8000, 1.350]	0.991 ± 0.016	1.006 ± 0.016	0.980 ± 0.016
ϵ	[-0.5000, 0.5000]	0.0275 ± 0.0183	0.0045 ± 0.0189	-0.0111 ± 0.0191
$f = d \ln(D)/d \ln(a)$	[0.1000, 1.200]	0.638 ± 0.080	0.715 ± 0.090	0.753 ± 0.088
σ_{FOG}	[0, 20]	3.21 ± 1.68	3.19 ± 1.80	2.65 ± 1.53
F_1	[0.5, 1.5]	0.91 ± 0.09	0.98 ± 0.10	0.98 ± 0.09
F_2	[-5.0, 5.0]	-0.47 ± 0.31	-0.2 ± 1.48	-0.51 ± 0.32
Derived Parameters				
$f\sigma_8$	x	0.430 ± 0.054	0.452 ± 0.057	0.457 ± 0.052
$b\sigma_8$	x	1.283 ± 0.039	1.249 ± 0.051	1.198 ± 0.036
D_A	x	1069.6 ± 24.0	1314.5 ± 27.4	1420.6 ± 33.4
H	x	79.3 ± 3.2	88.4 ± 4.1	99.5 ± 4.4

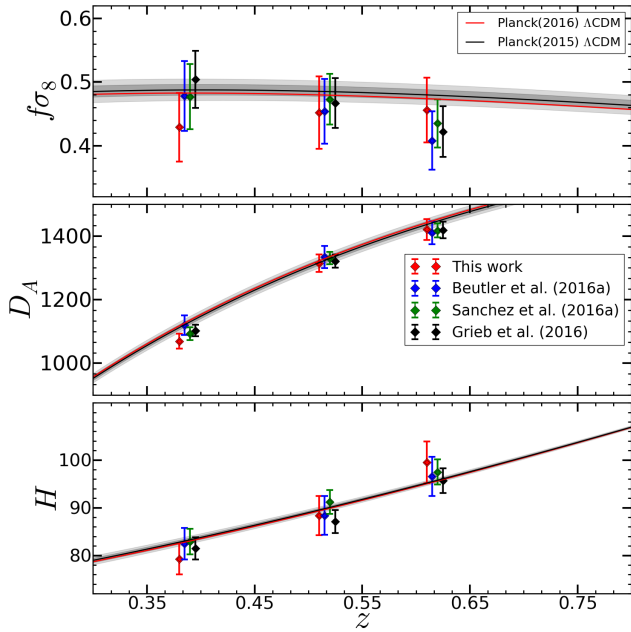


Figure 5. Here we compare our results of $f\sigma_8(z)$, $D_A(z)$ and $H(z)$ with the predictions of Planck Λ CDM and with the results of Beutler et al. (2016a), Grieb et al. (2016) and Sánchez et al. (2016a). The dark and the light shaded regions represent the 1σ and the 2σ spreads of the Planck Λ CDM (TTTEEE+lowP) predictions of $f\sigma_8$, D_A and H . The solid black line shows the Planck Collaboration et al. (2015) predictions for the variation of $f\sigma_8$ as a function of z while the solid red line shows the predictions of Planck Collaboration et al. (2016) for the variation of $f\sigma_8$ with respect to z .

of the BOSS DR12 galaxy dataset. At this stage, it is worthwhile to note that in our analysis of the BOSS DR12 galaxy data we use the covariance matrices obtained from the MD-P mocks. In Table 2, we list the parameters used in our analysis and the best-fit results returned by COSMOMC for the parameters for the three galaxy bins ($z_{\text{eff}} = 0.38, 0.51$ and 0.61). Our results for the parameters $f\sigma_8$, $b\sigma_8$, D_A and H for bins 1, 2 and 3 agree with reports of results of parameters from efforts by other groups (Alam et al. 2016).

Figure 1 illustrates the monopole and the quadrupole for BOSS DR12 data and the best fit models of the data for the three bins. The black dots in the plots of Figure 1 represent the measured correlation functions for BOSS DR12 galaxy sample. The diagonal elements of the covariance matrices corresponding to mocks in the three redshift bins give us the error bars for our observations in the three bins. As indicated in the plots, the red and the blue lines denote the best fits of monopole and quadrupole respectively using the fitting range $25 h^{-1}\text{Mpc} \leq s \leq 150 h^{-1}\text{Mpc}$ with a bin size of $5 h^{-1}\text{Mpc}$. Our results show that the means of the best-fit models for the monopole and the quadrupole tally with the multipoles from the data very well. This is evidenced by the values of χ^2/dof (53/44, 52/44, 41/44) that we obtain for the comparison of multipoles of data and theory of the three redshift bins. The values of χ^2/dof are as per expectations. Figure 4 shows the results of the marginalized likelihood of the derived parameters $\{f\sigma_8, b\sigma_8, D_A, H\}$.

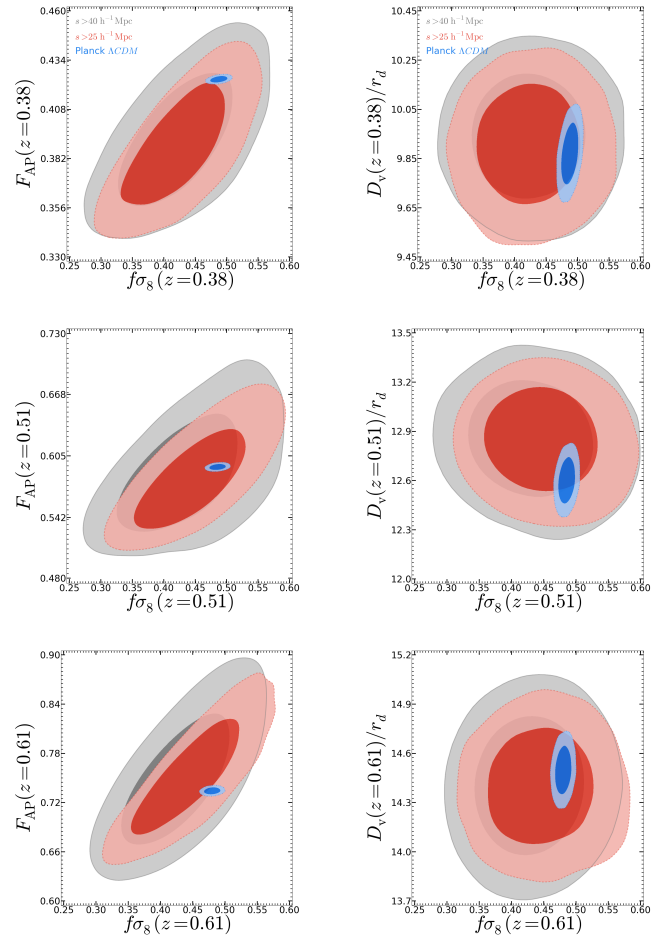


Figure 6. In this figure we show 1σ (68%) and 2σ (95%) confidence intervals for F_{AP} and $f\sigma_8$ on the left and D_V/r_d and $f\sigma_8$ on the right. The grey contours show constraints from larger scales ($s > 40 h^{-1}\text{Mpc}$) while the red contours depict constraints from all scales ($s > 25 h^{-1}\text{Mpc}$) from our RSD analysis. The blue contour shows the constraints from Planck 2015 results. The top, middle and the bottom rows are from the three redshift bins, *viz.* $z_{\text{eff}} = 0.38, 0.51$ and 0.61 respectively. All estimates are mutually consistent.

6 DISCUSSION

Our measurements of $f\sigma_8$, D_A and H using CLPT-GSRSD are consistent with the predictions of the Λ CDM for all the three redshift bins. Furthermore, our results for $f\sigma_8$, D_A and H obtained from CLPT-GSRSD agree very well with measurements of the same parameters obtained from other approaches which do investigations based on full-shape analyses of SDSS-III BOSS DR12 combined sample (Beutler et al. 2016a; Grieb et al. 2016; Sánchez et al. 2016a). At the same time, the theoretical models used by the four analyses are very different. A comparison of the different results is presented in Figure 5. The black and the red lines in the plot of $f\sigma_8$ vs z in Figure 5 show the predictions of Planck Collaboration et al. (2015) and Planck Collaboration et al. (2016) respectively. The difference between the predictions of Planck 2015 and Planck 2016 is not more than $1/2 \sigma$. The correspondence of results of $f\sigma_8$ at multiple redshifts which are obtained from different theories can be considered

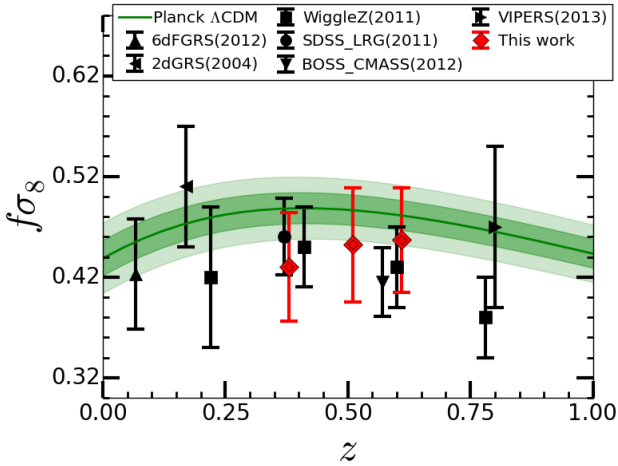


Figure 7. Here, we plot measurements of $f\sigma_8$ from different surveys and research studies. The included surveys and studies report measurements of $f\sigma_8$ over a redshift range of $0.06 < z < 0.80$. We have represented the 1σ and the 2σ spreads of the Planck Λ CDM prediction for the evolution of $f\sigma_8$ with redshift in the dark green and the light green shaded regions respectively.

as a useful probe of the theory of gravity. It also holds the promise of letting us place model independent constraints on other models of gravity.

One of the challenges in RSD analysis is to use the smaller scales as they have higher signal to noise by virtue of sampling large numbers of two point modes. But, perturbation theory based models find it difficult to describe measurements at smaller scales due to non-linear clustering. The model used in our analysis has been validated using various approximate mocks and N-body mocks. We fit to scales upto $25 h^{-1}$ Mpc in our final analysis. In order to understand the contribution from quasi-linear scales and to further look for biases in our analysis we have also run our analysis using a linear scale of $s > 40 h^{-1}$ Mpc. Figure 6 shows the comparison between results obtained from fitting only linear scale and results obtained while including the quasi linear scales. In Figure 6, we show 1σ (68%) and 2σ (95%) confidence intervals for F_{AP} and $f\sigma_8$ on the left and D_V/r_d and $f\sigma_8$ on the right. The grey contours show constraints from larger scales ($s > 40 h^{-1}$ Mpc) while the red contours depict constraints from all scales ($s > 25 h^{-1}$ Mpc) from our RSD analysis. The blue contour shows the constraints from Planck 2015 results. The top, middle and the bottom rows are from the three redshift bins, *viz.* $z_{\text{eff}} = 0.38, 0.51$ and 0.61 respectively. The inclusion of the quasi-linear scale improves the constraints without introducing any statistically significant shifts in the measurements. The improvement in F_{AP} and $f\sigma_8$ is larger compared to D_V/r_d because most of the information in D_V/r_d is contained in the BAO peak.

Figure 7 presents a compilation of $f\sigma_8$ measurements at different redshifts from different surveys and research studies. We expect our results to provide a robust test of the underlying theory of gravity at large distance scales.

We decided not to push for measurements of the linear growth rate of structure at fitting scales smaller than the minimum fitting scale that we have chosen here because of the lack of reliability in the behavior of the model at small

distance scales. From the perspective of a comprehensive RSD analysis it will be invaluable to gain knowledge of estimates of cosmological parameters at distance scales smaller than $20 h^{-1}$ Mpc. However, an analysis of smaller cosmological scales with presently available theoretical resources presents a significant challenge. The theoretical models that are currently available to us are unable to model small distance scales. It is worthwhile to investigate if this inability is due to the presence of the non-linear Finger of God effect. Are the contributions from non-linear clustering (which are different from the Fingers of God effect) modeled accurately? These are questions that we wish to seek answers to. We would like to investigate the efficacy of the Gaussian Streaming Model in explaining non-linear clustering at small scales. As an outlook for the future, we also plan to explore the feasibility of designing and using new estimators to probe scales smaller than $20 h^{-1}$ Mpc and also test the effectiveness of CLPT-GSRSD at such scales.

7 SUMMARY

We have used CLPT-GSRSD to measure cosmological parameters including the linear growth rate of structure f from the SDSS III BOSS DR12 combined galaxy sample. The BOSS DR12 combined galaxy dataset includes over a million massive galaxies encompassing a redshift range $0.2 < z < 0.75$. We divide this sample to three partially overlapping redshift bins with effective redshifts of 0.38, 0.51 and 0.61 and we work with multipole moments of two-point galaxy correlation functions in these three redshift bins. We use the measured and best fit multipole moments to place constraints on cosmological parameters including the linear growth rate of structure in the Universe. The fitting scale that we choose in this work is dictated by the performance, reliability and considerations of error of MD-P mocks and our theoretical model at small scales. Our measurements of the growth rate of structure, $f\sigma_8(z)$, angular diameter distance $D_A(z)$ and the Hubble expansion rate, $H(z)$ are in agreement with the results for the same parameters obtained by different groups (Beutler et al. 2016a; Grieb et al. 2016; Sánchez et al. 2016a). Furthermore, our results are combined with other BAO (Beutler et al. 2016b; Ross et al. 2016; Vargas Magaña et al. 2016) and full-shape methods in a set of final consensus constraints in Alam et al. (2016). Our results are in consonance with the predictions of the Planck Λ CDM model. We expect the results of our work to shed more light on the evolution of the linear growth rate of structure and contribute towards lifting the ambiguity in the choice between dark energy and modified theories of gravity. The measurements we report in this work can contribute to constrain cosmological parameters in different models of gravity. Through our work, we also provoke questions of whether it is possible to model non-linearities at small distance scales in the Universe.

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