The cosmological analysis of the SDSS/BOSS data from the Effective Field Theory of Large-Scale Structure

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Abstract. The Effective Field Theory of Large-Scale Structure is a formalism that allows us to predict the clustering of Cosmological Large-Scale Structure in the mildly non-linear regime in an accurate and reliable way. After validating our technique against several sets of numerical simulations, we perform the analysis for the cosmological parameters of the DR12 BOSS data. We assume ΛCDM, a fixed value of the baryon/dark-matter ratio, Ω_b/Ω_c, and of the tilt of the primordial power spectrum, n_s, and no significant input from numerical simulations. By using the one-loop power spectrum multipoles, we measure the primordial amplitude of the power spectrum, A_s, the abundance of matter, Ω_m, and the Hubble parameter, H_0, to about 13%, 3.2% and 3.2% respectively, obtaining ln(10^{10} A_s) = 2.72 ± 0.13, Ω_m = 0.309 ± 0.010, H_0 = 68.5 ± 2.2 km/(s Mpc) at 68% confidence level. If we then add a CMB prior on the sound horizon, the error bar on H_0 is reduced to 1.6%. These results are a substantial qualitative and quantitative improvement with respect to former analyses, and suggest that the EFTofLSS is a powerful instrument to extract cosmological information from Large-Scale Structure.

Keywords: cosmological parameters from LSS, power spectrum, redshift surveys, baryon acoustic oscillations

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1 Introduction

General remarks. In the past few decades, cosmology has evolved from a highly speculative science to a highly precise one, where percent-level error bars are the norm. This has happened thanks to our capability to measure the primordial density fluctuations first in the Cosmological Large-Scale Structure (LSS), and then, dominantly, in the Cosmic Microwave Background (CMB). Such precision in the measurements has allowed cosmology to contribute to the exploration of the fundamental laws of Nature, that is, those laws from which any other law can be derived.

In order to continue this splendid journey, it is essential that we continue to measure new primordial fluctuations. The next leading source of cosmological information will come from LSS, which will be observed either directly, or indirectly through CMB lensing. There is however a tremendous challenge from extracting information from LSS: at short distances the fluctuations are large and the dynamics is extremely complicated, with stars and galaxies forming and exploding. Because modes are coupled at non-linear level, these uncontrolled short-distance non-linearities affect the long-wavelength fluctuations. This makes the theoretical modeling very challenging, so that, in practice, either only very-long wavelength data are analyzed, as in this regime the corrections from non-linear scales become either negligible or easy to model, or also short-wavelength data are analyzed, but at the cost of adding potentially large, and often uncontrolled, systematic uncertainties. These limitations come with a huge cost in terms of information that we can extract from LSS data. In fact, in everything that we measure in cosmology, with very few exceptions, information comes from measuring a high number of modes, $N_{\text{modes}}$, i.e. the error bars scale as $1/\sqrt{N_{\text{modes}}}$. The number of modes scales like $N_{\text{modes}} \propto k_{\text{max}}^3$, where $k_{\text{max}}$ indicates the highest wavenumber where the theory prediction can be trusted, and, as long as the modes are not order one nonlinear, these modes are essentially uncorrelated.\footnote{There is a well-spread argument that suggests that the information does not scale as $N_{\text{modes}} \propto k_{\text{max}}^3$. It is often left implicit to state that one is usually referring to the information contained within the two-point function. This is true in principle, but in practice, as long as non-linearities are perturbatively small, information is spread hierarchically to the higher $N$-point functions, which can also be analyzed, as we will show later. Something even simpler is also true: in the same mildly non-linear regime, the non-linearities represent a small correction, and therefore most of the information is still in the two-point function, and it is still scaling as $k_{\text{max}}^3$. Of course this discussion does not deal with the need to include parameters in our predictions that account for uncontrolled physics and that therefore could in practice limit the amount of information available. We will discuss this explicitly later on.}

By dropping the highest wavenumbers, we drop most of the information.

The EFTofLSS. There are several ways in which we might try to make progress. One is the use of numerical simulations that can simulate the formation of galaxies at all wavelengths observed by the experiments. This appears to be a prohibitive task. In practice, a more realistic approach is to try to simulate dark matter dynamics exactly, and then model in some physically-motivated and data-guided way galaxy formation physics. Doing this at all wavelengths of interest appears again to be very hard, but there is hope that one can do this at long-enough wavelengths. This is a potentially fruitful approach, it has delivered many important results, but, so far, it has allowed to analyze only the data up to a somewhat limited $k_{\text{max}}$. It should definitely be pursued more, but it is not the approach we use here.

The approach we use here is instead much less ambitious. Instead of using all the modes that are measured by a survey, we try to analyze the ones for which the non-linearities are non-negligible, but still smaller than order one. Though this might be a humble aim, it has
the potential to bring us tremendous results. In fact, if we denote the modes for which non-linearities are negligible as the linear modes, $N_{\text{lin}}$, and the ones for which they are sizable but still less than order one, as ‘mildly non-linear’, $N_{\text{mild, non-lin}}$, and if we also assume that, roughly, at redshift zero, the validity of linear theory is up to $k_{\text{lin}}^{\text{max}} \sim 0.1h \text{ Mpc}^{-1}$, while the mildly non-linear regime is valid up to $k_{\text{mild, non-lin}}^{\text{max}} \sim 0.5h \text{ Mpc}^{-1}$, then we have that the $N_{\text{mild, non-lin}} \sim 100 \cdot N_{\text{lin}}$: even within our extremely-rough approximations, the number of modes in the mildly non-linear regime is much larger than the linear ones.

The idea behind our approach is the following. If we restrict the analysis to wavenumbers which are in the mildly non-linear regime, then it is possible to write the equations of motion for the fluctuations, within which the effect of short-distance non-linearities is parametrized in the most general way. These equations are therefore correct, i.e. cannot be wrong, unless some basic principle of physics, such as Lorentz invariance, matter conservation, etc., are violated, because they are the most general ones compatible with the aforementioned principle of physics. In fact, given a particular galactic process, there are coefficients in the equations of motion whose numerical value can be adjusted to exactly reproduce the effect of that specific galactic process at long wavelength. When we apply this construction, we obtain what is called the Effective Field Theory of Large-Scale Structure (EFTofLSS) [1, 2].

Let us give a rough overview of the EFTofLSS, referring the reader to some lectures notes and some relevant literature for more detailed information. In the EFTofLSS, we derive some equations of motion for the long-wavelength fields: dark matter density [1–3] and velocity [4, 5], baryons [6], neutrinos [7], dark energy [8], etc., which, as mentioned just above, contain some terms with coefficients that are unknown and depend on the short-distance physics. In the jargon of the EFTofLSS, these coefficients are named ‘counterterms’. Just to give a rough idea of these equations, if the universe is filled only with dark matter, with $\delta_\ell$ being the long-wavelength overdensity, $v_\ell$ the long-wavelength velocity and $\phi_\ell$ the long-wavelength Newtonian potential, these equations read approximately

$$
\dot{\delta}_\ell + \frac{1}{a} \dot{\ell} ((1 + \delta_\ell)v^i_\ell) = 0 ,
$$

$$
\ddot{v}^i_\ell + Hv^i_\ell + \frac{1}{a} \nabla^j_\ell \nabla_j_\ell + \frac{1}{a} \partial_i \phi_\ell = \int^{t'} dt' \left[ c_{s,1}(t,t') \delta_\ell(x_\ell, t, t'), t') \right. 
\left. + c_{s,2}(t,t') \delta_\ell(x_\ell, t, t'), t')^2 + \ldots \right] ,
$$

where $x_\ell(\vec{x}, t, t')$ represents the location at time $t'$ of a fluid element that at time $t$ is at position $\vec{x}$, and where $\ldots$ represents in principle an infinite number of terms. $c_{s,1}$, $c_{s,2}$, $\ldots$ represent the unknown coefficients that encode the effect of short distances at long distances.

The resulting equations are solved perturbatively in powers in the fluctuations. Schematically, we write

$$
\delta_\ell(\vec{k}, t) = \sum_n \delta_\ell^{(n)}(\vec{k}, t) ,
$$

where $\delta_\ell^{(n)}$ is of order $(\delta_\ell^{(1)})^n$, with $\delta_\ell^{(1)}$ being the solution to the linearized equations, and we solve eq. (1.1) to a given order in $n$, assuming $\delta_\ell^{(1)} \ll 1$. Importantly, at a given perturbative

\[2\] The fact that this is possible to do should not be surprising. If we think of the dielectric Maxwell equations, we know that the effect on the propagation of long-wavelength photons of any atomic structure of the dielectric material, no matter how complicated that could be, is described by a proper numerical value of the dielectric constants.

\[3\] See EFTofLSS repository.
order, only a finite number of counterterms needs to be kept, as, roughly, each counterterm starts contributing at a specific order in $n$.

The perturbative solution allows us to compute correlation functions of long-wavelength fields that depend only on a finite number of unknown coefficients. Then, long-wavelength fluctuations of galaxies are described as biased tracers of the long-wavelength fluctuations of these fields. Schematically, if $\delta_{\ell,g}$ is the long-wavelength overdensity of galaxies, we have [9]

$$
\delta_{\ell,g}(\vec{x}, t) = \int dt' \left[ \tilde{c}_1(t, t') \delta_{\ell}(x_{\parallel}(\vec{x}, t, t'), t') + \tilde{c}_2(t, t') \partial_i v_i^l(x_{\parallel}(\vec{x}, t, t'), t')^2 + \tilde{c}_3(t, t') \delta_{\ell}(x_{\parallel}(\vec{x}, t, t'), t')^2 + \ldots \right].
$$

This means that correlation functions of galaxies can be perturbatively computed as linear combination of more complicated correlation functions of matter density and velocity fields, weighted by unknown coefficients that are named ‘biases’ (we will also use the name ‘counterterms’ or ‘EFT parameters’). Again, at a given order $n$ in the perturbative expansion, only a finite number of biases should be included.

Finally, the overdensity of galaxies in redshift space, $\delta_{\ell,g,r}$, is given in terms of the galaxy fields in real space by a formula that, roughly, reads [10]

$$
\delta_{\ell,g,r}(\vec{k}, t) = \delta_{\ell,g}(\vec{k}, t) - \frac{i k_z}{aH} v_{\ell,g}(\vec{k}, t) + \frac{i^2}{2} \left( \frac{k_z}{aH} \right)^2 \left[ v^2_{\ell,g}(\vec{x}, t) \right]_{\vec{k}} + \ldots \tag{1.4}
$$

where $\delta_{\ell,g}(\vec{k}, t)$ is the Dirac three-dimensional $\delta$-function, $\left[ \ldots \right]_{\vec{k}}$ means that we take the Fourier transform at momentum $\vec{k}$ of the quantity inside the brackets, and $\hat{z}$ is the direction of the line of sight. $k_{\text{NL}}$ is the wavenumber associated to the non-linear scale, i.e. where $k^3 \delta^{(1)}(k)^2 \big|_{k \to k_{\text{NL}}} \sim 1$. Again, there is only a finite number of unknown numerical coefficients that needs to be used for the calculation at a given order $n$.

Therefore we can schematically summarize the full perturbative structure as follows. Correlation functions of the density of galaxies in redshift space are obtained as linear combination of correlation functions of galaxy density and velocities in configuration space, which, in turn, are obtained as linear combinations of correlation functions of long-wavelength matter fields satisfying eq. (1.1). These correlation functions are computed in a perturbative expansion in powers of $\delta^{(1)}_{\ell} \ll 1$, or, equivalently, in powers of $k/k_{\text{NL}} \ll 1$, with $k$ being the wavenumber of interest. For a given order $n$ used in perturbation theory, only a finite, albeit by now potentially large, number of free coefficients is used.

Therefore, the EFTofLSS is a perturbation theory for the LSS. While there is a plethora of perturbative approaches to LSS, dating back to the time of Zeldovich [11], it should be stressed that there can be only one correct perturbative approach. By this we mean that there is only one correct set of equations that describes the long-wavelength behavior of the universe. Others can be correct only either if they are just a change of variables of the same equations, and so essentially the same equations, or if they solve the same equations expanding in different subsets of parameters. The novelty of the EFTofLSS with respect to the other perturbative approaches is that it has free coefficients that encode the effects of...
short distance physics at long distances. Their numerical value is unknown to the EFTofLSS and must be provided by either a direct fit to the data, or by some knowledge of the short distance physics, i.e. of the galaxy formation physics. But it is thanks to these coefficients that the EFTofLSS can be made arbitrarily accurate\(^4\) at long distances. Explicitly, for a given order \(n\), the error associated with the lack of inclusion of the higher order terms scales as \((k/k_{NL})^{\gamma \cdot n}\), with \(\gamma\) being an order one number. This is known as systematic theoretical error [2, 4, 10, 12, 13]. So, for a given required precision and a given wavenumber, one has an idea, albeit rough, of what order \(n\) to choose and how large the mistake of the prediction can be. It is the presence of the unknown numerical coefficients that makes these bold-sounding claims correct, though it comes at the cost of harming at some level the predictivity of the theory. But these coefficients are as non-vanishing as the viscosity for honey is a non-vanishing coefficient. One could use a more predictive theory to simulate honey’s motion: a fluid without viscosity. The theory would be more predictive, as it would have one less parameter to measure from experiment, but would be wrong. This is why the EFTofLSS should be used.

Though the EFTofLSS provides the only correct set of equations for a perturbative approach to the large-scale clustering of LSS, it should be thought of as the result of an evolution of many attempts to formulate a consistent set of equations and a consistent perturbative approach to solve them and predict the clustering of LSS. Here we name, as an example, a few of the perturbative approaches that were developed prior to the EFTofLSS and that were essential for the development of the EFTofLSS, and we list a very-small sample of the relevant references: Zeldovich approximation [11], SPT [14, 15], LPT [16], RPT [17], MPTbreeze [18, 19], RegPT [20], CLPT [21], combined Perturbation Theory, Halo Model and Simulations [22], etc.

**Application to data.** In this paper, we apply the EFTofLSS to the cosmological data. More specifically, we compare the EFTofLSS prediction for the monopole and quadrupole of the power spectrum of galaxies in redshift space at one-loop order (i.e. at perturbative order \(n = 3\)), and for the monopole of the bispectrum of galaxies in redshift space at tree level (i.e. \(n = 2\)), to the DR12 sample of galaxies as measured in the BOSS/SDSS data [23–26]. We use such a comparison to infer a measurement of each of the following cosmological parameters: the amplitude of the primordial power spectrum, \(A_s\), the fraction of matter of the universe, \(\Omega_m\), and the present value of the Hubble constant, \(h\). For convenience, we will work in \(\Lambda\)CDM cosmology and fix the tilt \(n_s\) and the ratio of baryons versus dark matter. A reason of concern for the predictive power of the EFTofLSS is the potentially-large number of parameters its predictions depend on. Nevertheless, we believe our results show that very substantial qualitative and quantitative gains in extracting information from the LSS data can be achieved by employing the EFTofLSS in an analysis.

This paper is the completion of a long journey for the EFTofLSS, which started from the initial development of the theory [1–3] and involved many important steps in order for it to be compared to the data. Here we mention a few of them in order to give a broad orientation within the relevant literature to the interested reader. The dark matter power spectrum has been computed at one-, two- and three-loop orders and compared with simulations in [2, 4, 27–35]. This also involved some additional theoretical developments, such as a careful understanding of renormalization [2, 36, 37] (including some rather-subtle aspects such as lattice-running [2]) and of the way of extracting the numerical value of the counterterms from simulations [2, 38], of the non-locality in time [4, 9, 28], and of some

\(^4\)In the sense of asymptotic series such as Quantum Electrodynamics.
subtleties with the velocity field [4, 5]. These theoretical explorations also include a study in 1+1 dimensional spacetime [38]. In order to correctly describe the baryon acoustic oscillation (BAO) peak, an IR-resummation of the long displacement fields had to be performed. This has led to the so-called IR-resummed EFTofLSS [10, 39–42]. A method to describe baryons and baryonic effects has been developed in [6], and was shown to work extremely well when compared with simulations. Concerning the computation and comparison with simulations of higher \( n \)-point functions, the dark-matter bispectrum has been computed at one-loop in [43, 44] (and compared with simulations), the one-loop trispectrum in [45], the dark matter displacement field in [46] (and compared with simulations). Understanding the dark matter dynamics has allowed us to compute the first observable, lensing [12]. Passing to biased tracers, they have been studied in the context of the EFTofLSS in [9, 47–51] (see also [52]), the halo power spectrum and bispectrum including all cross correlations with the dark matter field have been studied and compared with simulations in [9, 48]. Redshift space distortions have been first developed for dark matter [10, 53], and then applied to biased tracers in [50]. Some potentially present subleading ingredients of the universe have also been included in the EFTofLSS, such as clustering dark energy [8, 34, 54, 55], primordial non-Gaussianities [48, 53, 56–59], and neutrinos [7, 60].

With the completion of [50], the one-loop power spectrum of biased tracers in redshift space in ΛCDM cosmology had been computed and compared to simulations. At that point, the EFTofLSS had become ready to be compared with observational data of galaxy clustering. This is where the journey of this paper begins.

**Technical aspects of the analysis.** Before actually starting to go through the details of the paper, we wish to point out that in our analysis we introduce some technical aspects that are not specifically related to the EFTofLSS, but that could apply to any other analysis. This is the treatment of the window function in section 3.1, of the redshift section effect in section 3.2, of the fiber collisions in section 3.3, as well as of the marginalization of any parameter of the model that appear at most quadratically in the Likelihood in section 3.4.

## 2 Summary of the theory model

### 2.1 Biased tracers in redshift space

With a handful of parameters, the EFTofLSS can describe the gravitational clustering of galaxies in redshift space at sufficiently long wavelengths with arbitrary accuracy. This is done by adding suitable counterterms, whose form is dictated by the symmetries obeyed by LSS, that are able to correct for the sensitivity to uncontrolled short-distance wavelengths that affect the perturbative expression. For completeness, we here review the basic equations of the EFTofLSS that we use to analyze the data.

In the standard very accurate approximation where the time dependence of the diagrams is approximated by the growth factors, at one loop, the galaxy power spectrum in redshift space depends on the cosmology and the redshift but also on 10 free parameters [50]. There are 4 galaxy biases \( b_i \) entering in the one-loop expression.\(^5\) Those biases are introduced when the galaxy density and velocity fields are expanded in terms of the underlying dark matter density and velocity fields that satisfy the effective ‘smoothed’ equations of motions for the

---

\(^5\)While at third order in perturbations the galaxy fields depend initially on 8 bias parameters [48], 4 are enough to describe the galaxy power spectrum at one loop due to some accidental degeneracies in the \( P_{13} \) diagram [50].
long wavelength modes [9]. These biases and 3 additional counterterms absorb the sizeable
sensitivity of the one-loop expressions to the uncontrolled short-distance non-linearities that
enter in the kernel of the convolution integrals. One captures the nonlinear effects of the
short-scale physics on long-wavelength dark matter modes, parametrized by the so-called
effective ‘sound speed’ of dark matter $c_{\text{ct}}$ [2] and that, at this order, is degenerate with higher-
derivative terms of the bias derivative expansion. Two additional counterterms, parametrized
by $c_{r,i}$, with $i = \{1, 2\}$, cancel the UV-dependence arising from redshift-space distortions [10].
Finally, there are 3 stochastic terms $c_{\epsilon,i}$ to account for the difference between the actual
realization of the galaxy fields in our universe and their expectation values taken in the
computation of the observables [50] (the first of them is the well-known shot-noise term).
They also correct the UV-dependence of some perturbative diagrams.

The one-loop redshift-space galaxy power spectrum reads\(^6\) [50]:

\[
P_g(k, \mu) = Z_1(\mu)^2 P_{11}(k) + 2 \int \frac{d^3q}{(2\pi)^3} Z_2(q, k - q, \mu) P_{11}(|k - q|) P_{11}(q) + 6Z_1(\mu) P_{11}(k) \int \frac{d^3q}{(2\pi)^3} Z_3(q, -q, k, \mu) P_{11}(q) 
+ 2Z_1(\mu)P_{11}(k) \left( c_{\epsilon,1} k^2_{\text{eff}} + c_{r,1} \mu^2 + c_{r,2} \mu^4 k^2_{\text{eff}} \right) + \frac{1}{n_g} \left( c_{\epsilon,1} + c_{r,2} + c_{\epsilon,3} \mu^2 k^2_{\text{eff}} \right).
\]

Here we set $k_{\text{eff}}^{-1}$, which controls the bias derivative expansion, to be $\lesssim k_{nl}^{-1}$, which is the scale
controlling the expansion of the dark matter derivative expansion. $n_g$ is the mean galaxy
density. Providing estimates for these physical parameters, all free parameters in the EFT
are expected to be order $O(1)$.\(^7\) This fact is of some importance when one is only interested
in the cosmological parameters and wants to marginalize over the others. We will build on
this consideration for the partial marginalization procedure described in section 3.4.

The redshift-space galaxy density kernels are given by:

\[
Z_1(q_1) = K_1(q_1) + f \mu_1^2 G_1(q_1) = b_1 + f \mu_1^2, \\
Z_2(q_1, q_2, \mu) = K_2(q_1, q_2) + f \mu_2^2 G_2(q_1, q_2) + \frac{1}{2} f \mu q \left( \frac{\mu_2}{q_2} G_1(q_2) Z_1(q_1) + \text{perm.} \right), \\
Z_3(q_1, q_2, q_3, \mu) = K_3(q_1, q_2, q_3) + f \mu_3^2 G_3(q_1, q_2, q_3) + \frac{1}{3} f \mu q \left( \frac{\mu_3}{q_3} G_1(q_3) Z_2(q_1, q_2, \mu_1 q_3) + \frac{\mu_2}{q_3} G_2(q_2, q_3) Z_1(q_1) + \text{cycl.} \right),
\]

where here $\mu = q \cdot \hat{z} / q$, $q = q_1 + \cdots + q_n$, and $\mu_{i_1 \cdots i_n} = q_{i_1 \cdots i_n} \cdot \hat{z} / q_{i_1 \cdots i_n}$, $q_{i_1 \cdots i_m} = q_{i_1} + \cdots + q_{i_m}$, with $\hat{z}$ being the unit vector in the direction of the line of sight, and $n$ is the order of the
kernel $Z_n$. $K_i$ and $G_i$ are the galaxy density and velocity kernels, respectively, given in
appendix A, and $f$ is the logarithmic growth rate, which can be calculated by first solving
for the linear growth factor, $D$, yielding:

\[
D(a) = \frac{5}{2} \Omega_m H_0^2 H(a) \int_0^a \frac{da'}{a'^3 H(a')^3},  \\
(2.3)
\]

\(^6\)In the following we make the time-dependence implicit in the notation when it is clear from the context.
In practice all observables need to be computed at all the effective redshifts, $z_{\text{eff}}$’s, of the survey sample.

\(^7\)More exactly, for the type of galaxies that are of interest here, all EFT parameters are expected to be of
order of the linear bias $b_1$, which is $\sim O(2)$, see for example [50, 51].
from which $f$ follows:

$$f(a) \equiv \frac{d \log D(a)}{d \log a} = \frac{(5a - 3D(a))\Omega_m}{2D(a)(\Omega_m + a^3(1 - \Omega_m))}.$$  \hspace{1cm} (2.4)

The form used here in (2.2) is a re-writing with different notations of the expressions in [50] (see eq. (3.23) there).

The bispectrum in redshift space is a function of five kinematic variables. Three describe the shape of the triangle, such as two sides $k_1$, $k_2$ and the angle between them $x = \cos \theta = \hat{k}_1 \cdot \hat{k}_2 = -\frac{k_1^2 + k_2^2 - k_3^2}{2k_1 k_2}$. The orientation of the triangle with respect to the line of sight $\hat{z}$ is characterized by two variables, such as the polar angle $\omega$ and the azimuthal angle $\phi$ of $\hat{k}_1$, or equivalently by $\mu_1 = \cos \omega = \hat{k}_1 \cdot \hat{z}$ and $\mu_2 = \hat{k}_2 \cdot \hat{z} = \mu_1 \cos \theta - \sqrt{1 - \mu_1^2} \sin \theta \cos \phi$.

The tree-level galaxy bispectrum in redshift space reads

$$B_g(k_1, k_2, k_3, \mu_1, \mu_2) = 2Z_2(k_1, k_2, \mu_{12})Z_1(k_1)Z_1(k_2)P_{11}(k_1)P_{11}(k_2) + \frac{c_{e4}}{n_g} P_{11}(k_1) + \text{cyc.} + \frac{c_{e4}^2}{n_g^2},$$ \hspace{1cm} (2.5)

where one needs to introduce only one extra (stochastic) parameter, $c_{e4}$, with respect to the power spectrum. Indeed, using a basis of renormalized bias coefficients allows one to identify the constant piece of the bispectrum as the square of the constant piece of the power spectrum [48].

For an efficient exploration of the cosmology dependence of the power spectrum, one can expand the EFTofLSS prediction as a sum of terms that are bias- and redshift-independent. Symbolically, the EFTofLSS power spectrum can be written as:

$$P_g(k,z) = \sum_n \mu_n^{2\alpha_n} f(z)^{\beta_n} b_{\alpha_n}(z)^{\gamma_n} D(z)^{2\delta_n} P_n(k),$$ \hspace{1cm} (2.6)

where $\{\alpha_n, \beta_n\} \in \{0, 1, 2, 3, 4\}$ (redshift-space distortions), $\{\gamma_n, \delta_n\} \in \{0, 1\}$ (biasing), $\rho_n \in \{1, 2\}$ (linear or loop term), with the pieces $P_n(k)$ depending only on the cosmological parameters. While the linear power spectrum and the counterterms are straightforward to obtain, efficient evaluations of the loop integrals are needed to explore the cosmological parameter space in a timely fashion. To this extent, we adopt two computational strategies, a numerical one [33] and an analytical one relying on the FFTLog decomposition of the linear power spectrum [61], described in appendix B. Proceeding in this way allows us to keep robust control over numerical instabilities while adequate performances can be reached for practical purposes. The description of biased tracers in redshift space would not be complete without resumming the long wavelength displacements whose effect can be accurately described in the Eulerian perturbative treatment outlined above only at very high perturbative order. We adopt the original IR-resummation scheme developed in [29], further extended to redshift space with controlled approximations in [10, 53]. Details on the IR-resummation can be found in appendix C.\footnote{With this publication we release a C++ code that computes the IR-resummed one-loop power spectrum multipoles and tree-level bispectrum monopole of biased tracers in the EFTofLSS, called CBiRd: Code for Biased tracers in Redshift space at CBiRd GitHub (see also for a repository of all EFTofLSS codes EFTofLSS repository).}

Finally, we make a comment on additional higher order terms. In the EFTofLSS, each term contributes with a typical size controlled, roughly, by an expansion in $k/k_{\text{NL}}$ (see for example [10] for details). Given that the power spectrum, around the $k$’s of interest for
us, scales approximately as $P_{11}(k) \sim \frac{1}{k_{\text{NL}}^2} (k/k_{\text{NL}})^2$, a term of order $k^4 P_{11}(k)$ is roughly comparable to the $k^2$ stochastic term, and could therefore be included in our model prediction. We will do this in this paper to test the dependence on this additional counterterm. We find that the numerical and observational data prefer the stochastic $k^2$ term, while the $k^4 P_{11}$ is hardly measurable. We therefore do not include it in the final analysis.

### 2.2 Alcock-Paczynski effect

To estimate the galaxy spectra from data, a reference cosmology (usually denoted as ‘fiducial’ cosmology) at a reference redshift is assumed to transform the measured redshifts and celestial coordinates into three-dimensional cartesian coordinates. The difference between the reference cosmology and the true cosmology (usually denoted as ‘observed’ cosmology) produces a geometrical distortion known as the Alcock-Paczynski (AP) effect [62]. The wavenumbers transverse $k_\perp$ and parallel $k_\parallel$ to the direction of the line-of-sight get distorted, such that the measured wavenumbers assuming a reference cosmology, denoted with a subscript ‘ref’,

$$k_\perp^{\text{ref}} = q_\perp k_\perp, \quad k_\parallel^{\text{ref}} = q_\parallel k_\parallel,$$

where the distortion parameters are given by:

$$q_\perp = \frac{D_A(z_{\text{eff}})H(z = 0)}{D_A(z_{\text{ref}})H(z = 0)}, \quad q_\parallel = \frac{H(z_{\text{ref}})/H(z = 0) - H(z_{\text{eff}})/H(z = 0)}{H(z_{\text{eff}})/H(z = 0)},$$

and wavenumbers are measured in units of $h$/Mpc. Here $z_{\text{eff}}$ denotes the effective redshift used to calculate the EFT prediction, while $z_{\text{ref}}$ refers to the redshift of reference chosen for the measurements, $H(z)$ is the Hubble parameter, and the angular-diameter distance $D_A(z)$ is defined as:

$$D_A(z) = \frac{1}{1 + z} \int_0^z \frac{dz'}{H(z')}.$$

When exploring the cosmological parameters, the distortion parameters are calculated at each probed cosmology using the usual formula for $H$ and $D_A$.\footnote{The procedure we describe here is quite different from other analyses present in the literature that usually assume a fixed cosmology for their template and measure the cosmology and redshift-space distortions through the AP effect (see e.g. [22, 25, 63]), where $q_\perp$ and $q_\parallel$ are taken as free parameters to fit. We rather choose to evaluate our theoretical predictions in each step of the regression such that the AP effect is a completely deterministic effect. The exploration of the parameter space stems from the cosmology and redshift dependence of the observables, for which no unnecessary degeneracies are introduced by an approximate modeling.}

The power spectrum in eq. (2.6) is expanded in Legendre polynomials $P_\ell(\mu)$ and its multipole moments read:

$$P_\ell^{\text{ref}}(k^{\text{ref}}) = \frac{2\ell + 1}{2q_\parallel q_\perp} \int_{-1}^1 d\mu^{\text{ref}} \mu^{\text{ref}}(\mu^{\text{ref}})^{2\alpha_n} P_n(k^{\text{ref}}, \mu^{\text{ref}}) P_\ell(\mu^{\text{ref}}),$$

where the true $(k, \mu)$ are related to the reference $(k^{\text{ref}}, \mu^{\text{ref}})$ by [64]:

$$k = \frac{k^{\text{ref}}}{q_\perp} \left[ 1 + (\mu^{\text{ref}})^2 \left( \frac{1}{F^2} - 1 \right) \right]^{1/2},$$

$$\mu = \frac{\mu^{\text{ref}}}{F} \left[ 1 + (\mu^{\text{ref}})^2 \left( \frac{1}{F^2} - 1 \right) \right]^{-1/2},$$

where $F = q_\parallel/q_\perp$.\footnote{The procedure we describe here is quite different from other analyses present in the literature that usually assume a fixed cosmology for their template and measure the cosmology and redshift-space distortions through the AP effect (see e.g. [22, 25, 63]), where $q_\perp$ and $q_\parallel$ are taken as free parameters to fit. We rather choose to evaluate our theoretical predictions in each step of the regression such that the AP effect is a completely deterministic effect. The exploration of the parameter space stems from the cosmology and redshift dependence of the observables, for which no unnecessary degeneracies are introduced by an approximate modeling.}
For the bispectrum, we focus on the monopole, given by \cite{63}:

$$B_0(k_1^{\text{ref}}, k_2^{\text{ref}}, k_3^{\text{ref}}) = \frac{2 \cdot 0 + 1}{2q_0^4} \int_{-1}^{1} d\mu_1^{\text{ref}} \int_0^{2\pi} d\phi^{\text{ref}} B_g(k_1, k_2, k_3, \mu_1, \mu_2), \quad (2.13)$$

where the true components \((k_i, \mu_i)\) are again related to the respective fiducial ones \((k_i^{\text{ref}}, \mu_i^{\text{ref}})\) by eq. (2.11) and (2.12) \cite{65}.

3 Technical aspects of analysis

Given our theory model for the clustering of LSS as specified by the EFTofLSS, it is, at least in principle, a well-established procedure to perform the analysis of the data. We sample the posterior distribution using an MCMC routine, in a standard way that we review later in the section. However, the use of the EFTofLSS enables us to perform some interesting manipulations that simplify the analysis. We therefore describe them in some detail. In this section, we focus on the analysis that includes only the power spectrum. The manipulations when we include the bispectrum are very similar, and we describe them in detail in appendix D.

We denote the data by \(d\) and the model parameters by \(\{\Omega, \vec{b}\}\), \(\Omega\) being the cosmological parameters, \(\vec{b}\) being the EFT parameters. Explicitly,

$$\vec{b} = \{b_1, b_2, b_3, b_4, c_{c1}, c_{c2}, c_{c3}, c_{c4}\}. \quad (3.1)$$

Similarly, we denote the multipole \(\ell\) of the power spectrum of the data as \(P^{(d)}_\ell\). The Likelihood of the data given the model parameters is assumed to be a Gaussian distribution in the residuals:

$$L(d|\{\vec{\Omega}, \vec{b}\}) =$$

$$\text{Exp} \left[ -\frac{1}{2} (P^{(\text{EFT})}_\ell(k, \{\vec{\Omega}, \vec{b}\}) - P^{(d)}_\ell(k)) \cdot C^{-1}(k, k') \cdot (P^{(\text{EFT})}_\ell(k', \{\vec{\Omega}, \vec{b}\}) - P^{(d)}_\ell(k')) \right], \quad (3.2)$$

where repeated indexes are summed over. Here \(C^{-1}(k, k')\) is the inverse of the covariance matrix of the data, and \(P^{(\text{EFT})}_\ell(W)(k)\) is the EFT power spectrum after being convolved with the window function of the survey:

$$P^{(\text{EFT})}_\ell(W)(k) = \int W(k, k') P^{(\text{EFT})}_\ell(k') dk' \simeq \sum_{k'} \Delta k' W(k, k', \ell, \ell') P^{(\text{EFT})}_\ell(k'), \quad (3.3)$$

where \(W(k, k', \ell, \ell')\) is the window function in Fourier space and where for the second equality we approximate the integral with a sum in \(k'\) weighted by thin ‘log-spaced length’ elements \(\Delta k'\) given by the FFTLog of the window functions, see eq. (3.6) below. In the following, we will drop the \(\sum_{k'} \Delta k'\) in the notation.

In practice, we choose 4096 points. This follows that the FFTLog is performed on another spherical Bessel function, which makes the overall integrand highly oscillating. Thus, a rather large number of points is needed for this FFTLog.

3.1 Window function

The real space correlation function multipoles \(\xi_\ell(s)\) after application of the survey window function are given in terms of the configuration-space window functions, \(Q_\ell,\ell'(s)\), by \cite{25, 66, 67}:

$$\xi^{(W)}_\ell(s) = Q_\ell,\ell'(s) \cdot \xi_\ell(s). \quad (3.4)$$
Explicitly, $Q_{\ell,j}(s) \equiv C_{\ell,j}\cdot Q_s(s)$, where:

$$C_{0,j} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$, $C_{2,j} = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 7 & 2 \\ 0 & 1 & 9 \end{bmatrix}$, $C_{4,j} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 8 \\ 0 & 7 & 2 \\ 0 & 2 & 7 \\ 0 & 1 & 9 \end{bmatrix}$, and the $Q_s(s)$ are given in [25, 66, 67]. The window functions for the power spectrum, $W(k,k')_{\ell,j}$ of (3.3), are then given by:

$$W(k,k')_{\ell,j} = \frac{2}{\pi \ell} \int ds s^2 \cdot j_\ell(ks) Q_s(s) \cdot j_{\ell'}(k' s). \quad (3.6)$$

We defer to appendix D to discuss how we apply the window function for the bispectrum.

Computationally, it is quite straightforward to pre-compute $W(k,k')_{\ell,j}$ for the values of interest. In fact, for each couple $\{k,\ell\}$, eq. (3.6) can be computed with a simple Bessel-transform of $j_\ell(ks) Q_s(s)$. Therefore, $W(k,k')_{\ell,j}$ can be computed by evaluating $N_\ell \times N_k$ Bessel transforms, with $N_\ell$ being the number of multipoles, and $N_k$ being the one of $k$’s ($N_\ell = 3$, $N_k = 68$ in the case of the analysis presented in this paper). We plot $W(k,k')_{\ell,j}$ for $k \approx 0.1 h \text{ Mpc}^{-1}$ in figure 1. One easily sees that indeed, being the window function a multiplication in real space, it is a convolution in Fourier space. One can also see that the computation of $W(k,k')_{\ell,j}$ does not present numerical challenges, and, on the linear power spectrum, we have checked that we obtain the same result on the power spectrum as the standard procedure which acts on configuration space.

Now that we have $W(k,k')_{\ell,j}$ at our disposal, we can follow two approaches. If one uses a grid of cosmological predictions, it is possible to apply directly (3.3) to the points of the grid, so that the evaluation-time in each step in the MCMC is shortened as the window function is applied only once, directly at the level of the grid, for a given choice of the model parameters. The result of the application of the window function to $P_\ell(k)$ is given in figure 2. This is the approach we will use in this paper.
Alternatively, one can notice that one can expand eq. (3.2) and write:

\[
\mathcal{L}(d|\{\vec{\Omega}, \vec{b}\}) = \text{Exp} \left[ -\frac{1}{2} P_0^{(\text{EFT})}(k, \{\vec{\Omega}, \vec{b}\}) \cdot C_{WW}^{-1}(k, k')_{\ell,\ell'} \cdot P_0^{(\text{EFT})}(k', \{\vec{\Omega}, \vec{b}\}) + P_0^{(\text{EFT})}(k, \{\vec{\Omega}, \vec{b}\}) \cdot C_{WW}(k, k')_{\ell,\ell'} \cdot P_0^{(d)}(k') - \frac{1}{2} P_0^{(d)}(k) \cdot C_{WW}(k, k')_{\ell,\ell'} \cdot P_0^{(d)}(k') \right],
\]

(3.7)

where

\[
C_{WW}^{-1}(k_1, k_2)_{\ell_1,\ell_2} = W(k'_1, k_1)_{\ell'_1,\ell_1} \cdot C_{WW}^{-1}(k'_1, k'_2)_{\ell'_1,\ell'_2} \cdot W(k'_2, k_2)_{\ell'_2,\ell_2},
\]

\[
C_{WW}(k_1, k_2)_{\ell_1,\ell_2} = W(k'_1, k_1)_{\ell'_1,\ell_1} \cdot C_{WW}(k'_1, k'_2)_{\ell'_1,\ell'_2} \cdot W(k'_2, k_2)_{\ell'_2,\ell_2},
\]

(3.8)

where \(\dagger\) means hermitian conjugate and applies to both the \(\{k, k'\}\) and \(\{\ell, \ell'\}\) indexes. In practice, \(C_{WW}^{-1}\) is obtained by a left multiplication of \(C^{-1}\) with \(W^\dagger\), while \(C_{WW}^{-1}\) is obtained by a left and right multiplication of \(C^{-1}\) with \(W\) and \(W^\dagger\) respectively. Notice that eq. (3.7) has drastically simplified the implementation of the window function. Once \(W(k, k')_{\ell,\ell'}\) has been computed, \(C_{WW}^{-1}\) and \(C_{WW}\) in eq. (3.8) can be pre-computed. So, there is no need to apply the window function at each step of the MCMC.

In terms of computational cost, this advantage compensates for the fact that now to evaluate eq. (3.7) at each step of the MCMC, we have to evaluate two contractions of the power spectra with the covariance matrix: one for the term in \(C_{WW}^{-1}\) and one for the term in \(C_{WW}^{-1}\). The term in \(C^{-1}\) does not depend on the cosmological parameters and so can be precomputed. This is in contrast with the evaluation of eq. (3.2), which, after \(P_0^{(\text{EFT})}(W)\) has been computed, has only one contraction. However, the computational advantage of eq. (3.7) is quite large.

In the case of the EFTofLSS, the advantage of using eq. (3.3) or eq. (3.7) is particularly important also for another reason. In the standard approach of applying the window function, one takes \(P_0^{(\text{EFT})}(k)\), Bessel transforms it to obtain \(\xi_0^{(\text{EFT})}(s)\), applies \(Q_{\ell,\ell'}(s)\), that acts diagonally in \(s\) space, and Bessel transforms it back to obtain \(P_0^{(\text{EFT})}(W)(k)\). The EFTofLSS has the peculiarity that it is able to reach arbitrarily large accuracy at low wavenumbers, and this is achieved somewhat at the cost of producing large mistakes at large wavenumbers, where the theory is not reliable. These mistakes take the form of steep functions that make it
particularly hard to Fourier transform \( P_{\ell}^{(\text{EFT})}(k) \). It is therefore challenging to go back-and-forth from \( P_{\ell}^{(\text{EFT})}(k) \) to \( \xi_{\ell}^{(\text{EFT})}(s) \). Being able to use the window function in Fourier space avoids this difficulty.

### 3.2 Redshift selection effects

The BOSS data we analyze in this paper have been binned in broad redshift bins. For example, the largest data set, the high-\( z \) North Galactic Cap (NGC), has the redshift distribution \( n(z) \) given in figure 3, where \( n(z) \) is meant to be normalized to one. In principle, it is possible to account for this distribution exactly in the EFTofLSS by performing suitable integrations over time of the various kernels. However, in the approximation in which galaxies that are broadly separated in redshift do not contribute to the power spectrum, and in the limit in which the redshift bin is not very wide, it is enough to average the EFTofLSS prediction over this redshift distribution (see e.g. [67, 68] for some recent precise expressions):

\[
P_{\ell}^{(\text{EFT})}(k) \approx \int dz \, n(z)^2 P_{\ell}^{(\text{EFT})}(k, z). \tag{3.9}
\]

The \( z \)-dependence of \( P_{\ell}^{(\text{EFT})}(k, z) \) is rather straightforward, as it factorizes in the sum of terms with different \( z \)-dependence, and the \( z \)-dependence is factorized from the \( k \)-dependence. The dark matter power spectra depend on \( z \) as \( D(z)^2 \), \( f(z)D(z)^2 \) and \( f(z)^2D(z)^2 \) at linear level, according if we consider matter-matter, matter-velocity or velocity-velocity power spectra. At one loop level, one has, respectively, \( D(z)^4 \), \( f(z)D(z)^4 \) and \( f(z)^2D(z)^4 \). On top of this, the EFT-parameters depend on \( z \) in a way that is expected to be similar to the one of powers of the growth factors. Therefore, implementing eq. (3.9) amounts to pre-compute several integrals of the form

\[
\int dz \, n(z)^2 b_i(z)^n b_j(z)^m f(z)^3 D(z)^4.
\]

In the limit that the redshift bin is much smaller than the scale of variation of \( D(z) \) and the EFT-parameters, we expect that for all these quantities the binning corresponds to evaluating all the quantities at the same effective redshift. Let us start by discussing the terms with \( n_1 = n_2 = 0 \). In this case, the integral in eq. (3.10) can be performed, and one can set it equal to a \( f(z_{\text{eff}}(n_3, n_4))^m \) \( D(z_{\text{eff}}(n_3, n_4))^4 \). Mathematically, \( z_{\text{eff}} \) depends on \( n_3, n_4 \). However, in figure 4, we show, as an example, that for \( n_4 = 2, 4 \) and \( n_3 = 0, 2 \), the choice of \( z_{\text{eff}} = 0.55 \) is accurate for all the terms. In interpreting this, one should also account that, for the worst terms, which occur for \( n_3 = 0 \) and, at some level, for \( n_3 = 1 \), the contribution is multiplied by factors of the EFT-parameter, which can absorb the offset in \( z_{\text{eff}} \). Additionally, notice that the loop terms, i.e. \( n_4 = 4 \), will contribute by a smaller amount to the final answer (about 10%), so one can afford a larger inaccuracy in those. Notice also that the contribution of the term in \( f^2 \) to the monopole, which is the better measured quantity, is suppressed by several numerical factors. Since the dependence on \( z \) of the EFT-parameters is expected to be comparable, we extend this conclusion to all the terms. This allows us to conclude that we can safely approximate \( P_{\ell}^{(\text{EFT})}(k) \approx P_{\ell}^{(\text{EFT})}(k, z_{\text{eff}}) \). Furthermore, we have checked that for the BOSS data an inaccuracy in \( z_{\text{eff}} \) of \( \pm 0.02 \) does not affect the results.

### 3.3 Fiber collisions

In the BOSS survey, fiber collisions affect the determination of the redshift of nearby galaxies, systematically affecting the power spectrum at \( k \gtrsim 0.1 h \text{ Mpc}^{-1} \). Several methods have been proposed to correct for this systematic error. In this paper, we will employ the ‘effective
window method’ of [69]. This method has been shown to be extremely accurate in cancelling the systematic error induced by fiber collisions up to very small scales, well beyond the ones we will be interested in this paper: $k \gtrsim 0.3h\text{Mpc}^{-1}$ and, possibly, even much higher than this [69]. The method consists of adding to the model power spectrum a $\Delta P(k)$ (defined in eq. (24) of [69]), which is computed in terms of the model and in terms of the window function of the fiber collision, plus two nuisance parameters that correspond exactly to the shot noise of the monopole and the $k^2$ stochastic term in the quadrupole that already appear in the EFTofLSS prediction. The resulting power spectrum has to be compared to the data measured with the method dubbed ‘Nearest Neighbor’ (NN), which already takes into partial account the effect of the fiber collisions.
Figure 5. Residuals between the monopole (dotted, blue) and quadrupole (dotted, green) power spectra and the ones reconstructed using the NN&shot-noise method, measured by the mean of 30 BOSS-like Patchy mocks, together with the error bars of one box. The dashed line represents the correction given by the ‘Effective Window Method’ of [69] with $k_{\text{trust}} = 0.25 h \text{Mpc}^{-1}$. In the figure we use the nuisance parameters of [69] equal to $-43$ for the monopole shot-noise and $-9$ for the quadrupole $k^2$ stochastic term (Notice that these values are much smaller than the natural ones we have for the stochastic EFT-parameters). We see that the signal is reconstructed with extreme accuracy, allowing us to neglect any residual systematic error from fiber collision effects.

In order to get a sense of how much, potentially-undetected, residual inaccuracies of this method might affect the determination of cosmological parameters, we compare the results of the Effective Window Method with the ones we obtain if we instead use the less-accurate (but still, as we will see, sufficiently so) method of [70], dubbed ‘Nearest Neighbor Method+adjustable shot noise’ (NN&shot-noise).

The result of the implementation of the corrections for fiber collisions can be seen in figure 5, where we plot the residuals between the true power spectra and the ones reconstructed using the NN&shot-noise method, measured in BOSS-like Patchy mock catalogues whose characteristics we describe later in section 4, together with the error bars of the data. On the same plot, we show the reconstructed signal obtained with the Effective Window Method. We see that, after the implementation of the Effective Window Method, the residuals are negligible, safely guaranteeing that we have fully accounted for the effect of fiber collisions. As an extra precautionary note, we analyze our data also using the NN&shot-noise method, finding that the mildly inaccurate implementation shifts our $A_s$ parameter by less than $1/6$ of a $\sigma$, and unmeasurably so the other parameters. Given that the Effective Window Method is much more accurate than the NN&Shot-noise method, we therefore conclude that fiber collisions present no residual non-negligible source of systematic error.

3.4 EFT-parameter marginalization

The prediction $P^{(\text{EFT})}_\ell$ depends on several counterterms and galaxy bias parameters (EFT parameters). Their measurement is interesting, as they contain information about galaxy formation: it is the way different galaxy formation mechanisms affect large scales. However if somebody is only interested in the cosmological parameters, as we are in this paper, one will eventually marginalize over them.

In this paper we take two approaches. We will explore the Likelihood in eq. (3.7), which depends on 10 EFT parameters and the cosmological parameters, by sampling with MCMC’s, and then marginalize over them. We will use this as a consistency check of a more
efficient approach, which is the following. If we are ultimately interested in marginalizing over
the bias parameters, it means that we are interested in the Likelihood of the cosmological
parameters, which is the Likelihood after integration over the EFT parameters:

\[ \mathcal{L}(d|\{\vec{\Omega}\}) = \int db \mathcal{L}(d|\{\vec{\Omega}, \vec{b}\}). \]  

(3.11)

We now split the EFT parameters as:

\[ \vec{b} = \{\vec{b}_G, \vec{b}_{NG}\}, \]  

(3.12)

where the distinction between \( \vec{b}_G \), and \( \vec{b}_{NG} \) is that the \( \vec{b}_G \)'s appear in the exponent of eq. (3.7)
only up to quadratic level. This means that the \( b_G \)'s appear in \( P_{\ell}(EFT)(k, \{\vec{\Omega}, \vec{b}\}) \) only up to
linear level. In our notation, \( \vec{b}_G = \{b_3, c_\xi, c_{r,1}, c_{r,2}, c_{\epsilon,1}, c_{\epsilon,2}, c_{\epsilon,3}\} \), and \( \vec{b}_{NG} = \{b_1, b_2, b_4\} \). We
therefore can write:\(^10\)

\[ P_{\ell}(EFT)(k, \{\vec{\Omega}, \vec{b}\}) \equiv \sum_i b_{G,i} P_{\ell,lin,i}(k, \{\vec{\Omega}, \vec{b}_{NG}\}) + P_{\ell, const}(k, \{\vec{\Omega}, \vec{b}_{NG}\}). \]  

(3.14)

Plugging back in eq. (3.11), we realize that the integral over the \( b_G \)'s can be done analytically,
as it is a Gaussian integral, without the need to explore the Likelihood numerically. In order
to make the Gaussian structure more apparent, we can define:

\[
\mathcal{L}(d|\{\vec{\Omega}, \vec{b}_G, \vec{b}_{NG}\}) = \text{Exp} \left[ -\frac{1}{2} \sum_{ij} b_{G,i} F_{2,ij} \left( P_{\ell,lin}(k, \{\vec{\Omega}, \vec{b}_{NG}\}), C_{WW}^{-1}\right) b_{G,j} + \right.
\]

\[
\left. + \sum_i b_{G,i} F_{1,i} \left( P_{\ell,lin}(k, \{\vec{\Omega}, \vec{b}_{NG}\}), P_{\ell, const}(k, \{\vec{\Omega}, \vec{b}_{NG}\}), P_{\ell}(d), C_{WW}^{-1}, C_{W}^{-1}, P_{\ell}(d)\right) \right. 
\]

\[
\left. + F_0 \left( P_{\ell, const}(k, \{\vec{\Omega}, \vec{b}_{NG}\}), P_{\ell}(d), C_{WW}^{-1}, C_{W}^{-1}\right) \right]. 
\]  

(3.15)

where

\[
F_{2,ij} = P_{\ell,lin,i}(k, \{\vec{\Omega}, \vec{b}_{NG}\}) \cdot C_{WW}^{-1}(k, k') \ell, \ell' \cdot P_{\ell,lin,j}(k', \{\vec{\Omega}, \vec{b}_{NG}\}), 
\]  

(3.16)

\[
F_{1,i} = -P_{\ell, const}(k, \{\vec{\Omega}, \vec{b}_{NG}\}) \cdot C_{WW}^{-1}(k, k') \ell, \ell' \cdot P_{\ell,lin,i}(k', \{\vec{\Omega}, \vec{b}_{NG}\}) + 
\]

\[
+ P_{\ell,lin,i}(k, \{\vec{\Omega}, \vec{b}_{NG}\}) \cdot C_{W}^{-1}(k, k') \ell, \ell' \cdot P_{\ell}(d)(k'), 
\]

\[
F_0 = -\frac{1}{2} P_{\ell, const}(k, \{\vec{\Omega}, \vec{b}_{NG}\}) \cdot C_{WW}^{-1}(k, k') \ell, \ell' \cdot P_{\ell, const}(k', \{\vec{\Omega}, \vec{b}_{NG}\}) + 
\]

\[
+ P_{\ell, const}(k, \{\vec{\Omega}, \vec{b}_{NG}\}) \cdot C_{W}^{-1}(k, k') \ell, \ell' \cdot P_{\ell}(d)(k') - 
\]

\[
\frac{1}{2} P_{\ell}(d)(k) \cdot C^{-1}(k, k') \ell, \ell' \cdot P_{\ell}(d)(k'), 
\]

\(^10\)Explicitly,

\[
P_{\ell, const}(k, \{\vec{\Omega}, \vec{b}_{NG}\}) = P_{\ell}(EFT)(k, \{\vec{\Omega}, \vec{b}\}) \bigg|_{b_G \to 0}, \quad P_{\ell,lin,i}(k, \{\vec{\Omega}, \vec{b}_{NG}\}) = \frac{\partial P_{\ell}(EFT)(k, \{\vec{\Omega}, \vec{b}\})}{\partial b_{G,i}} \bigg|_{b_G \to 0}. 
\]  

(3.13)
and then we can perform a Gaussian integral to define the partially marginalized Likelihood:

\[
\mathcal{L}(d|\{\vec{\Omega}, b_{NG}\}) = \int db_G \mathcal{L}(d|\{\vec{\Omega}, b_G, b_{NG}\}) = \exp \left[ \frac{1}{2} F_{1,i}((\vec{\Omega}, b_{NG})) \cdot F_{2}(\{\vec{\Omega}, b_{NG}\})^{-1} \cdot F_{1,j}(\{\vec{\Omega}, b_{NG}\}) + F_{0}(\{\vec{\Omega}, b_{NG}\}) - \frac{1}{2} \log \left( \det \left( F_{2}(\{\vec{\Omega}, b_{NG}\}) \right) \right) \right],
\]

where in the last passage we neglected an irrelevant constant. \( \mathcal{L}(d|\{\vec{\Omega}, b_{NG}\}) \) depends on only 3 EFT parameters, so it is quite easier to explore numerically with an MCMC. This is compensated, but only in part, by the fact that this Likelihood is slower to evaluate numerically for each choice of the cosmological parameters, as the functional form is more complex than a simple \( \chi^2 \)-one. We will refer to this Likelihood as the partially marginalized Likelihood. The partially marginalized Likelihood in the case in which we include the bispectrum is discussed in appendix D. Notice finally that this procedure does not require to include priors on the EFT-parameters, but these can always be simply included before performing the integral in eq. (3.17).

### 3.5 Details on the MCMC analysis

We obtain the parameter Likelihood described above using the Python-based Monte-Carlo Markov Chain (MCMC) sampler *emcee* [71]. We run four separate chains in parallel and use the Gelman-Rubin convergence criteria [72] with a threshold of \( \epsilon \leq 0.005 \) on the scale-reduction parameters, to determine convergence.

Beyond the Gelman-Rubin convergence criteria, we have performed several additional convergence tests. Here we list a few. We have checked that the non-marginalized Likelihood of eq. (3.7) and the partially marginalized Likelihood of eq. (3.17) give consistent results. Similarly, we obtain consistent results if we impose the Gelman-Rubin test threshold to be 0.005 instead of 0.003, or if we change other settings of the MCMC (such as the number of walkers).

At each step of the likelihood evaluation, we take the model from a pre-computed grid of the cosmology-dependent pieces of eq. (2.6). The dependence on the EFT parameters is instead analytic, and so evaluation for the model concerning the variation of these parameters can be done directly at the level of the MCMC. This setup allows for an extremely rapid evaluation of the likelihood. For this first analysis of the BOSS data using the EFTofLSS, we limit ourself to a three-dimensional grid scanning over the cosmological parameters \( \ln(10^{10} A_s) \), \( \Omega_m \) and \( h \), where \( A_s \) is the amplitude of the primordial power spectrum, \( \Omega_m \) is the matter abundance, and \( h \) is the present day Hubble parameter measured in units of 100 km/(sec Mpc). We will therefore fix the ratio of the baryons versus dark-matter abundance, \( \Omega_b/\Omega_c \), and the tilt of the primordial power spectrum, \( n_s - 1 \), to the Planck2018 best-fit values. However, there should be no difficulty in extending our analysis to include these additional cosmological parameters as well as others, such as, for example, the dark-energy equation of state, \( w \). In fact, one possible option is the following. Since the cosmological parameters are already quite well-determined, we are interested only in small relative variations of them. Therefore, the predictions for all models of interest can be obtained by Taylor expanding the prediction around the best cosmology (say, for example, Planck2018+BOSS DR12), to the desired order. In this way, the dependence on all the parameters would be analytical, allowing for a rapid evaluation of the likelihood directly at the level of the chain, without
building grids, and potentially for some additional simplifications. This approach has been developed in [33] for the two-loop dark-matter power spectrum in the EFTofLSS, and was found to give sub-percent accuracy. We foresee no obstacles in doing this also for the power spectrum of biased tracers in redshift space. We leave this to future work [73].

Finally, in the EFTofLSS, the EFT parameters are supposed to be order one, as in this case all operators become strongly coupled at around the same scale. We will therefore impose the following priors on the parameters that encompass the physically-motivated range:

\[ b_1 \in [0,4]_{\text{flat}}, \quad c_2 \in [-4,4]_{\text{flat}}, \quad c_4 \in 2_{\text{gauss}}, \quad b_3 \in 2_{\text{gauss}}, \quad (3.18) \]

\[ c_{ct} \in 2_{\text{gauss}}, \quad c_{r,1} \in 4_{\text{gauss}}, \quad c_{r,2} \in 4_{\text{gauss}}, \quad (3.18) \]

\[ c_{\epsilon,1}/\bar{n}_g \in \left(\frac{400}{(h \text{ Mpc}^{-1})^3}\right)_{\text{gauss}}, \quad c_{\epsilon,\text{mono}} \in 2_{\text{gauss}}, \quad c_{\epsilon,\text{quad}} \in 2_{\text{gauss}}, \quad (3.18) \]

where \( b_2 = \frac{1}{\sqrt{2}}(c_2 + c_4), \quad b_4 = \frac{1}{\sqrt{2}}(c_2 - c_4), \quad c_{\epsilon,\text{mono}} = c_{\epsilon,1} + \frac{1}{3}c_{\epsilon,2}, \quad c_{\epsilon,\text{quad}} = \frac{2}{3}c_{\epsilon,2}. \)

We choose \( k_{\text{NL}} = k_{M} = 0.7h \text{ Mpc}^{-1}. \) Notice that \( \bar{n}_g = 4.5 \cdot 10^{-4}/(h \text{ Mpc}^{-1})^3 \) for CMASS and \( \bar{n}_g = 4.0 \cdot 10^{-4}/(h \text{ Mpc}^{-1})^3 \) for LOWZ.\(^{11}\) Here the subscript ‘flat” meant that we impose a flat prior with the stated boundaries, while the subscript ‘gauss’ means that we impose a gaussian prior centered at zero with that standard deviation. Notice that all priors are symmetric around zero, because we have no theoretical or observational prejudice on the actual sign of the terms. The only exception to this is \( b_1, \) which however, as we will see, is determined by the data to be \( b_1 \simeq 2 \) with about 5% precision, and therefore the prior plays evidently no role in this case.

We perform the change of variables between \( (b_2, b_4) \) and \( (c_2, c_4) \) as we find for all simulations and in the observational data that the former are almost completely anti-correlated (at \( \sim 99\% \)). Thus, \( c_4 \) that parametrizes the difference between the contributions of \( b_2 \) and \( b_4 \) in the 1-loop term, which are highly degenerate, is practically undetermined, and we put a gaussian prior with standard deviation of 2 on it. Then, upon comparison with the numerical and observational data, we find that for the signal-to-noise ratios that are relevant in our study (for BOSS-like volume), the functions that are multiplied by \( c_4 \) and \( c_{\epsilon,\text{mono}} \) are too small to significantly affect the results. We therefore set \( c_4 \) and \( c_{\epsilon,\text{mono}} \) to zero. As we do not analyze the hexadecapole in our analyses, we set \( c_{r,2} = 0 \) and put a prior \( 8_{\text{gauss}} \) on \( c_{r,1} \) to absorb the main contribution of \( c_{r,2} \) in the quadrupole. Also, we impose a prior on the shot noise consistent with fiber collisions and selection effects, since (most of) the Poisson shot noise is subtracted from the data. Our choice of the priors plays no role as long as all parameters are well determined by the data, as we will find to be the case in simulations with volume as large as the Challenge boxes that we describe later, and whose volume is about 16 times the BOSS volume. We have checked for those boxes that significantly enlarging the priors does not affect our results, and all parameters stay of order of their expected physical size. For volumes of the size of the BOSS data, the priors make certain that the fit lies within the physical region of the EFT parameters, as potential unbroken degeneracies allowed by the large error bars of the data might lead to undesirable systematic errors. We will test our prior choice using the Patchy mocks that have the same volume as the data we want to analyze. When we fit the bispectrum, we do not set \( c_4 \) to zero, as \( b_2 \) and \( b_4 \) now enter in the observables in a less degenerate way. Furthermore, as we fit the bispectrum only up to a low \( k_{\text{max}}, \) the bispectrum shot noise terms are negligibly small, especially since most of them are subtracted from the data. We therefore put them to zero.

\(^{11}\)For the Challenge boxes, we use \( \bar{n}_g = 3 \cdot 10^{-4}/(h \text{ Mpc}^{-1})^3. \)
4 Tests on simulations

While there is no question that the EFTofLSS is accurate at low enough wavenumbers, it is unclear what is the wavenumber at which, given the error bars on the data, the EFTofLSS becomes too inaccurate and measurements begin to be biased. We call this error the ‘theory-systematic’, to be distinguished from the observational and instrumental systematic errors, and from the bias parameters of the EFTofLSS. As discussed above, it is possible to have a rough idea of what this wavenumber is, because, in the EFTofLSS, one can estimate the theoretical error associated with the higher-order terms that have not been computed [2, 4, 10, 12, 13]. However, this procedure is quite approximate, and, given the importance of going to higher wavenumbers to gain more information, it is preferable to have additional methods to measure the theory systematic error in the measurements.

Testing the EFTofLSS against simulations can provide a systematic way to quantify these different errors, even though it is a bit unclear if the systematic error associated with the simulations is much smaller than percent level (see for example [74]). We analyze two sets of simulations.

**BOSS DR12 challenge boxes.** The BOSS DR12 analysis [23, 25] performed a blind mock challenge on a set of high-resolution N-body simulations populated with galaxies using Halo Occupation Distribution (HOD) modeling. There are seven different catalogs with different HOD prescriptions, constructed from periodic simulation boxes with varying cosmologies, and they do not include any sky-geometry selection function effect. The seven catalogs are labeled A through G. Several of the boxes are based on the Big MultiDark simulation [75]. The catalogs are constructed out of simulation boxes with a range of 3 underlying cosmologies, and boxes with the same cosmology have varying galaxy bias models, with resulting variance of the overall galaxy bias by about 5%. The redshift of the boxes ranges from $z = 0.441$ to $z = 0.562$. We analyze the simulations at high redshift, which are relevant for our analysis, i.e. A, B, F, G and D. These cases were designed to quantify the sensitivity of RSD models to the specifics of the galaxy bias model over a reasonable range of cosmologies. The cosmology and relevant simulation parameters for each of the boxes that we use are given in table 1. A comparison of the results from the mock challenge in the context of the BOSS DR12 results is presented in [23], and individual Fourier-space clustering results from the challenge are discussed in [25, 76].

**Patchy mocks.** The BOSS collaboration provides a set of mock catalogs, which mimic the clustering properties and geometry of the observed dataset [77]. These mock catalogues have been produced using approximate gravity solvers and analytical-statistical biasing models. The catalogues have been calibrated to an N-body based reference sample extracted from one of the BigMultiDark simulations [78], which were performed using Gadget-2 [79] with $3840^3$ particles in a volume of $[2500h^{-1}\text{Mpc}]^3$ assuming a $\Lambda$CDM cosmology with $\Omega_m = 0.307115$, $\Omega_b = 0.048206$, $\sigma_8 = 0.8288$, $n_s = 0.9611$, and a present-day Hubble constant of $H_0 = 67.77\text{km s}^{-1}\text{Mpc}^{-1}$. Here $\sigma_8$ is the root mean square of the linearly-evolved matter power spectrum within a top-hat radius of $8h\text{Mpc}^{-1}$ at redshift zero. Halo abundance matching is used to reproduce the observed BOSS 2- and 3-point clustering measure-

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12 Not all systematic errors in the simulations are very problematic for this test. As long as the simulations implement some mechanism that preserves the laws of physics that are used to derive the EFTofLSS, such as for example coordinate reparametrization invariance, the EFTofLSS should be able to describe it, notwithstanding that the actual physics of galaxy formation in our universe might be different than the simulated one.
Table 1. Cosmological parameters and specifications of the N-body simulations analyzed in this work. $V_{\text{eff}}$ is the effective volume of the box, while $z_{\text{eff}}$ is the effective redshift, accounting for the window and redshift selection. The factor of ‘×16’ in the $V_{\text{eff}}$ of the Patchy mocks represents the fact that we will use 16 independent boxes, each with effective volume equal 3 Gpc$^3$.
All the MCMC are run with the baryons-to-dark-matter ratio $\Omega_b/\Omega_c$ and the tilt of the primordial power spectrum $n_s$ fixed and given in table 1, while we freely scan over $A_s$, $\Omega_m$ and $h$. We have checked on Fisher forecasts of the CMASS NGC dataset that removing the priors on $\Omega_b/\Omega_c$ and $n_s$ to about ten times the errors from Planck2018 degrades our constraints just roughly by about 18% for $\ln A_s$, 27% for $\Omega_m$, and 53% for $h$.\footnote{If instead we put an extremely wide prior on $n_s$ to roughly the interval $[0.70 - 1.22]$, which is about 60 times the error bar from Planck2018, and put a prior on $\Omega_b/\Omega_c$ of the order of four times the percentage error on $\Omega_b h^2$ obtained by Big Bang Nucleosynthesis (BBN) (which is about 20$s^2$ of Planck2018)\cite{80}, the error bars are degraded by about 117%, 158% and 267% for $\ln A_s$, $\Omega_m$ and $h$ respectively. Given that we will measure these quantities to 12%, 3.2%, and 2.9% respectively, we still would have a significant and valuable measurement.}

4.1 Size of the loop terms

Before analyzing the measurements in a systematic way, a simple $\chi^2$-fit at the best fit of the simulations can provide insights on how much the various terms in the power spectrum of eq. (2.1) contribute overall (one could perform this study equally well by fixing the fit to the fiducial cosmology). One can get a sense of the applicability range of the EFTofLSS along with other interesting features that we detail in the following. In figure 6 are shown various contributions to the monopole and quadrupole for the challenge boxes and Patchy mocks.}

**Figure 6.** Size of various contributions to the monopole and quadrupole of the one-loop power spectrum of eq. (2.1) evaluated at the best fit cosmology at $k_{\text{max}} = 0.23 h \text{ Mpc}^{-1}$ for the challenge boxes (top) and for the Patchy mocks (bottom). Dashed stays for negative value. We also plot, in dotted, the size of the statistical error bars. 

First, at high wavenumbers, $k_{\text{max}} \simeq 0.23 h \text{ Mpc}^{-1}$, on the monopole, we notice that, on the Challenge A box, the linear part contributes to $\sim 85\%$ of the total while the one-loop contributes to $\sim 15\%$. On the quadrupole, the contribution of the loops is instead larger,
and reaches order half of the linear one. On Patchy, on the monopole, the loop contribution is about 10\%, while on the quadrupole the loop contribution is about half of the linear one. This allows us to very roughly estimate the theoretical error which corresponds to the size of the next-to-leading contribution, namely, the two-loop. The size of the $L$-loop $P^{(L)}(k)$ can be roughly estimated as $P^{(L)}(k) \sim \alpha^L P_{11}(k)$, where $P_{11}$ is the linear power spectrum and $\alpha$ is a small number representing the size of the loop contribution. It follows that around $k \sim 0.25h\,\text{Mpc}^{-1}$, the theoretical error is about 1\% of the power spectrum for the monopole, which is of the same order as the measurement statistical error, and a similar conclusion holds, approximately, for the quadrupole, where the statistical error is larger. This suggests, if not completely justifies, that we can use the EFTofLSS at one-loop order up to $k_{\text{max}}$ in the range $0.20 h\,\text{Mpc}^{-1}$ to $0.25h\,\text{Mpc}^{-1}$ for surveys with volume similar to the BOSS CMASS one. We will discuss further the $k$-range accessible by the theory at this order in the following, when concluding our systematic analysis.

Second, representative contributions are also plotted in figure 6. We first observe that keeping only one galaxy bias $b_i$ to its best-fit value ($\sim O(2)$) while setting all other EFT parameters to zero gives the typical size of the contributions entering in the one-loop contribution. We notice that there are some cancellation at play in loops in the monopole, but they are not extremely large. Additionally, in the quadrupole, the $k^2$ stochastic term seems to be the larger contribution.

To finish on preliminary considerations, we would like to make a comment on the constant stochastic term. In principle, it is of size $\bar{n}^{-1}_g$, which is of the order of the typical significant contributions to the one-loop. In practice, the shot-noise contribution is often subtracted from the measurements. However, this procedure relies on estimators for weighting and selecting the galaxies to account for the spatial variance of the number density. For these reason, we fit without shot noise the Challenge boxes as the shot-noise subtraction is reliable in simulations where the number density is known, while we fit with a shot noise the data to account for small inaccuracies in the estimation of $\bar{n}_g$. We indeed find that keeping the constant stochastic term with reasonable priors (a few percents of $\bar{n}^{-1}_g$) in regards of the above discussion helps the performance of the fit. Furthermore, we remind the reader that the ‘Effective Window Method’ correcting fiber collisions contains a nuisance parameter in the form of a shot noise.

### 4.2 Theory-systematic error

On the left of figure 7, we show the marginalized posterior distributions of the cosmological parameters measured from the Challenge boxes, by fitting the individual power spectrum of each box with the partially marginalized likelihood of eq. (3.17). By power spectrum, here and in the rest of the paper, we mean the monopole and quadrupole of the power spectrum. On top of each posterior distribution, we give the expectation value of the cosmological parameter with its statistical error and theory-systematic error, to be read from top to bottom. In figure 9 we plot level contours of the partially marginalized likelihood, after marginalizing over all but two EFT parameters.

We see that each box is fitted in very comparable terms. Concerning the cosmological parameters, we see that in each box we can measure $A_s, \Omega_m$ and $h$, breaking all degeneracies without any significant input from Planck or other CMB experiments.\footnote{This statement refers just to the inclusion of the sound horizon prior, which is a typical procedure to calibrate BAO data (see for example [81, 82]). We remind the reader that we are fixing $\Omega_b/\Omega_c$ and $n_s$ to their Planck value, and that we discussed in footnote 13 about the degradation of our measurements when this priors are relaxed, concluding that we would still have a significant measurement of the parameters.} We will comment in the next subsection on the physical origin of this.
Figure 7. Marginalized posterior distributions of the cosmological parameters obtained from fitting the power spectrum of the four Challenge boxes ABFG and their average (top) and for box D (bottom), without (left), and with (right) the inclusion of Planck sound horizon prior. For each cosmological parameter, the expectation value is given together with the statistical error and the theory-systematic error (from top to bottom in each subplot). The vertical lines represent the fiducial cosmology of the simulation.

We call $\sigma_{\text{sys}}$ an estimate of the theoretical error associated with the higher-order terms in the EFTofLSS prediction that we have not computed yet and so are not included in the model we fit to the data. Usually $\sigma_{\text{sys}}$ is called a systematic bias, but we do not use this nomenclature in order to avoid confusion with the EFTofLSS parameters or with other forms.
of systematic errors. The challenge boxes A B F and G are highly correlated, as their initial conditions are the same (since they are extracted from the same dark-matter simulation, the Big MultiDark). Thus, $\sigma_{\text{sys}}$ is measured by computing the distance between the true value and the 68% confidence region of the average over the four boxes of the distribution of the posteriors of each box for a given parameter. In particular, $\sigma_{\text{sys}}$ is taken to be zero when the true value is within the 68% confidence region. We perform the same procedure for box D. Then, since box D and boxes A B F and G are constructed from independent simulations, we combine the two measurements of the distance of cosmological parameters from the fiducial error even as large as the statistical error can be accounted for by shifting the posterior of the systematic error in $\Omega_{\text{CMASS}}$. The challenge boxes are high precision N-body simulations with HOD modeling. On box D, we find no evidence of theoretical systematic error in $\Omega_{\text{CMASS}}$ and $h$ all the way to $k_{\text{max}} = 0.25 h\, \text{Mpc}^{-1}$, except a small $\sigma_{\text{sys}} = 0.007 \simeq 0.3 \sigma_{\text{stat,CMASS}}$ on $h$ at $k_{\text{max}} = 0.15 h\, \text{Mpc}^{-1}$. On $A_s$, we find a theoretical systematic error that starts very small at $k_{\text{max}} = 0.2$, where we measure $\sigma_{\text{sys}} = 0.02 \simeq \sigma_{\text{stat,CMASS,x low-z}}/7$, and becomes comparable to $\sigma_{\text{sys}} \simeq 0.4 \sigma_{\text{stat,CMASS,x low-z}}$ and $0.5 \sigma_{\text{stat,CMASS,x low-z}}$ at $k_{\text{max}} = 0.23 h\, \text{Mpc}^{-1}$ and $k_{\text{max}} = 0.25 h\, \text{Mpc}^{-1}$ respectively.

On the combination of boxes ABFG, we similarly find no evidence of theoretical systematic error in $\Omega_{\text{m}}$ and $h$ all the way to $k_{\text{max}} = 0.25 h\, \text{Mpc}^{-1}$, with the exception of a small $\sigma_{\text{sys}} = 0.006 \simeq 0.4 \sigma_{\text{stat,CMASS}}$ on $\Omega_{\text{m}}$ at $k_{\text{max}} = 0.15 h\, \text{Mpc}^{-1}$ and a tiny $\sigma_{\text{sys}} = 0.002 \simeq 0.1 \sigma_{\text{stat,CMASS}}$ on $h$ at $k_{\text{max}} = 0.25 h\, \text{Mpc}^{-1}$. Similarly, on $A_s$, we find a theoretical systematic error that is tiny up to $k_{\text{max}} = 0.20 h\, \text{Mpc}^{-1}$, and it is larger at $k_{\text{max}} = 0.23$, where we measure $\sigma_{\text{sys}} = 0.04 \simeq \sigma_{\text{stat,CMASS,x low-z}}/3$, while it becomes a sizable $\sigma_{\text{sys}} = 0.09$ at $k_{\text{max}} = 0.25 h\, \text{Mpc}^{-1}$. When we combine the measurements of the mean of ABFG and D, we can measure the theoretical systematic error more accurately by a factor of $\sim 1/\sqrt{2}$. We find no evidence of any systematic error on $h$ up to $k_{\text{max}} = 0.25 h\, \text{Mpc}^{-1}$, with a minimum detectable error of $\sigma_{\text{sys, min}} = 0.008$. For $\Omega_{\text{m}}$, we detect a systematic error $\sigma_{\text{sys}} = 0.0037$ at $k_{\text{max}} = 0.15 h\, \text{Mpc}^{-1}$ and $\sigma_{\text{sys}} = 0.0017$ at $k_{\text{max}} = 0.20 h\, \text{Mpc}^{-1}$, with a minimum detectable error $\sigma_{\text{sys, min}} = 0.0032$. In comparison to the error bars we will obtain on the observational data, all these systematic errors are negligibly small ($\lesssim \sigma_{\text{stat,CMASS,x low-z}}/6$), with the possible exception of $k_{\text{max}} = 0.15 h\, \text{Mpc}^{-1}$ where the error is about $\sigma_{\text{stat,CMASS,x low-z}}/4$, which we however assume to be due to a statistical fluctuation. Instead, on $\ln(10^{10} A_s)$, we detect a systematic error that becomes significant at $k_{\text{max}} = 0.23 h\, \text{Mpc}^{-1}$, equal to 0.06, and at $k_{\text{max}} = 0.25 h\, \text{Mpc}^{-1}$, equal to 0.09. At $k_{\text{max}} = 0.20 h\, \text{Mpc}^{-1}$, we detect a systematic error on $\ln(10^{10} A_s)$ equal to 0.036, which is about $1/4 \cdot \sigma_{\text{stat,CMASS,x low-z}}$, which is sufficient for being negligible.\footnote{This interpretation is probably reinforced by the fact that our measurement of the theoretical systematic error is in this case potentially affected by a statistical fluctuation that is already visible at low $k_{\text{max}}$.}

We conclude that we can safely perform the analysis of the data with negligible theoretical error up to $k_{\text{max}} = 0.20 h\, \text{Mpc}^{-1}$. It is now worthwhile to remark that a systematic error even as large as the statistical error can be accounted for by shifting the posterior of the
parameter by about this quantity. This has a very small effect on any statistically-significant statement that one can make on the posterior of the parameters. Therefore, overall, the systematic errors appear to be small even in the highest wavenumbers that we consider. This in particular means that even though we will conservatively decide not to quote the result obtained $k_{\text{max}} = 0.23 h \, \text{Mpc}^{-1}$ as our main results, the systematic error in this case is still quite small, and one could decide to consider the ones obtained with this $k_{\text{max}}$ to be the most significant results.

Though the challenge boxes are high precision N-body simulations with accurate HOD modeling, they lack the window function, redshift sampling and fiber collisions, that characterize a survey such as BOSS. While the challenge boxes are therefore a good tool to test for the theoretical systematic, they do not test for those. However, as we have discussed in section 3.1, 3.2, and 3.3, all of these effects have been accurately implemented in our pipeline, and so we expect negligible systematic error from them.

Overall, one should keep in mind that the window function does not give a huge effect: even analyzing the data without accounting for it, the shift in the cosmological parameters is of the order of the statistical error. So, an implementation which is accurate just to about 10% is enough to make any error negligible. However, to check the accuracy of the treatments of these survey systematics in our pipeline to an even higher level of safety, we analyze 16 Patchy lightcone NGC boxes. The total volume is roughly equivalent to the one of one Challenge box, which allows us to measure survey-systematic errors to the same precision as we measured the theory-systematic errors using the Challenge boxes, though one should keep in mind that the level of accuracy of the mocks is less than the one of an N-body simulation. We fit the 16 boxes separately and do the product of the individual 1D marginalized posteriors. Even if this procedure is not fundamentally correct, as the correct procedure would be to multiply the full likelihoods, it is rather conservative and good enough to measure the theoretical systematic error. Another reason to fit the Patchy boxes individually is to measure the response of the EFT to our choice of priors on the EFT parameters for one volume equivalent to the one of the data. Indeed, for a volume as large as the one of Challenge box, all EFT parameters are well determined. However, when we turn to a volume like the one of Patchy or data, we might encounter unbroken degeneracies that can potentially drive the fit to an unphysical region. Therefore, it is important to test if our choice of priors, which follows solely from theoretical considerations, is enough to keep the fit in the physical region.

On the left of figure 8, we show the marginalized posterior distributions of the cosmological parameters measured from the Patchy mocks, by fitting the individual power spectrum of each box with the partially marginalized likelihood of eq. (3.17). We find for each cosmological parameter at each $k_{\text{max}}$ that the product of the 1D posteriors appear with negligible small systematics errors, with vanishing ones for $A_s$ and $\Omega_m$, and tolerably small $\sigma_{\text{sys}} \lesssim 0.005 \simeq \sigma_{\text{stat}, \, \text{CMASS x low}} / 4$ on $h$.

We conclude that, on one hand, all survey geometrical and selection effects are well accounted for in our pipeline, and, on the other hand, our theoretical priors are suitable to keep the parameters in the physical region for a volume like BOSS data.

Before moving on, we would like to make the following comment about the theory-systematic error as extracted from simulations. One can take the point of view that the simulations and the HOD modeling are exact, i.e. they represent the exact distribution of a population of galaxies generated according to some physical mechanism. Any deviation from the exact representation of the physical model (for example if the trajectory of the particles had a numerical error) would induce a numerical systematic error, that we here assume...
Figure 8. Marginalized posterior distributions of the cosmological parameters obtained from fitting the power spectrum of 16 Patchy lightcone NGC boxes, without (left), and with (right) the inclusion of Planck sound horizon prior. For each cosmological parameter, we quote the expectation value together with the statistical error and the theory-systematic error of the product of the individual 1D posteriors (from top to bottom in each subplot). The vertical lines represent the fiducial cosmology of the simulation.

to be negligible.\footnote{16} It is a well-recognized limitation that the HOD models do not actually simulate the formation of galaxies from first principles, but rather ‘model’ them according to some rule guided by a mixture of physical processes and data. We point out that, from the EFTofLSS point of view, it is absolutely not important that the HOD model correctly represents a population of galaxies that actually exists in the universe. In fact, as long as the galaxy-formation physics is modeled by the HOD with a process that satisfies the principle of general relativity, i.e. it is controlled by locally observable quantities, then the EFTofLSS is expected to be able to describe the distribution of this population with arbitrary accuracy.

Of course, for a given model, the non-linearities induced by the higher order biases can be larger than in other models. In this case, to reach a given accuracy, one might need to include higher-order terms in the EFTofLSS. For example, in [50, 51], higher-order derivative terms were included in order to reproduce to a given accuracy the distribution of highly-biased tracers. This discussion makes it clear that, in the absence of numerical systematic error, the inferred cosmological parameters obtained from the four HOD models should not only be equal to the true ones, but also the same for each model. By looking at figure 9, we notice that indeed all HOD models of the challenge boxes give very similar results for all the cosmological parameters. All the EFT-parameters that have not been already integrated out have almost identical values apart from the linear bias $b_1$, which seems to be the one over which the differences in the HOD models accumulate. It is therefore not a complete surprise that the model with the smallest systematic error is the Challenge box G, which has the smallest $b_1$, and so the neglected loops are the smallest.

\footnote{16}Of course, due to modeling issues, it is unlikely that this assumption is entirely true.
4.3 Physical considerations

We have seen above that we can measure the three cosmological parameters $A_s$, $\Omega_m$, and $h$, without any significant theory-systematic error. As we will see, this measurement can be done also on the SDSS/BOSS data. Given that the official analysis of the SDSS/BOSS data do not provide independent measurements of all of these quantities without the addition of external priors, here we briefly highlight why it is actually possible to measure them.\footnote{We stress again that we are fixing $n_s$ and $\Omega_b/\Omega_c$ to the fiducial value of the simulations, or, for the data, to the central value from Planck2018. However, we fix these two cosmological parameters just for computational reasons, the analysis could be generalized to include them as well. We discussed earlier on the implication of this choice on the error bars of the parameters.}

First, we remind the reader that we are working in the $\Lambda$CDM setting. Thanks to the accurate modeling of LSS provided by the EFTofLSS, we are allowed to use, implicitly or explicitly, all the relations among quantities that are implied by this (as for example the dependence of $f$ on $\Omega_m$ from eq. (2.4), or the dependence of the sound horizon at decoupling on $\Omega_m$ that is implicitly encoded in the transfer functions), without the need to introduce artificial free parameters beyond the ones given by the EFTofLSS. In this sense, our analysis is identical to the standard CMB analysis of Planck or WMAP.

We also remind the reader that we are analyzing all the simulation data up to $k_{\text{max}} = 0.2h$ Mpc$^{-1}$ and $0.23h$ Mpc$^{-1}$, without making any splitting between oscillating and broadband signal. However, for the purpose of explaining the physical origin of the signal, we can separate the part of the signal that originates from the BAO and from the broad-band shape of the power spectrum.

Figure 9. 2D posterior distributions of the cosmological parameters $\ln(10^{10} A_s)$, $\Omega_m$, and $h$, and the EFT parameters $b_1$ and $c_2$, obtained from fitting the power spectrum up to $k_{\text{max}} = 0.2h$ Mpc$^{-1}$ of the challenge boxes ABFG (left). On the right, we plot the same quantities for box A, as a function of $k_{\text{max}}$ (right). We can see that we can measure all the cosmological parameters as well as the EFT parameters. All degeneracies are broken, though some to a different extent than others. We see that the four HOD models on the left differ mainly for their value of $b_1$. The results for the cosmological parameters, which should be equal for all of them, appear to be remarkably consistent.
Focusing on the BAO signal, this is well approximated by an oscillating signal in the power spectrum, corresponding to a peak in the real space correlation function. The wavelength of this oscillation, equivalent to the location of the peak in real space, is the sound horizon at the end of the baryon-drag epoch. In the case of the CMB, the observed quantity is the angle at which this peak appears, $\theta_{\text{CMB}}$, which scales as $\theta_{\text{CMB}} \simeq r_d(z_{\text{CMB}})/D_A(z_{\text{CMB}})$, where $r_d(z_{\text{CMB}})$ is the sound horizon at the time of decoupling, $z_{\text{CMB}} \sim 1100$, and $D_A$ is the angular diameter distance. Ref. [83] (see section 4.4) carefully derives that such a measurement is mainly dependent on the combination $\Omega_m$. In the case of LSS, the angle (or ratio of length scales) under which the BAO oscillation is observed, $\theta_{\text{LSS}}$, scales as the geometric mean of what is observed parallel and perpendicular to the line of sight (see for example [81, 82])

$$\theta_{\text{LSS}} \simeq \frac{r_d(z_{\text{CMB}})}{(D_A(z_{\text{LSS}})^2 \cdot c z_{\text{LSS}}/H(z_{\text{LSS}}))^{1/3}}.$$ \hspace{1cm} (4.1)

Straightforwardly repeating the analysis of [83], we find that,

$$\frac{\partial \log \theta_{\text{LSS}}}{\partial \log \Omega_m} \bigg|_{\Omega_m=0.3, \ h=0.7} = -0.12, \quad \frac{\partial \log \theta_{\text{LSS}}}{\partial \log h} \bigg|_{\Omega_m=0.3, \ h=0.7} = 0.48,$$ \hspace{1cm} (4.2)

for $z_{\text{LSS}} = 0.5$. $\theta_{\text{LSS}}$ therefore mainly depends on the combination $\Omega_m h^{-4}$.18

The relative amplitude of the BAO peak with respect to the smooth part instead gives a measurement of $\sim \Omega_m h^2$. This is quite intuitive as the amplitude of the oscillating part, unlike its wavelength, is not affected by projection effects, and therefore simply scales as the density of baryons and dark matter at the time of recombination, which, in the case of fixed $\Omega_b/\Omega_c$, simply scales as $\Omega_m h^2$ (see [85] for a very pedagogical review and derivation).

We now pass to include the broad-band signal. The linear power spectrum monopole and quadrupole give a measurement of $b_1^2 A_s^{(k_{\text{max}})}$ and $b_1 f A_s^{(k_{\text{max}})} \sim b_1 \Omega_m^{1/3} A_s^{(k_{\text{max}})}$. Here $A_s^{(k_{\text{max}})}$ represents the amplitude of the linearly evolved power spectrum at the maximum wavenumber of our analysis, as this is where the signal peaks. $A_s^{(k_{\text{max}})}$ is not quite the same as the primordial amplitude, $A_s$, because the evolution of modes that enter the horizon before or after matter-radiation equality is different. In fact, a scale-invariant primordial power spectrum decays, roughly, as $k^{-2}$ around the scales of interest (see for example [4]). We therefore have:

$$A_s^{(k_{\text{max}})} \equiv A_s \left(\frac{k_{\text{eq}}}{k_{\text{max}}}\right)^2,$$ \hspace{1cm} (4.4)

where $k_{\text{eq}}$ is the wavenumber that enters the horizon at matter-radiation equality, and scales as $k_{\text{eq}}[h \text{ Mpc}^{-1}] \propto (\Omega_m h^2)^{1/2}$. We also approximated $f$ with the well known approximate

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18 This derivation applies if we keep $\Omega_b$ as an independent parameter. If the ratio $\Omega_b/\Omega_c$ is held fixed, then the scaling is rather $\Omega_m h^5$ (see eq. (11) of [84]). However, for the purpose of our general discussion, we are only interested in describing how the degeneracy between $\Omega_m$ and $h$ can be broken by single-redshift LSS observations alone, and we therefore focus on the more general case where $\Omega_c$ is left as a free parameter.

19 Eq. (4.1) summarizes the information that we acquire if we only observe the monopole. In reality the parallel and perpendicular directions to the line of sight can be disentangled by observing also the quadrupole. In this case, there are two angles or ratios of length scales that are observed:

$$\theta_{\text{LSS}, \parallel} \simeq \frac{r_d(z_{\text{CMB}})}{D_A(z_{\text{LSS}})}, \quad \theta_{\text{LSS}, \perp} \simeq \frac{r_d(z_{\text{CMB}})}{c z_{\text{LSS}}/H(z_{\text{LSS}})}.$$ \hspace{1cm} (4.3)

$\theta_{\text{LSS}, \parallel}$ mainly depends on the combination $\Omega_m h^{-3}$, while $\theta_{\text{LSS}, \perp}$ mainly depends on the combination $\Omega_m h^{-10}$. This also allows to break the degeneracy between $\Omega_m$ and $h$. 

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relation $f(z) \sim \Omega_m(z)^{4/7}$ (see, for example, [86]), which, for $z \simeq 0.5$ and accounting for the $z$-dependence of the matter density, gives $f(z \simeq 0.5) \sim \Omega_m(z = 0)^{1/3}$.

These scalings allow us to solve for $b_1 \propto \Omega_m^{1/3}$, $A_s^{(k_{\text{max}})} \propto b_1^{-2} \propto \Omega_m^{-2/3}$. The constraint on the sound horizon from the BAO gives $h \propto \Omega_m^{1/4}$. Having expressed $b_1, h$, and $A_s^{(k_{\text{max}})}$ in terms of $\Omega_m$, we can finally use the amplitude of the BAO peak, that, after substituting the scaling of $h$, goes as $\propto \Omega_m^{1/2}$, to determine $\Omega_m$ and, in consequence, $b_1, h$, and $A_s^{(k_{\text{max}})}$ and, finally, from eq. (4.4), $A_s$.

On top of this information, the power spectrum has a broad shape which also carries information. For example, the power spectrum changes slope for $k$’s around $k_{\text{eq}}$. This carries further information about the combination $\Omega_m h^2$. Furthermore, the nonlinear power spectrum additionally contributes to lower the error bars on the cosmological parameters, as, though there are many new EFT parameters that are introduced at this stage, each contribution carries a different functional shape dependence, and so can be measured, at least in principle, independently. Since the number of shapes is larger than the number of parameters, this adds information.

The way the cosmological and EFT parameters are determined can indeed be recognized by the contours of figure 9. The constraint on the linear power spectrum from the monopole clearly must play a leading role, due to the smallness of the error bars. This amplitude scales as $b_1^2 A_s^{(k_{\text{max}})} \sim b_1^2 A_s k_{\text{eq}}^2 \propto b_1^2 A_s \Omega_m h^2$. This explains the anticorrelated degeneracies between $A_s$ & $b_1$, between $A_s$ & $h$. As just described, all the degeneracies are broken by the information from the quadrupole and the BAO, but in general in a subleading way so that these (anti)correlated degeneracies are still measurable. This might explain why we do not find a strong anticorrelation in $\Omega_m h$, but maybe, at least in the Challenge boxes, a slight correlation between $\Omega_m$ and $h$. The scaling of the observed BAO angle in LSS, $\theta_{\text{LSS}}$, in fact might explain this.

The fact that we are able to independently determine the cosmological parameters allows, at least in principle, for another interesting application of our results to the realm of astrophysics, beyond the study of primordial cosmology. In fact, the breaking of the degeneracy between $A_s$ and $b_1$ allows for a measurement of $b_1$. As it is evident from the left panel of figure 9, we can also measure the other EFT parameters. But figure 9 tells us that the bias $b_1$ correlates with the specific HOD model, and, in general, all the EFT parameters correlate with it. This means that by measuring the power spectrum, we can infer the EFT parameters, and, from them, the galaxy formation mechanism that is at play. Of course the success of this potential application relies on the capability of relating the bias parameters to a particular HOD model, and the HOD model to some galaxy formation physics mechanism. We leave the exploration of these possibilities to future work.

### 4.4 Inclusion of Planck sound horizon prior

As we have just discussed in section 4.3, the measurement of the sound horizon at decoupling using the BAO is very important in order for us to measure the cosmological parameters, as it strongly constraints the linear combination $\Omega_m/h^4$. While we have just shown that this can be done using only redshift-space clustering data, it is a fact that the CMB experiments, and the Planck satellite most lately, have provided exquisite measurement of the sound horizon at decoupling, or rather of the angle $\theta_{\text{CMB}}$ described earlier, to per-mille precision. It is therefore interesting to use this measurement as a prior for the LSS analysis. In fact, contrary to other
The combination of the LSS data with the CMB-measured sound horizon has a long history of important applications in the community, dating back to the first measurements of this effect (see for example [81, 82, 87] and references therein), and it is normally considered essential to obtain constraints on the Hubble parameter from LSS data. However, contrary to the usual cases, here we will use it simply to improve our constraints, as we already have interesting constraints, and in general to explore the effect of such a CMB-motivated prior.

As we saw in the former section, the theory-systematic error that we measured is much smaller than the statistical error of BOSS DR12. However, once we apply the Planck prior on the sound horizon, one might worry that our statistical errors will drop and our theoretical systematics, which naively will not change, will dominate the error budget. This motivates us to test the EFTofLSS against the simulations again, using now a prior on the sound horizon. We refer to appendix E for details on the formulas for the sound horizon at decoupling and for the numerical value of the prior we use.

The reason why the result of this test is non-trivial is because it can happen that the Planck prior will reduce not only the statistical error, but also the theory-systematic error. On the right of figure 7, we plot the marginalized likelihood for the cosmological parameters, after the inclusion of the Planck prior on the sound horizon. In figure 10 we plot the contours of the posterior for the partially marginalized likelihood. By focusing on Challenge ABFG and D for definiteness, we see that, typically, the statistical error on $\ln(10^{10}A_s)$ has been reduced in a marginal way, the one on $\Omega_m$ by about around 25% on average, while the one on $h$ is reduced by about a factor of two. In terms of theoretical systematic error, on ABFG
we find overall tolerably small systematic errors for \( k_{\text{max}} \leq 0.23h\,\text{Mpc}^{-1} \):

- On \( \ln(10^{10}A_s) \), \( \sigma_{\text{sys}} = 0.02 \lesssim \sigma_{\text{stat,CMASS}}/7 \) at \( k_{\text{max}} = 0.15h\,\text{Mpc}^{-1} \) and \( k_{\text{max}} = 0.2h\,\text{Mpc}^{-1} \), \( \sigma_{\text{sys}} = 0.04 \lesssim \sigma_{\text{stat,CMASS x low-z}}/3 \) at \( k_{\text{max}} = 0.23h\,\text{Mpc}^{-1} \), and \( \sigma_{\text{sys}} = 0.08 \lesssim 0.6 \cdot \sigma_{\text{stat,CMASS x low-z}} \) at \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \);

- On \( \Omega_m \), \( \sigma_{\text{sys}} = 0.006 \lesssim 0.4 \sigma_{\text{stat,CMASS}} \) at \( k_{\text{max}} = 0.15h\,\text{Mpc}^{-1} \), \( \sigma_{\text{sys}} = 0.002 \approx \sigma_{\text{stat,CMASS}}/5 \) at \( k_{\text{max}} = 0.2h\,\text{Mpc}^{-1} \) and \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \), and \( \sigma_{\text{sys}} = 0.001 \approx \sigma_{\text{stat,CMASS x low-z}}/9 \) at \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \);

- On \( h \), \( \sigma_{\text{sys}} = 0.006 \approx 0.4 \sigma_{\text{stat,CMASS}} \) at \( k_{\text{max}} = 0.15h\,\text{Mpc}^{-1} \), \( \sigma_{\text{sys}} = 0.002 \approx \sigma_{\text{stat,CMASS}}/6 \) at \( k_{\text{max}} = 0.2h\,\text{Mpc}^{-1} \) and \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \), and \( \sigma_{\text{sys}} = 0.001 \approx \sigma_{\text{stat,CMASS x low-z}}/10 \) at \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \).

The potentially dangerous systematic errors are at \( k_{\text{max}} = 0.15h\,\text{Mpc}^{-1} \) on \( \Omega_m \) with \( \sigma_{\text{sys}} = 0.006 \approx 0.4 \sigma_{\text{stat,CMASS}} \) and on \( h \) with \( \sigma_{\text{sys}} = 0.006 \approx 0.4 \sigma_{\text{stat,CMASS}} \); but we assume that at such a low \( k_{\text{max}} \) they are due to a statistical fluctuation. At \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \) we find a large systematic error on \( \ln(10^{10}A_s) \) of \( \sigma_{\text{sys}} = 0.08 \), and a marginally important one at \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \).

On box D, we detect vanishing errors for \( \Omega_m \) and \( h \) at all \( k_{\text{max}} \)'s. On \( \ln(10^{10}A_s) \), we find a theoretical systematic error that grows with \( k_{\text{max}} \). It starts with \( \sigma_{\text{sys}} = 0.03 \approx \sigma_{\text{stat,CMASS}}/5 \) at \( k_{\text{max}} = 0.15h\,\text{Mpc}^{-1} \), then we find \( \sigma_{\text{sys}} = 0.05 \approx \sigma_{\text{stat,CMASS}}/3 \) at \( k_{\text{max}} = 0.2h\,\text{Mpc}^{-1} \) and becomes \( \approx \sigma_{\text{stat,CMASS}}/2 \) at \( k_{\text{max}} = 0.23h\,\text{Mpc}^{-1} \) and \( \sigma_{\text{sys}} = 0.1 \approx 0.8 \sigma_{\text{stat,CMASS}} \) at \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \).

When we combine the measurements of ABFG & D, we find on \( \Omega_m \) and \( h \) negligible small systematic errors with respect to what we find in the data (\( \sigma_{\text{sys}} \leq \sigma_{\text{stat,CMASS}}/5 \)). On \( \ln(10^{10}A_s) \), we find a large systematic error of \( \sigma_{\text{sys}} = 0.11 \) at \( k_{\text{max}} = 0.25h\,\text{Mpc}^{-1} \), while we have \( \sigma_{\text{sys}} = 0.07 \) at \( k_{\text{max}} = 0.23h\,\text{Mpc}^{-1} \) and \( \sigma_{\text{sys}} = 0.05 \approx 0.4 \sigma_{\text{stat,CMASS}} \) at \( k_{\text{max}} = 0.15h\,\text{Mpc}^{-1} \) and \( k_{\text{max}} = 0.2h\,\text{Mpc}^{-1} \). Though the systematic error so detected in \( A_s \) at \( k_{\text{max}} = 0.20h\,\text{Mpc}^{-1} \) is not completely negligible, we notice that it is the same already at \( k_{\text{max}} = 0.15h\,\text{Mpc}^{-1} \), so we assume that its detected magnitude is affected by a statistical fluctuation, and so we neglect it and proceed with the analysis at \( k_{\text{max}} = 0.20 \) by assuming negligible systematic error.

It is worthwhile to explain why the systematic error has in some cases decreased, after the inclusion of the Planck sound horizon prior. This is naively surprising given the reduction in the statistical error, which also affects the size of the smallest systematic error we can detect. The reason is explained by figure 11. In the absence of any prior, the theoretical systematic error can be represented as a vector living in the space of the cosmological parameters that goes from the true value to the average maximum likelihood point as obtained from the MCMC’s. \( \sigma_{\text{sys}} \) for each parameter is nothing but the component of this vector in the corresponding direction (taking also into account the uncertainty associated with the finite number of simulations at our disposal). If we impose a prior, for example on the sound horizon at decoupling, the likelihood is forced to have support only on a region that is narrower or equal to the one without prior. Therefore, the resulting theoretical-systematic error that is obtained after imposing a prior is equal to the one we have without any prior, projected to the region where the prior has support. This reduces the systematic error, and, in general, it suggests that great care should be used when estimating how the systematic error changes after the addition of priors.
Figure 11. In the absence of any prior, the theoretical systematic error can be represented as a vector living in the space of the cosmological parameters that goes from the true value to the maximum likelihood point as obtained from the MCMC’s. $\sigma_{\text{sys}}$ for each parameter is the component of this vector in the corresponding direction. If we impose a prior, the likelihood is forced to have support only on a region that is narrower or equal to the one without prior. Therefore, roughly, the resulting theoretical-systematic error obtained after imposing a prior is equal to the one we have without any prior, projected on the region supported by the prior. This tends to reduce the systematic error. On the $\Omega_m-h$ plane, we plot the level contour of the Planck sound horizon at decoupling $r_d$.

These considerations suggest an effective way to include the theoretical error for a given $k_{\text{max}}$ without the use of simulations but taking into account the effect induced by the priors. One could add to the EFTofLSS prediction evaluated at a given order, i.e. the model, a reasonable expression for the functional form of a higher-order term, for example $c_{\text{higher}}(k/k_{\text{NL}})^4 P_{11}(k)$, whose amplitude is parametrized by a parameter, here $c_{\text{higher}}$. Then one can run the MCMC including the new contribution, with a prior on this parameter, here $c_{\text{higher}}$, to be a number of order one. One can then measure the resulting shift induced on the maximum likelihood point for a cosmological parameter, and call that number the theory-systematic error. We leave the development of this potential method to estimate the theoretical systematic error to future work.

4.5 Inclusion of bispectrum

We now discuss the effect of the inclusion of the bispectrum monopole in the likelihood, obtaining the likelihood given by eq. (D.14) in appendix D. Here we use the Patchy mock catalogues to test our analysis pipeline, as we have measurements of the bispectrum monopole only for this kind of simulations.

We point out that, as discussed at the end of appendix D, the application of the window function to the bispectrum is done approximately, and this might induce additional systematic error bars. In order to better limit the systematic error induced by the window function, we analyze the bispectrum monopole including only triangles whose minimum wavenumber is $k_{\text{min}} = 0.04h\,\text{Mpc}^{-1}$, and whose maximum one is either $k_{\text{max}} = 0.08h\,\text{Mpc}^{-1}$ or $k_{\text{max}} = 0.1h\,\text{Mpc}^{-1}$, adding 10 and 34 triangles, respectively. For the power spectrum, we use $k_{\text{max}} = 0.23h\,\text{Mpc}^{-1}$, when analyzed in combination with the bispectrum, though we noticed earlier that in this case the systematic error might be marginally important.\footnote{Since the theoretical error grows with $k_{\text{max}}$, and since we will have negligible systematic error on simulation, this is justified.}

When analyzing the Patchy mocks using the partially marginalized likelihood of eq. (D.14), we obtain the results presented in figure 12. In the same figure, as we did in...
Figure 12. Marginalized posterior distributions of the cosmological parameters obtained from fitting the power spectrum and the bispectrum monopole of 16 Patchy lightcones NGC (left, thin lines) and their average (left, thick lines), and with the inclusion of Planck sound horizon prior (right). We analyze the bispectrum either with $k_{\text{max}} = 0.08h\text{ Mpc}^{-1}$ or with $0.1h\text{ Mpc}^{-1}$. For each cosmological parameter, the expectation value of the product of the posteriors is given together with the statistical error and the theory-systematic error (from top to bottom in each subplot).

section 4.2, we also plot the product of the 1-D posteriors, which is a non-rigorous procedure, given that we should multiply the full posteriors, but we checked that, given our error bars, it is quite a conservative procedure. We see that the error bars are reduced in a marginal way by the inclusion of the bispectrum monopole both with and without the inclusion of the Planck prior. We detect negligible systematic error upon the inclusion of the bispectrum monopole, though it is barely so when the bispectrum is analyzed up to $k_{\text{max}} = 0.1h\text{ Mpc}^{-1}$.

We also point out that the improvement in the determination of the EFT parameters is limited. Comparing the statistical errors between the cases with and without the bispectrum in table 2, we see that we get a marginal improvement on $A_s$ of about 10% and no improvement on $\Omega_m$ and $h$. Furthermore, we have checked that the improvement in the measurement of the EFT-parameters is much more significant if one were to use larger priors for the same parameters.

Our conclusions about the inclusion of the bispectrum are clearly limited by several factors: the fact that we do not perform a high order calculation does not allow us to go to high wavenumbers; our approximate implementation of the window function forces us to exclude low wavenumbers. It would be interesting to improve in all of these aspects in future work.

4.6 Summary of tests on simulations

We summarize the results of our tests on numerical simulations in table 2. We find that we can determine $A_s$, $\Omega_m$ and $h$ from the power spectrum data, with no significant role played by other observational data. We also can measure several of the EFT parameters, allowing, in principle, to learn about the galaxy formation mechanism.

We give in table 3 the best-fit results from the fit of the power spectrum up to $k_{\text{max}} = 0.2h\text{ Mpc}^{-1}$, from both the simulations and the BOSS DR12 data we analyze next. All minimum $\chi^2$ per degree of freedom (min $\chi^2$/d.o.f.) give an acceptable $p$-value, attesting for the goodness of the fit. The values of the EFT parameters are all of order of the linear
bias $b_1$ ($\sim O(2)$), all lying well within the chosen priors for this analysis. We remind that the Challenge boxes were run without shot noise, as this has been subtracted exactly from
Table 2. Summary of the individual analysis over simulations up to $k_{\text{max}} = 0.20\, h\, \text{Mpc}^{-1}$ for the power spectrum and up to $k_{\text{max}} = 0.1\, h\, \text{Mpc}^{-1}$ for the bispectrum monopole (in which case we use the power spectrum up to $k_{\text{max}} = 0.23\, h\, \text{Mpc}^{-1}$). As explained in the text, we use a conservative way to assess the systematic error, which is however limited to about $\sigma_{\text{stat}}$ for Challenge boxes, to $\sigma_{\text{stat}}/\sqrt{2}$ for the combination of all the Challenge boxes, and to $\sigma_{\text{stat}}/4$ for Patchy mocks.

|                                    | $\ln(10^{10} A_s)$ | $\Omega_m$ | $h$ |
|------------------------------------|---------------------|------------|-----|
|                                    | $\sigma_{\text{stat}}$ | $\sigma_{\text{sys}}$ | $\sigma_{\text{stat}}$ | $\sigma_{\text{sys}}$ | $\sigma_{\text{stat}}$ | $\sigma_{\text{sys}}$ |
| Challenge ABFG                     | 0.07 0.04           | 0.005 0.000 | 0.011 0.000 |
| Challenge ABFG with $r_d$          | 0.07 0.04           | 0.004 0.001 | 0.005 0.001 |
| Challenge D                        | 0.06 0.02           | 0.005 0.000 | 0.012 0.000 |
| Challenge D with $r_d$             | 0.06 0.05           | 0.004 0.000 | 0.005 0.000 |
| Challenge ABFG & D                 | 0.05 0.036          | 0.004 0.002 | 0.008 0.000 |
| Challenge ABFG & D with $r_d$      | 0.05 0.05           | 0.003 0.001 | 0.004 0.001 |
| Patchy NGC                         | 0.18 0.00           | 0.014 0.001 | 0.030 0.002 |
| Patchy NGC with $r_d$              | 0.16 0.00           | 0.013 0.000 | 0.014 0.000 |
| Patchy NGC with Bisp.              | 0.15 0.04           | 0.014 0.000 | 0.029 0.002 |
| Patchy NGC with Bisp. with $r_d$   | 0.14 0.03           | 0.013 0.000 | 0.015 0.000 |

The relative statistical error on the cosmological parameters on BOSS-NGC-like Patchy mocks is about 18%, 4% and 4.4% for $A_s$, $\Omega_m$ and $h$ respectively at $k_{\text{max}} = 0.2h\, \text{Mpc}^{-1}$. The inclusion of the bispectrum monopole reduces the error bars marginally. Upon inclusion of the Planck prior on the sound horizon, the error bars are decreased by roughly about 10%, 20% and 50% for $A_s$, $\Omega_m$ and $h$ respectively. We have argued that the systematic error is always comfortably smaller than the statistical errors we will find in the observational data as long as $k_{\text{max}} \lesssim 0.2h\, \text{Mpc}^{-1}$: $\lesssim 1/4 \cdot \sigma_{\text{stat}}$, CMASS \times low−z from the Challenge boxes analyses for all cosmological parameters, and on $h \lesssim 1/4 \cdot \sigma_{\text{stat}}$, CMASS \times low−z from the analyses of the Patchy mocks (otherwise vanishing for the other cosmological parameters). These are negligibly small as long as $k_{\text{max}} \lesssim 0.20h\, \text{Mpc}^{-1}$, which we therefore take as the maximum wavenumber for our CMASS analysis.

We also observe how the statistical error decreases with $k_{\text{max}}$. Ideally, the errors should decrease as $k_{\text{max}}^{-3/2}$, but this is affected by the fact that some parameters are dominated by the information in the BAO peak, which is already well measured at low $k_{\text{max}}$, as well as by the fact that there are degeneracies with some cosmological and EFT parameters. Taking for definiteness the power spectrum-only analysis of the Challenge ABFG boxes, we find that the error decreases roughly as $k_{\text{max}}^{-0.4}$, $k_{\text{max}}^{-0.7}$ and $k_{\text{max}}^{-0.6}$ for $A_s$, $\Omega_m$ and $h$, respectively.\textsuperscript{21}

\textsuperscript{21}In order to check, at least partially, the correctness of our pipeline, we have performed a Fisher-matrix study of the expected error bars on the parameters around the best fit point for the power spectrum analysis for Challenge box A. We find results similar, though somewhat smaller than expected, to the ones we find with the MCMC. We notice here that we also performed the same Fisher-matrix analysis around the best fit
**Best fits**

|               | \(\ln(10^{10} A_s)\) | \(\Omega_m\) | \(h\) | \(b_1\) | \(\text{min } \chi^2/\text{d.o.f.}\) | \(p\)-value |
|---------------|------------------------|----------------|------|-------|-----------------------------------|-----------|
| Challenge A   | 3.18                   | 0.317          | 0.668| 1.94  | 50/(38-9)                         | 0.007     |
| Challenge B   | 3.24                   | 0.312          | 0.671| 1.96  | 50/(38-9)                         | 0.008     |
| Challenge F   | 3.17                   | 0.314          | 0.671| 2.09  | 54/(38-9)                         | 0.003     |
| Challenge G   | 3.17                   | 0.312          | 0.673| 1.80  | 48/(38-9)                         | 0.014     |
| Challenge D   | 3.19                   | 0.290          | 0.708| 1.83  | 51/(38-9)                         | 0.006     |
| Patchy NGC box 1 | 3.12                   | 0.304          | 0.698| 1.82  | 68/(68-10)                        | 0.18      |
| data CMASS NGC | 2.70                   | 0.311          | 0.725| 2.22  | 65/(68-10)                        | 0.27      |
| data CMASS SGC | 2.98                   | 0.306          | 0.670| 2.06  | 60/(68-10)                        | 0.41      |
| data LOWZ NGC  | 3.24                   | 0.312          | 0.638| 1.73  | 69/(76-10)                        | 0.40      |
| data CMASS NGC with bisp. | 2.73                   | 0.303          | 0.686| 2.31  | 103/(68+34-11)                    | 0.18      |
|               | \(c_2/c_4\)            | \(b_3\)        | \(c_{ct}\) | \(c_r,1\) (+\(c_r,2\)) | \(c_{r,1}/\bar{n}_g\) | \(c_{r,quad}\) |
| Challenge A   | 1.0                    | -1.6           | -0.7 | -6.9  | -                                 | -4.1      |
| Challenge B   | 1.1                    | -2.0           | -1.1 | -8.6  | -                                 | -3.9      |
| Challenge F   | 1.1                    | -2.3           | -2.2 | -7.0  | -                                 | -4.2      |
| Challenge G   | 0.9                    | -1.8           | 1.1  | -6.7  | -                                 | -3.3      |
| Challenge D   | 1.0                    | -1.5           | -0.8 | -7.3  | -                                 | -3.6      |
| Patchy NGC box 1 | 1.0                    | -1.2           | 0.2  | -9.9  | 0                                 | -1.5      |
| data CMASS NGC | 1.2                    | 0.1            | 0.4  | -7.7  | 0                                 | -3.7      |
| data CMASS SGC | 1.2                    | -0.6           | -0.1 | -9.0  | 0                                 | -1.1      |
| data LOWZ NGC  | 1.0                    | -1.0           | 0.2  | -10.3 | 0                                 | -2.1      |
| data CMASS NGC with bisp. | 1.8/-0.5             | 0.2            | -2.0 | -9.5  | 0                                 | -1.1      |

**Table 3.** Best-fit values and goodness of fit of the power spectrum up to \(k_{\text{max}} = 0.2 h \text{ Mpc}^{-1}\), with \(k_0 = 0.7 h \text{ Mpc}^{-1}\) and \(\bar{n}_g = \{4.5, 4.0\} \times 10^{-4} h \text{ Mpc}^{-3}\), respectively. The values of \(\bar{n}_g\) for Challenge boxes is given in footnote 11. For \(b_1\) of challenge A, the true value of \(b_1\) is 2.04. For the observational data, the bins below \(k \lesssim 0.07\) are correlated, therefore the value of the \(\chi^2/\text{d.o.f.}\) should be taken with care. Performing a similar calculation for the data with the minimum wavenumber \(k = 0.05 h \text{ Mpc}^{-1}\) still gives good values of the fit. Regarding the \(p\)-value of the Challenge boxes, we remind that they have error bars roughly a factor of four smaller than the data we analyze in the observational data. Fitting without imposing priors results into a \(p\)-value of \(\sim 0.03\) both for Challenge A and D, while the EFT-parameters best-fit values are still of physical size. As we do not fit the hexadecapole in this analysis, we set \(c_{r,2} = 0\). Thus, its contribution (mainly in the quadrupole) is absorbed by \(c_{r,1}\), which takes overall higher values about two times its natural size.

We have also checked through the Fisher-matrix analysis that we obtain very similar error bars if we do not include the AP effect in the Fisher forecast. This is consistent with the fact that the AP effect is just a calculable geometric effect that relates how the Fourier-space power-spectra are obtained out of the sky map, and therefore should carry no cosmological information. Indeed, if we analyzed directly the three-dimensional information in the \(z\)-dependent angular correlation function, usually referred as \(C_\ell(z, z')\), there would be no need of the AP parameters and no information would be lost.

We use the same Fisher-matrix analysis to study the degradation of our error bars if we lower the \(k_{\text{max}}\) of the analysis. In particular, we find from the Fisher-matrix analysis of the CMASS NGC sample that the error bars on the parameters at \(k_{\text{max}} = 0.1 h \text{ Mpc}^{-1}\) are worse than the ones at \(k_{\text{max}} = 0.25 h \text{ Mpc}^{-1}\) by a factor of 133\%, 180\%, and 92\% for \(A_s\), \(\Omega_m\) and \(h\) respectively. Maybe even more significantly, if we do not
would be interesting to see how this scaling is modified if one adds higher $n$-point function in the analysis without the limitation we described earlier for our analysis.

It is worth to finally notice that the EFT of LSS at one-loop order is able to fit accurately, without introducing unacceptably large systematic errors, the power spectrum measured from the challenge boxes, whose volume is sixteen times larger than the effective one of BOSS NGC, up to $k_{\text{max}} = 0.2h\text{Mpc}^{-1}$. This is promising in light of the forthcoming large-size surveys.

5 Analysis of the SDSS/BOSS data

5.1 Main results

Having successfully tested our pipeline with simulations, we are ready to apply the same pipeline to the data. When we analyze the power spectrum only, we obtain the results shown in figure 14, 15, 16, 17, 20 and 23. They correspond, respectively, to the analysis of the CMASS NGC, CMASS SGC, CMASS NGC combined with SGC, LOWZ NGC, and finally to the combination of these three data sets. When we apply the bispectrum, we obtain the results presented in figure 22 and 19. The best-fit values for all parameters are given in table 3 and the 68% confidence intervals for the cosmological parameters are given in table 4. In figure 18, we plot the best-fit power spectrum for CMASS NGC, CMASS SGC and LOWZ NGC, against the BOSS DR12 measurement and the residuals. We provide the correlation among all parameters and the full 2D posterior surfaces in appendix F.

Let us give some details of the analysis that are specific to the observational data. For the AP effect, we use $\Omega_m = 0.310$ and $z = 0.57$ for CMASS and $z = 0.32$ for LOWZ. Since we know neutrinos have a mass, we analyze the data as done in Planck2018 by fixing one massive neutrino at 0.06eV. When we fit the different data sets, we keep all EFT parameters as independent. While this is completely justified as we go from the CMASS to the LOWZ sample, as the $z$-dependence of the EFT parameters is unknown, in principle CMASS NGC and SGC should have the same EFT parameters if they observe the same population of galaxies. However, following BOSS collaboration [25], we conservatively keep the bias parameters separate because we wish to account for different selection criteria among the samples.22

Our comparisons with numerical simulations have shown that we do not have any significant systematic error until $k_{\text{max}} = 0.2h\text{Mpc}^{-1}$ for $z \simeq 0.55$ (in reality, due to $k$-binning, we will use $k_{\text{max}} = 0.188h\text{Mpc}^{-1}$ at $z = 0.55$, but we will refer to it as $k_{\text{max}} = 0.2h\text{Mpc}^{-1}$ for simplicity). We will analyze also the LOWZ NGC sample, which is centered at $z \simeq 0.32$. In order to establish until which wavenumber we can trust our LOWZ prediction, we rescale the theoretical error as measured at $z = 0.55$ to $z = 0.32$, and compare it to the statistical

fix $n_s$ and, for $\Omega_b/\Omega_c$, we rather put a prior on $\Omega_b/\Omega_c$ of the order of four times the percentage error on $\Omega_b h^2$ obtained by BBN (which is about 20$\sigma$’s of Planck2018) [80], the expected error bars at $k_{\text{max}} = 0.1h\text{Mpc}^{-1}$ are increased by a factor of 3.7, 2.2, 4.2 for $A_s$, $\Omega_m$, and $h$ respectively, to roughly about 60%, 30% and 50% for $\ln A_s$, $\Omega_m$, and $h$ respectively, for CMASS NGC, making the overall measurement scarcely significant. On the other hand, as mentioned earlier, the relative degradation of the measurements when we loosen the prior on $n_s$ and $\Omega_b/\Omega_c$ is much smaller at the $k_{\text{max}}$’s of interest here.

22We observe that, however, given our priors and the size of the error bars of CMASS, all EFT parameters are statistically consistent between the CMASS NGC and SGC samples, as it is can be seen from the full 2D posterior surfaces in figure 26. We show in an appendix G the results on CMASS if the NGC and SGC samples are analyzed using the same set of EFT parameters with the exception of the shot noise. We find that results are quite consistent, with a very marginal reduction of the error bars by about 10%, and, as we will see, most significantly a slightly larger value of $A_s$ by about half a statistical error.
error of the data finding that we can fit the LOWZ data up to $k_{\text{max}} = 0.18 \, \text{h} \, \text{Mpc}^{-1}$ for the power spectrum of LOWZ. The results at $k_{\text{max}} = 0.20 \, \text{h} \, \text{Mpc}^{-1}$ for CMASS and at $k_{\text{max}} = 0.18 \, \text{h} \, \text{Mpc}^{-1}$ for LOWZ should be considered as our main results.

There are several features to discuss.

- **Size of error bars:** fitting the power spectrum up to $k_{\text{max}} = 0.20 \, \text{h} \, \text{Mpc}^{-1}$ for the CMASS sample and $k_{\text{max}} = 0.18 \, \text{h} \, \text{Mpc}^{-1}$ for the LOWZ sample, determines $A_s$, $\Omega_m$ and $h$ to about 13%, 3.2% and 3.2% respectively, as seen from figure 17. If we add the Planck prior on the sound horizon, $A_s$, $\Omega_m$ and $h$ are determined to about 13%, 2.9% and 1.9%, respectively. This is quite consistent with what we see in simulations. Especially, the statistical errors on CMASS NGC are quite similar to the ones we find in Patchy mocks, see figure 14.

- **Goodness of fit:** the best fit model has a good p-value and the EFT parameters are well within what is theoretically expected, see table 3, and comparable to what we see in simulations. From figure 18 we do not see any surprising feature in the comparison of the data to the model.

- **Consistency of the datasets:** we obtain consistent constraints on the cosmological parameters from the power spectrum-only analysis as we change the data set, with and without the inclusion of the Planck sound horizon prior. The change in the statistical error as we include the Planck sound-horizon prior is quite consistent with what we saw in simulations. More in detail, if we look at figure 19, we can see that the 1-σ

\begin{table}
| CMASS NGC | $\ln(10^{10} A_s)$ | $\Omega_m$ | $h$ |
|-----------|------------------|-----------|-----|
|           | $2.60 \pm 0.18$  | $0.309 \pm 0.013$ | $0.722 \pm 0.032$ |
| CMASS NGC with $r_d$ | $2.67 \pm 0.17$  | $0.304 \pm 0.012$ | $0.684 \pm 0.013$ |
| CMASS SGC | $2.75 \pm 0.27$  | $0.302 \pm 0.023$ | $0.658 \pm 0.045$ |
| CMASS SGC with $r_d$ | $2.71 \pm 0.26$  | $0.306 \pm 0.022$ | $0.681 \pm 0.024$ |
| CMASS NGC x SGC | $2.65 \pm 0.15$  | $0.307 \pm 0.011$ | $0.703 \pm 0.026$ |
| CMASS NGC x SGC with $r_d$ | $2.68 \pm 0.14$  | $0.305 \pm 0.010$ | $0.684 \pm 0.012$ |
| LOWZ NGC | $2.93 \pm 0.28$  | $0.319 \pm 0.024$ | $0.622 \pm 0.043$ |
| LOWZ NGC with $r_d$ | $2.88 \pm 0.27$  | $0.327 \pm 0.023$ | $0.659 \pm 0.024$ |
| CMASS NGC x SGC x LOWZ NGC | $2.72 \pm 0.13$  | $0.309 \pm 0.010$ | $0.685 \pm 0.022$ |
| CMASS NGC x SGC x LOWZ NGC with $r_d$ | $2.73 \pm 0.13$  | $0.309 \pm 0.009$ | $0.678 \pm 0.011$ |

Table 4. 68% confidence interval for the cosmological parameters from the individual analyses over the CMASS sample and LOWZ NGC sample of the BOSS data up to $k_{\text{max}} = 0.20 \, \text{h} \, \text{Mpc}^{-1}$ for the power spectrum of CMASS, up to $k_{\text{max}} = 0.18 \, \text{h} \, \text{Mpc}^{-1}$ for the power spectrum of LOWZ.
Figure 14. Marginalized posterior distributions of the cosmological parameters obtained from fitting the power spectrum of the CMASS sample NGC (top) or SGC (bottom) at $z_{\text{eff}} = 0.55$ without (left), and with (right) the inclusion of Planck sound horizon prior (right). The green lines represent the 1-$\sigma$ constraints from Planck2018 while the orange lines represent the constraints from WMAP9yrs.

contours in the 2D posterior distributions for both the cosmological parameters and the EFT-parameters intersect or barely miss to do so, with CMASS SGC and LOWZ NGC having a very strong overlap. This consistency remains true at the level of the 1D posterior distributions.
Figure 15. Marginalized posterior distributions of the cosmological parameters obtained from fitting the power spectrum of the CMASS sample NGC×SGC at $z_{\text{eff}} = 0.55$ (left), and with the inclusion of Planck sound horizon prior. The green lines represent the 1-$\sigma$ constraints from Planck2018 while the orange lines represent the constraints from WMAP9yrs.

- **Comparison of error bars with CMB, supernovae and strong lensing:** in figure 17, we plot our results for the posterior distribution of the cosmological parameters together with the 68% confidence interval from Planck2018 [84] and WMAP9yr [89]. We see that while our measurements are clearly less precise than the ones of Planck2018 for $A_s$ and $h$, while the error bar is very similar for $\Omega_m$. Once we compare to WMAP9yr, we find that the size of our error bars is similar for $h$ and smaller for $\Omega_m$. Furthermore, our error bar on $h$ is only about 20% worse than the one obtained with strong lensing [90] and with the latest Supernovae results [91]. Given that, roughly speaking, BOSS is currently not the latest generation LSS experiment (in comparison with DES, DESI, etc.) similarly to how WMAP is not currently the latest generation CMB experiment, this comparison with WMAP9yr (and also with Planck2018) is highly suggestive and bears lots of hopeful expectations for the potentialities of LSS science in the next decade or two.

We remind the reader that the accurate modeling of LSS provided by the EFTofLSS has allowed us to use all the relations among parameters implied by the $\Lambda$CDM cosmology (such as for example the use of the transfer functions or the dependence of $f$ on $\Omega_m$), without the need to add some artificial free parametrization beyond the one provided by the EFTofLSS. In this sense, our analysis is identical to the standard CMB ones, such as those of WMAP and Planck.

- **Statistical compatibility with CMB:** as can be seen from figure 17 and table 5, without the inclusion of the Planck prior on the sound horizon, the preferred values of the cosmological parameters are in statistical agreement with the ones measured from Planck2018.
LOWZ NGC

Figure 16. Marginalized posterior distributions of the cosmological parameters obtained from fitting the power spectrum of LOWZ NGC at \( z_{\text{eff}} = 0.32 \), without (left) and with (right) the inclusion of Planck sound horizon prior. The green lines represent the 1-\( \sigma \) constraints from Planck2018 while the orange lines represent the constraints from WMAP9yr.

Our value of \( \ln(10^{10}A_s) \) is slightly, 2.3\( \sigma \), lower than the one measured in Planck2018, while the values of \( \Omega_m \) and \( h \) are extremely consistent with CMB measurements. Though statistically non-significant, the trend we see in our results is in accordance with trends found in other LSS measurements (see for example results from the DES collaboration [88]). For BOSS, this is in the same direction as the measurement of \( f\sigma_8 \) from of the high redshift bin [23]. By looking at figure 19, we see that, in the 2D posteriors, the push to lower values of \( A_s \) is driven by the CMASS NGC sample.

- **Effect of Planck sound-horizon prior:** as expected, the central values of the posterior distribution for the cosmological parameters is slightly pushed towards Planck2018 values when we add the Planck prior on the sound horizon. The compatibility with Planck2018 is similar to the case of no sound horizon prior.

- **\( f\sigma_8 \) and comparison with BOSS and DES standard analyses:** The derived quantity \( f\sigma_8 \) at \( z = 0 \) from our analysis of the power spectrum up to \( k_{\text{max}} = 0.20 h \text{Mpc}^{-1} \) for CMASS and up to \( k_{\text{max}} = 0.18 \) for LOWZ NGC reads: \( f(z = 0)\sigma_8(z = 0) = 0.363 \pm 0.026 \). If use only the CMASS sample, we obtain the constraint on this quantity at \( z = 0.55 \) to be \( f(z = 0.55)\sigma_8(z = 0.55) = 0.399 \pm 0.031 \). We remind the reader that this is a derived quantity for us, as, since we assume ΛCDM and all the relations implied by that, our natural parameters as \( A_s \), \( \Omega_m \) and \( h \). The measurement does not improve significantly if we add the Planck prior on the sound horizon. Our statistical error is roughly 20\% better than the bound obtained by the final BOSS power spectrum analysis in [25, 70], \( f(z = 0.57)\sigma_8(z = 0.57) = 0.444 \pm 0.038 \),...
and our central value is a bit lower, though in a statistically compatible way. We plot the 2D posterior distribution for $h$ and $f_{\sigma_8}$ in figure 21. The origin of our improvement in the constraint in $f_{\sigma_8}$, as well as in our measurement of the other cosmological parameters, is due to a combination of factors. First, the EFTofLSS allows us to reach higher wavenumbers. Second, we update the full shape of the power spectrum as we change the cosmological parameters, while in [25, 70] a reference shape is fixed, and only the AP parameters are updated as one changes the cosmological parameters. This does not allow to exploit all the available information without including sizable systematic errors. Just to mention an example, in our case the logarithmic derivative of the growth factor $f$ is appropriately updated to the cosmology dependent $\Lambda$CDM value.

In figure 21, we also plot the 2D posterior distribution for $\Omega_m$ and the quantity $S_{\delta} = \sigma_8 (\Omega_m/0.3)^{0.5}$. The 1D posterior distribution reads, at $z = 0$, $S_{\delta} = 0.704 \pm 0.051$ and is lower than the one of DES1Yr [88], though, again, in a statistically very compatible way.

- **Inclusion of bispectrum:** finally, when we include the bispectrum monopole in the analysis of CMASS NGC, we find the results presented in figure 22 and 23. As from table 3, the $p$-value of the best fit is 0.29. The values of the EFT-parameters and of the residuals, that we do not show to avoid clutter, do not show any particular feature.
The bispectrum monopole brings no significant statistical improvement on our results, quite consistent with what we saw in the Patchy mocks. The marginality of the improvement could be originating from several limitations in our bispectrum analysis that we mentioned in section 4.5. We therefore do not use these data in our final constraints, though we here highlight its importance as a complementary statistical probe.

• **Comments of potential systematic errors:** we finally note that the unprecedented precision of these measurements, as well as the breaking of previously unbroken degeneracies, should warn us against possibly yet unmeasured observational systematics. Though we find no issues on the theory side and in the comparison with the simulations, we make no assessment of the potential observational systematics that could have become relevant. To fully trust our findings, a reassessment of the observational systematic error is quite required.

5.2 Neutrino mass

We conclude the section on data analysis by highlighting an important fact that is shown by our analysis. It is well known that the leading effect of the non-vanishing of the neutrino masses is to induce a damping in the amplitude of the evolved power spectrum at distances shorter than the free-streaming scale of the neutrinos. Since it is in general believed to be
Table 5. Value of the probability (i.e. \( p \)-value) for the closest 1σ value of the cosmological parameters as measured by Planck2018, with respect to the individual analyses of the BOSS data up to \( k_{\text{max}} = 0.2h \text{ Mpc}^{-1} \) for the CMASS and LOWZ power spectrum, respectively, and \( k_{\text{max}} = 0.1h \text{ Mpc}^{-1} \) for the bispectrum monopole. We also include the effective number of \( \sigma \)'s such a \( p \)-value would correspond to in a putative Gaussian distribution. We conclude that our results are consistent with Planck2018, with the largest, very-mild, discrepancy present in \( \ln(10^{10}A_s) \) only in the case we include the bispectrum in the analysis.

|                   | \( \ln(10^{10}A_s) \) | \( \Omega_m \) | \( h \) | \( \ln(10^{10}A_s) \) | \( \Omega_m \) | \( h \) |
|-------------------|----------------------|---------------|-------|----------------------|---------------|-------|
| CMASS NGC         | 0.02                 | 1.00          | 0.18  | 2.3                  | 0.0           | 1.3   |
| CMASS NGC with \( r_d \) | 0.04                 | 0.72          | 0.69  | 2.1                  | 0.4           | 0.4   |
| CMASS SGC         | 0.29                 | 0.77          | 0.82  | 1.1                  | 0.3           | 0.2   |
| CMASS SGC with \( r_d \) | 0.21                 | 0.93          | 0.92  | 1.3                  | 0.1           | 0.1   |
| CMASS NGC x SGC   | 0.014                | 0.91          | 0.37  | 2.46                 | 0.1           | 0.9   |
| CMASS NGC x SGC with \( r_d \) | 0.02                 | 0.74          | 0.64  | 2.3                  | 0.3           | 0.5   |
| LOWZ NGC          | 0.96                 | 1.00          | 0.33  | 0.1                  | 0.0           | 1.0   |
| LOWZ NGC with \( r_d \) | 0.73                 | 1.00          | 1.00  | 0.3                  | 0.0           | 0.0   |
| CMASS NGC x SGC x LOWZ NGC | 0.02                 | 1.00          | 0.78  | 2.3                  | 0.0           | 0.3   |
| CMASS NGC x SGC x LOWZ NGC with \( r_d \) | 0.03                 | 1.00          | 1.00  | 2.2                  | 0.0           | 0.0   |
| CMASS NGC with bisp. | 0.0014               | 1.00          | 0.14  | 2.46                 | 0.0           | 1.5   |
| CMASS NGC with bisp. with \( r_d \) | 0.02                 | 0.60          | 0.58  | 2.3                  | 0.5           | 0.5   |

Figure 19. 2D posterior distributions of the cosmological parameters \( \ln(10^{10}A_s) \), \( \Omega_m \), and \( h \), and the EFT-parameters \( b_1 \) and \( c_2 \), obtained from fitting the CMASS NGC and SGC, and the LOWZ NGC power spectra up to \( k_{\text{max}} = 0.2h \text{ Mpc}^{-1} \). We can see the determination of the cosmological and the EFT parameters, as well as the consistency of NGC and SGC data sets.

... hard to obtain a measurement of \( A_s \) and \( \Omega_m \) from the clustering power spectrum alone, in order to detect the value of the neutrino masses from the analysis of this observable, it is therefore generically assumed that one has either to detect the step in the power spectrum.
Figure 20. 2D posterior distributions of the cosmological parameters $\ln(10^{10} A_s)$, $\Omega_m$, and $h$ obtained from fitting the CMASS and LOWZ NGC power spectra up to $k_{\text{max}} = 0.2 h \text{Mpc}^{-1}$ and $k_{\text{max}} = 0.18 h \text{Mpc}^{-1}$ respectively (left), and with the inclusion of the sound horizon prior (right), together with the same quantities from Planck2018 [84] and DES1Yr [88]. We can see the consistency with Planck2018.

Figure 21. 2D posterior distributions of the cosmological parameters $f(z = 0.55) \sigma_8(z = 0.55)$ and $h$ obtained from fitting the CMASS power spectrum sample (left), and of $S_8 = \sigma_8(z = 0) \left( \frac{\Omega_m}{0.3} \right)^{0.5}$ and $\Omega_m$ obtained from fitting the CMASS and the LOWZ NGC sample (right). We also plot the 2D posteriors for the same quantities from Planck2018 [84] and DES1Yr [88] and the 1D posterior for $f\sigma_8$ from BOSS CMASS sample [70]. We find that the results are consistent.
induced by neutrinos, or to measure $A_s$ (or equivalently $\Omega_m$) using some other observable (for example, CMB, lensing, cluster abundance, etc.). In our analysis, we have seen that we can measure $A_s$, $\Omega_m$ and $b_1$ directly using only the clustering power spectrum. This clearly adds additional constraining power, as we are going to show.

An analysis of the neutrino masses in BOSS data goes beyond the scope of the current paper, and we leave it to future work [73]. As a proof of principle and to show the potential power of using the EFTofLSS to detect neutrino masses, here we perform an analysis in flat $\Lambda$CDM with non-vanishing neutrino masses: we fix all the cosmological parameters to Planck2018, and fit the data by varying the mass of neutrinos, assuming normal mass hierarchy for definiteness.

At the level of the EFT model, we do not implement the rigorous derivation of the neutrino effects developed in [7], but rather simply perform the EFT computation described in section 2 using the linear power spectrum produced with such neutrino spectrum. Such an analysis has several shortcomings: the dark matter and neutrino clustering could be computed more accurately, we should allow the galaxy clustering to independently depend on the neutrinos overdensity, we should properly account for the $k$-dependence of the growth factors in the EFT-parameters (see for example [92]), one should impose Planck priors on the cosmological parameters, rather than fix them to the most likely ones from Planck2018. However, the effect we include should be the leading one [7]. We stress instead that it is acceptable to impose the Planck priors on all the cosmological parameters, as it is a fair assumption to study first the physics that we are sure is present in the ‘standard cosmological model’, such as flat $\Lambda$CDM with neutrino masses, and not to limit the power of such a ‘standard model’ analysis by looking at the same time for more exotic physics, such as dynamical dark energy.

With all these caveats in mind, we show on the left of figure 24 the posterior distribution for the sum of the neutrino masses and for $A_s$ from the analysis of the power spectrum of the CMASS sample of the BOSS DR12 data, for the analysis we mentioned earlier. We see that we have quite a tight constraint on the measurement of the sum of the neutrino masses,
Figure 23. 2D posterior distributions of the cosmological parameters $\ln(10^{10}A_s)$, $\Omega_m$, and $h$, and the galaxies biases $b_1$ and $c_2$, obtained from fitting the CMASS NGC power spectrum up to $k_{\text{max}} = 0.20 h \text{Mpc}^{-1}$, with and without the bispectrum monopole up to $k_{\text{max}} = 0.1 h \text{Mpc}^{-1}$, and with and without the Planck sound horizon prior. This plot highlights the contribution of the various data sets.

with error bars that are roughly comparable to the ones from Planck2018. Of course, given that our preferred value of $A_s$ is lower than Planck2018, we find a statistically-non-significant preference for non-vanishing masses once we fix $A_s$ to Planck2018. The origin of such a tight error bar can be traced back to the plot on the right of figure 24, where we plot the ratio of the linear power spectra with and without neutrino masses, together with the error bars of the monopole and the quadrupole. Given that $A_s$, $h$ and $\Omega_m$ are fixed, for given neutrino masses, the monopole can be fitted by lowering $b_1^2$. The amplitude of the quadrupole then drops by $b_1$, but once the drop is below the error allowed by the quadrupole, the chosen neutrino masses are excluded. We indeed can see that as the sum of the neutrino masses is above $\sim 0.26 eV$, the neutrino should begin to be excluded. This is indeed comparable to the error bars we have in the data.

In summary, though we find no evidence of a non-vanishing neutrino mass, and with the caveats that we described earlier in this section, we see a slight statistical preference for a mass above the minimal one imposed by neutrino oscillation experiments once we fix $A_s$ to the Planck2018 value. Most importantly, the error bars are quite tight: with some optimism, a detection of the neutrino masses, using LSS, might be around the corner.
\[ \sum m_{\nu} = 0.24 \pm 0.12 \pm 0.10 \]

Figure 24. Left: posterior distribution for the sum of the neutrinos in normal hierarchy, obtained from fitting the CMASS NGC, SGC and NGCxSGC power spectrum up to \( k_{\text{max}} = 0.2 h \text{Mpc}^{-1} \), and by fixing all the remaining cosmological parameters to Planck2018. Right: ratio of the linear power spectrum with and without neutrino masses, as a function of the sum of the neutrino masses in normal hierarchy. We see the step in the linear power spectrum. We also plot the monopole and quadrupole errors in red dashed and black dotted respectively. For fixed cosmological parameters, we expect a value of \( \sum_i m_{\nu_i} \) to be detectable once the drop in the power spectrum is below the error from the quadrupole, suggesting \( \sum_i m_{\nu_i} \sim 0.25\text{eV} \) should be detectable.

6 Conclusions

We have used the predictions from the Effective Field Theory of Large-Scale Structure (EFTofLSS) to analyze the data from the DR12 BOSS CMASS and LOWZ samples. In \( \Lambda \text{CDM} \), by fixing \( \Omega_b/\Omega_c \) and \( n_s \) to the best fit values of Planck2018, we have analyzed first the power spectrum up to \( k_{\text{max}} = 0.2 h \text{Mpc}^{-1} \) for the CMASS sample and up to \( k_{\text{max}} = 0.18 h \text{Mpc}^{-1} \) for the LOWZ sample, finding that we can measure \( A_s, \Omega_m \) and \( H_0 \) to about 13\%, 3.2\% and 3.2\% respectively. We have then added the Planck prior on the sound horizon, finding that this further improves the constraints to about 13\%, 2.9\% and 1.6\% respectively. We have also included the data on the bispectrum monopole, finding a limited further improvement on the measurement of \( A_s \). We have performed extensive checks with numerical simulations, that include several different HOD models, finding that the theoretical systematic error of the EFTofLSS is negligibly small. We have also performed a careful implementation of the window function and fiber collision effects, as well as studied the redshift selection effects, concluding that they do not induce any residual sizable systematic error. We have confirmed this conclusion using simulations that include these observational effects.

The analysis of the BOSS data using the EFTofLSS has required us to develop some interesting technical implementations. For example, we have developed a numerically-non-demanding way to apply the survey window function directly in Fourier space (as the standard procedure of applying it in real space was numerically challenging due to peculiarities of the
observables in the EFTofLSS);\textsuperscript{24} to reduce the number of parameters to be explored in the MCMC, we have marginalized analytically over some of the EFT parameters that appear linearly in the power spectrum. We stress that the technical aspects of the analysis that are discussed in section 3 could be used also for analyses not based on the EFTofLSS.

It appears to us that these measurements, that were enabled by the EFTofLSS, represent a very substantial qualitative and quantitative improvement of former analyses of redshift-clustering data. In fact, these parameters were never measured individually, and in such an accurate way. If confirmed, our results suggest that the Large-Scale Structure (LSS) of the Universe might contain much more exploitable cosmological information than previously believed. They therefore open up some research directions that we now briefly highlight.

On the side of the EFTofLSS and of the numerical simulations, we point out the following interesting research avenues that could be pursued:

- Our analysis is limited by the order at which we have performed our calculation. Here we have used the one-loop power spectrum and the tree-level bispectrum. It would be useful to go to the next perturbative order in both these observables, as well as to compute the trispectrum. This would allow us to additionally increase the accuracy of our predictions, as we mention briefly in the main text.

- By the use of simulations, the theory systematic error has been measured to be small up to the $k_{\text{max}}$ at which we have stopped the analysis, and we have therefore simply neglected it. It would be desirable to account for it in a more systematic way, either within the EFTofLSS itself (as we have highlighted in the main text), or by using more accurate and precise simulations.

- Notwithstanding the presence of many EFT parameters, our result show the great constraining power provided by the EFTofLSS. In this paper we have made no use of simulations to determine any prior about the EFT parameters. Simulations have been used only to estimate the theoretical systematic error. However, our statistical errors would be much smaller if we had stronger priors on the EFT parameters. We have checked that if we were to put a 15% prior on $b_1$, and more stringent priors on the EFT-parameters, our error bars would shrink. It would be very useful if simulations could provide reliable priors of this order. For example, one could consider if it is possible to extend techniques such as the ‘separate universe approach’ (see for example [94–98]) to include galaxies in a reliable way. Similarly, one could consider to use the information from physically-motivated phenomenological models, such as the very-well-known ‘halo model’ (see for example [99]), to provide stringent priors on the EFT parameters.

- We have highlighted how the measurement of the galaxy-bias coefficients (and more generally of the EFT parameters) that we can perform, could, at least in principle, allow us to determine the galaxy formation mechanism in the LSS. It would be interesting if a mapping between galaxy formation models and EFT parameters could to be performed.

- Our measurements are statistically compatible with the ones from Planck2018 [84], with a slight lower value for $A_s$. It would be interesting to see if this extremely mild tension can be ameliorated, for example, by including curvature or an equation of state for dark energy (i.e. $\Omega_K \neq 0$ or $w \neq -1$), or allowing the mass of the neutrinos is larger (see [73]).

\textsuperscript{24}See also [93] for a different numerical implementation.
Of course, we should not forget that these tensions are not statistically significant, but however go together with a general trend observed in analysis of other LSS data (see for example results from the DES collaboration [88]).

On the side of observations, we highlight a few interesting improvements and directions that could be pursued:

- In the EFTofLSS, several $n$-point functions, with $n > 2$, are predicted using approximately the same parameters as the power spectrum. We have not analyzed any of them beyond the bispectrum monopole, simply because we did not have such measurements at our disposal. Given the opportunity provided by the EFTofLSS, it would be desirable if measurements of multipoles of higher $n$-point functions, such as, for example, the bispectrum quadrupole or the trispectrum monopole, were available to us. The analysis of these data is rather straightforward from the EFTofLSS point of view. Similarly, we have not tried to optimize the analysis, for example studying the effect of the binning of the data or the use of better estimators. Indeed, the main objective of this paper has been to provide a first comparison of the EFTofLSS with data.

- Similarly, it would be interesting to repeat our analysis using the configuration-space correlation function. In fact, the relatively-limited reach in $k_{\text{max}}$ of the EFTofLSS at one-loop forces us to neglect the signal associated to the BAO oscillations at wavenumber higher than $k_{\text{max}}$. In fact, as already shown in [10], the BAO features can be accurately reproduced in their entirety by the IR-resummed EFTofLSS. A configuration-space analysis, or a modification of the current analysis in Fourier space, should allow for this additional information to be used.

- Clearly, the fact that the EFTofLSS allows us to break degeneracies and measure several cosmological parameters with such an accuracy requires that we re-asses the measurement of the observational systematics. This is particularly true given the slight disagreement with CMB experiments for the measurement of $A_s$, in particular when we add the bispectrum measurements. We stress that this is very important to do to fully trust our findings.

- Of course, it would be nice to apply our analysis to larger LSS surveys, such as DES, DESI, eBOSS, Euclid, LSST, etc., as well as to intensity mapping experiments. Similarly, it would be interesting to include the most common extensions to $\Lambda$CDM, such as neutrinos (for which we presented a preliminary analysis), non-Gaussianities, $N_{\text{eff}}$, curvature, etc., etc. Overall, we have performed this first analysis of the BOSS data using the EFTofLSS having in mind the current and next generation LSS surveys, and more general cosmological parameters.

We leave all of this, and more, to future work, some of which is already in progress [73].

Acknowledgments

Preliminary, but already at an advanced stage of completion, results of this work were presented at the following avenues (with link to the publicly-available slides): conference ‘the non-linear universe’, Slovenia, July 2018; conference ‘Analytic Methods in Cosmology’, Institute Poincare, Paris, Sept. 2018; conference ‘Modern Inflation’, Simons Center, NY, Sept.
2018; seminar, Harvard University, Nov. 2018; ‘South American Workshop on Cosmology in the LSST Era’, ICTP-SAIFR, Brazil, Dec. 2018; seminar, EPFL, Switzerland, Jan 2019; Stanford physics colloquium, Feb. 2019 [100]; conference ‘Accelerating universe in the dark’, Kyoto, March 2019 [101]; 2019 CCNU-USTC Junior Symposium, Wuhan, April 2019; conference ‘24th Rencontres Itzykson - Effective Field Theory in Cosmology, Gravitation and Particle Physics’, Paris, June 2019; conference ‘The nature of Dark Matter and Large Scale Structure’, Cyprus, June 2019, workshop ‘Large-Scale Structure Formation’, Munich, July 2019, conference ‘EPS-HEP219’, July 2019 [102].

While this work was near completion, ref. [103] appeared, which has some overlap with a few parts of this paper.

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The Boltzmann code CLASS was used to compute the matter linear power spectrum [104]. Some plots were done using the Chainconsumer package [105]. Part of the analysis was performed with the Mercury cluster at the University of Science and Technol-

\textsuperscript{25}See EFTofLSS repository.
ogy of China, and PZ thanks Jie Jiang and Tian Qiu for support. Part of the analysis was also performed on the Sherlock cluster at the Stanford University and we thank the support team.

A Galaxy kernels

We give the explicit expressions for the galaxy kernels appearing in the one-loop power spectrum and refer to [50] for details about their derivation. We choose to work in the basis of descendants (this is the first complete set of bias coefficient for LSS, established in [9, 48] and with some typos corrected in [49]; see [9, 48] for connection to former bases of bias coefficients, as for example [52]). For the one-loop power spectrum and tree-level bispectrum, all kernels can be described with 4 bias parameters $b_i$.

The first and second order galaxy density kernel are:

$$K_1 = b_1,$$  
(A.1)

$$K_2(q_1, q_2) = b_1 \frac{q_1 \cdot q_2}{q_1^4} + b_2 \left( F_2(q_1, q_2) - \frac{q_1 \cdot q_2}{q_1^4} \right) + b_4 + \text{perm.}.$$  
(A.2)

The galaxy velocity kernels $G_n$ are simply the standard perturbation theory ones since the galaxy velocity field follows the dark matter velocity field, up to higher-derivative terms which are degenerate with other counterterms that appear in the renormalization of the redshift space expression (see e.g. [15] for the expressions of $F_n$ and $G_n$).

The third-order galaxy density kernel has a much more involved expression. However, for the one-loop calculation, degeneracies appear in the one-loop diagram obtained from $\langle \delta^{(3)}(q) \delta^{(1)}(q) \rangle$, when UV-divergences are removed and the integral over the angular coordinates is performed, leading to the following simple expression:

$$K_3(k, q) = -\frac{b_1}{504 k^3 q^2} \left( -38 k q + 48 k^3 q^3 - 18 k^3 q^5 + 9(k^2 - q^2)^3 \log \left[ \frac{k - q}{k + q} \right] \right)$$

$$+ \frac{b_3}{756 k^3 q^3} \left( 2kq(k^2 + q^2)(3k^4 - 14k^2 q^2 + 3q^4) + 3(k^2 - q^2)^4 \log \left[ \frac{k - q}{k + q} \right] \right).$$

(A.3)

The integrals are computed using the Cuhre routine of the CUBA library [106]. This cubature algorithm happens to be very competitive in this low-dimension problem compared to Monte-Carlo algorithm such as Vegas. As the integral over one angular coordinate can be calculated analytically, the integrals are only two-dimensional. 

B Evaluation of the loop integrals

We present two strategies to evaluate the loop integrals appearing in the galaxy power spectrum (2.1): a simple numerical one and an analytical one relying on the decomposition of the linear power spectrum in log $k$-space.

For the numerical evaluation we adopt the IR-safe and UV-safe integrand version of the 22-diagram + 13-diagram [34], such that most of the (unphysical) divergences that are hard to handle numerically are absent (see appendix D of [50] for an explicit expression). To gain additional computational speed, we calculate only their differences to a precomputed cosmology [33]. The loop integrals are of the form:

$$P_n(k) = \int \frac{d^3 q}{(2\pi)^3} P_n,\text{integrand}(k, q),$$

(B.1)

The integrals are computed using the Cuhre routine of the CUBA library [106]. This cubature algorithm happens to be very competitive in this low-dimension problem compared to Monte-Carlo algorithm such as Vegas. As the integral over one angular coordinate can be calculated analytically, the integrals are only two-dimensional.
which can be calculated with respect to a reference cosmology:

\[ P_{\text{target}}^{n}(k) = P_{\text{ref}}^{n}(k) + \Delta P_{n}(k), \]  

where “target” and “ref” refer to a cosmology being evaluated and a fiducial (reference) cosmology, respectively. The difference \( \Delta P_{n}(k) \) is the integral of the difference between the reference and the target integrands:

\[
\Delta P_{n}(k) = \int \frac{d^3q}{(2\pi)^3} \left[ P_{\text{target}}^{n,\text{integrand}}(k, q) - \left( \frac{A_{\text{target}}}{A_{\text{ref}}} \right)^2 P_{\text{ref}}^{n,\text{integrand}}(k, q) \right],
\]

where the ratio of the scalar amplitudes \( \left( \frac{A_{\text{target}}}{A_{\text{ref}}} \right)^2 \) is inserted to rescale the amplitude of the reference integrand to the target one such that \( \Delta P_{n}(k) \) is as small as possible.

Once the loop integrals for a reference cosmology \( P_{\text{ref}}^{n}(k) \) are precomputed to high-enough precision, we are free to evaluate only the difference \( \Delta P_{n}(k) \) instead of the full integral, to a relatively low accuracy.\(^{27}\) This simple manipulation speeds up the computation by at least a factor 2 without precision loss.

As we mentioned, following the procedure of [27, 34, 50], we make our integrals UV-safe, by removing the UV-limit of the integrand. This contribution indeed, after integration, is degenerate with the counterterms. This has the effect to make the integrals smaller, and so easier to evaluate. It also diminishes the size of the ‘bare’ part of the counterterms, so that they are, effectively, directly the renormalized ones, and therefore, expected to be of order one.

One can also choose to evaluate the loop integrals on a set of complex power-law cosmologies produced by decomposing the linear power spectrum using an FFTLog algorithm [61]. This procedure has several advantages. First, the loop integrals in eq. (2.1) are analytic for power-law cosmologies. In fact, they are formally equivalent to massless scalar propagator loop integrals in quantum field theory. As such, all UV divergences are automatically removed from the loop thanks to dimensional regularization and the counterterms needed to absorb these divergences are assured to be order \( O(1) \). The second advantage is practical. The loop integrals can be evaluated once and for all as they do not depend on the cosmology. Evaluating the loop integrals in eq. (2.1) at a new cosmology reduces to decomposing the linear power spectrum in a sum of complex power-law cosmologies and a few matrix multiplications.\(^{28}\)

\(^{27}\) It can be shown that the allowed relative error \( \epsilon_{\Delta} \) on the evaluation of \( \Delta P_{n}(k) \) is inversely proportional to its size:

\[
\epsilon_{\Delta} \approx \frac{|P_{\text{target}}^{n}(k)|}{|\Delta P_{n}(k)|} \epsilon_{\text{target}},
\]

where \( \epsilon_{\text{target}} \) is the required precision on the evaluation of \( P_{n}^{\text{target}}(k) \). A close inspection of the integrals involved in the 1-loop expression shows that the internal momenta that will contribute the most are of order \( k \). The kernels being independent of the cosmology, we can estimate the integrals as follows with the corresponding linear power spectra evaluated at \( k \):

\[
\left| \frac{\Delta P_{n}(k)}{P_{n}^{\text{target}}(k)} \right| = \left| 1 - \left( \frac{A_{\text{target}}}{A_{\text{ref}}} \right)^2 \frac{\int d^3q P_{\text{ref}}^{n,\text{integrand}}(k, q)}{\int d^3q P_{\text{target}}^{n,\text{integrand}}(k, q)} \right| \approx \left| 1 - \left( \frac{A_{\text{target}}}{A_{\text{ref}}} \right)^2 \left( \frac{P_{\text{ref}}^{n}(k)}{P_{\text{target}}^{n}(k)} \right)^2 \right|. \]

This estimate has been shown to be good enough for the 2-loop matter power spectrum [33]. We thus confidently use this estimate in our 1-loop calculation and choose \( \epsilon_{\Delta} \approx 5 \epsilon_{\text{target}} \).

\(^{28}\) We acknowledge the use of the C++ library Eigen [107].
For \( N \) sampling points, the linear power spectrum can be approximated using FFT in log-space by:
\[
P_{11}(k) = \sum_{-N/2}^{N/2} c_m k^{\nu + i\eta_m}, \tag{B.6}
\]
where
\[
c_m = \frac{1}{N} \sum_{l=0}^{N-1} P_{11}(k) k_l^{\nu} e^{-2\pi iml/N} \quad \eta_m = \frac{2\pi m}{\log(k_{\text{max}}/k_{\text{min}})}. \tag{B.7}
\]
For \(-3 < \nu < -1.5\) the IR divergences in the loop cancel. Expanding the kernels in integer powers of \( k^2, q^2 \) and \(|k-q|^2\), the loop integrals can be performed analytically on the power-law cosmologies \( \eta_m \). The 22-loop and 13-loop diagrams become simple matrix multiplications:
\[
P_{n,22}(k) = k^3 \sum_{m_1,m_2} c_{m_1} k^{\nu + i\eta_{m_1}} \cdot M_{n,22} \left( -\frac{\nu + i\eta_{m_1}}{2}, -\frac{\nu + i\eta_{m_2}}{2} \right) \cdot c_{m_2} k^{\nu + i\eta_{m_1}}, \tag{B.8}
\]
\[
P_{n,13}(k) = k^3 P_{11}(k) \sum_{m_1} c_{m_1} k^{\nu + i\eta_{m_1}} \cdot M_{n,13} \left( -\frac{\nu + i\eta_{m_1}}{2} \right), \tag{B.9}
\]
where the matrix \( M_{n,22}(\nu_1,\nu_2) \) and the vector \( M_{n,13}(\nu_1) \) for each term \( P_n \) (which is either of the 22 or the 13 form) in the decomposition of the kernels as induced by (2.6) can be obtained by decomposing the kernels in power laws of momenta, and performing the resulting momentum integral with the following expression:
\[
\frac{1}{2\pi^3} \int d^3q \frac{1}{q^{2\nu_1} |k-q|^{2\nu_2}} = \frac{1}{8\pi^{3/2}} k^{3-2\nu_1} \frac{\Gamma(\frac{3}{2} - \nu_1) \Gamma(\frac{3}{2} - \nu_2) \Gamma(\nu_1 - \frac{3}{2})}{\Gamma(\nu_1) \Gamma(\nu_2) \Gamma(3 - \nu_1)}. \tag{B.10}
\]
In our analysis, we have implemented both procedures for the evaluation of the integrals, finding identical results for all practical purposes.

### C IR-resummation scheme

The parameters parametrizing the tidal forces of long-wavelength modes \( k \) on the short-wavelength modes and the effect of short-wavelength displacements are, respectively:
\[
\epsilon_{\delta<} = \int_k d^3q P_{11}(q), \quad \epsilon_{s>} = k^2 \int_k d^3q \frac{P_{11}(q)}{q^2}. \tag{C.1}
\]
Both of these scale as powers of \( k/k_{\text{NL}} \) and are thus small: the perturbative expansion is valid and the convergence is assured. However, long-wavelength displacements, which are parametrized by:
\[
\epsilon_{s<} = k^2 \int_k d^3q \frac{P_{11}(q)}{q^2}. \tag{C.2}
\]
are generically of order one. They can not be treated perturbatively and therefore need to be resummed. We here follow the parametrically-controlled procedure developed in [29].
The IR-resummation in redshift space given in terms of multipole moments reads [53] (see also [10]):

\[ P_n^e(k)|_N = \sum_{j=0}^{N} \sum_{\ell} \frac{d\ell}{2\pi^2} M_{||N-j}^{\ell e}(k, k') P_n^{e}(k')_{j}, \quad (C.3) \]

\[ M_{||N-j}^{\ell e}(k, k') = \int dq j_{\ell}'(k'q) Q_{\ell j}^{N-j}(k, q), \quad (C.4) \]

\[ Q_{\ell j}^{N-j}(k, q) = i\ell q^2 \frac{2\ell + 1}{2} \int_{-1}^{1} d\mu_k \]

\[ \times \int d^2\hat{q} e^{-i\hat{q} \cdot \hat{k}} T_{0,r}(k, q) \times T_{0,r}^{-1}(||N-j)(k, q) P_k(\mu_k) P_\ell(\mu_\ell), \quad (C.5) \]

where we denoted by \(|N\) a quantity that is expanded up to order \(N\) in both \(\epsilon_{\delta<}, \epsilon_{s>}\) and \(\epsilon_{s<}\), while with \(|N|\) a quantity is expanded \emph{up to} the order \(N\) (loop level) in \(\epsilon_{\delta<}, \epsilon_{s>}\) and \(\emph{all orders in} \ \epsilon_{s<}.\) \(T_{0,r}(k, q)\) is given by

\[ T_{0,r}(k, q) = \exp \left\{ -\frac{k^2}{2} \left[ X_1(q) \left( 1 + 2j_\mu_\ell k + j_\mu_\ell^2 k^2 \right) + Y_1(q) \left( (\hat{k} \cdot \hat{q})^2 + 2j_\mu_\ell \mu_\ell k \hat{k} \cdot \hat{q} + j_\mu_\ell^2 \mu_\ell^2 \right) \right] \right\}, \quad (C.6) \]

where

\[ X_1(q) = \frac{1}{2\pi} \int_0^{\infty} dk \exp \left( -\frac{k^2}{\Lambda_{IR}^2} \right) P_{11}(k) \left( \frac{2}{3} - 2\frac{j_1(kq)}{kq} \right), \quad (C.7) \]

\[ Y_1(q) = \frac{1}{2\pi} \int_0^{\infty} dk \exp \left( -\frac{k^2}{\Lambda_{IR}^2} \right) P_{11}(k) \left( -2j_0(kq) + 6\frac{j_1(kq)}{kq} \right), \quad (C.8) \]

\(j_i\) are the spherical Bessel functions of type \(i\) and \(\mu_k = \hat{k} \cdot \hat{z}\). \(\Lambda_{IR}\) is the scale to which we resum the linear IR modes, taken to be \(\Lambda_{IR} = 0.066 \ h\ Mpc^{-1}\) in this analysis.

The \(Q\) function can be approximated for efficient numerical evaluation [53]. The idea is to expand the exponential in \(X_1(q)(1 + 2j_\mu_\ell k + j_\mu_\ell^2 k^2)\) and \(Y_1(q)(\hat{k} \cdot \hat{q})^2 + 2j_\mu_\ell \mu_\ell k \hat{k} \cdot \hat{q} + j_\mu_\ell^2 \mu_\ell^2\) in powers of \(k^3X_1(\mu_k^2 - 1/3)\) and \(k^2Y_1\). Up to order 3 and 1, respectively, the convergence has been checked to be good enough for a 1-loop calculation. Under this approximation, the integrals in eq. (C.5) can be done exactly, hence allowing a fast numerical computation.

We finally perform another numerical improvement. As it is well known (see for example the introduction of [10] and [39, 42]), the IR-resummation is trustable, in the sense that it resums terms that are larger than the rest, only for as long as it acts only on the oscillating part of the power spectrum. For this reason, we split the power spectrum at linear level and at loop level in a smooth and an oscillating part. At linear level, the smooth part is defined with the Eisenstein and Hu no-wiggle power spectrum [108], while the one-loop smooth part is defined by performing the loop integrals with the Eisenstein and Hu no-wiggle power spectrum. The oscillating part is what is left. Then, we act with the resummation matrices only on the oscillating part of the linear and loop power spectrum. This approach has the advantage that numerical inaccuracy in the resummation matrices and in the final resummation
integral are relegated to the oscillating part, which is small. However, it has some disadvantages, such as, for example, the fact that whatever is oscillating and is accidentally included in the smooth part, as the split in oscillating and non-oscillating part of the power spectrum is somewhat arbitrary, is not resummed (see for example [41] for a accurate discussion about this). We have checked that this numerically optimized procedure gives results that are practically indistinguishable from the original one, for the error bars we are dealing with now.

The bispectrum should also in principle be resummed. However, given that we use the tree level expression and we stop at rather long wavenumbers, where the errors from cosmic variance are very large, we assume the effect of the IR-resummation can be neglected for the bispectrum.

D Technical aspects of bispectrum likelihood

We first re-derive the formula analogous to section 3 with the inclusion of the bispectrum in the likelihood and then briefly discuss the application of the window function. The full covariance matrix can be decomposed into four blocks $C^{-1}_{\text{PP}}, C^{-1}_{\text{PB}} \equiv C^{-1}_{\text{PP}}$ and $C^{-1}_{\text{BB}}$, where P and B stand for power spectrum and bispectrum, respectively. The full likelihood reads:

\[
\mathcal{L}_{\text{FULL}}(d|\{\tilde{\Omega}, \tilde{b}\}) = \mathcal{L}_{\text{PP}}(d|\{\tilde{\Omega}, \tilde{b}\}) + \mathcal{L}_{\text{PB}}(d|\{\tilde{\Omega}, \tilde{b}\}) + \mathcal{L}_{\text{BB}}(d|\{\tilde{\Omega}, \tilde{b}\})
\]

\[
= \text{Exp} \left[ -\frac{1}{2} \left( P^{(\text{EFT})}_{\ell}(k,\{\tilde{\Omega}, \tilde{b}\}) - P^{(d)}_{\ell}(k) \cdot C^{-1}_{\text{PP}}(k,k')_{\ell,\ell'} \cdot (P^{(\text{EFT})}_{\ell}(k',\{\tilde{\Omega}, \tilde{b}\}) - P^{(d)}_{\ell}(k')) \right) \right] \\
\times \text{Exp} \left[ -(P^{(\text{EFT})}_{\ell}(k,\{\tilde{\Omega}, \tilde{b}\}) - P^{(d)}_{\ell}(k)) \cdot C^{-1}_{\text{PB}}(k,k')_{\ell} \cdot (B^{(\text{EFT})}_{0}(k',\{\tilde{\Omega}, \tilde{b}\}) - B^{(d)}_{0}(k')) \right] \\
\times \text{Exp} \left[ -\frac{1}{2} (B^{(\text{EFT})}(k,\{\tilde{\Omega}, \tilde{b}\})) - B^{(d)}_{0}(k)) \cdot C^{-1}_{\text{BB}}(k,k') \cdot (B^{(\text{EFT})}(k',\{\tilde{\Omega}, \tilde{b}\}) - B^{(d)}_{0}(k')) \right],
\]

where on the second line we have the likelihood $\mathcal{L}_{\text{PP}}(d|\{\tilde{\Omega}, \tilde{b}\})$ of the power spectrum only, eq. (3.2), the third line represents the cross-likelihood $\mathcal{L}_{\text{PB}}(d|\{\tilde{\Omega}, \tilde{b}\})$ of the power spectrum and the bispectrum, and the fourth line is the likelihood $\mathcal{L}_{\text{BB}}(d|\{\tilde{\Omega}, \tilde{b}\})$ of the bispectrum only. The formulas with the explicit appearance of the window function and with the partial marginalization for $\mathcal{L}_{\text{PP}}(d|\{\tilde{\Omega}, \tilde{b}\})$ are given in section 3 and we do not repeat them here. In this analysis, following [63], we apply the window function to the bispectrum in a rather approximate way, where we apply the window function to each of the two power spectra that appear in the tree-level expression in eq. (2.5). Such an approximate treatment should be sufficient given the large error bars of the data at large wavenumbers. Thus for $\mathcal{L}_{\text{BB}}(d|\{\tilde{\Omega}, \tilde{b}\})$ the covariance matrix is unchanged while for the cross-likelihood, we have:

\[
\mathcal{L}_{\text{PB}}(d|\{\tilde{\Omega}, \tilde{b}\}) = \text{Exp} \left[ -P^{(\text{EFT})}_{\ell}(k,\{\tilde{\Omega}, \tilde{b}\}) \cdot C^{-1}_{\text{PB,W}}(k,k')_{\ell} \cdot (B^{(\text{EFT})}_{0}(k',\{\tilde{\Omega}, \tilde{b}\}) - B^{(d)}_{0}(k')) \right. \\
\left. + P^{(d)}_{\ell}(k) \cdot C^{-1}_{\text{PB}}(k,k')_{\ell} \cdot (B^{(\text{EFT})}_{0}(k',\{\tilde{\Omega}, \tilde{b}\}) - B^{(d)}_{0}(k')) \right],
\]

where

\[
C^{-1}_{\text{PB,W}}(k_1,k_2)_{\ell} = W(k_1,k')_{\ell,\ell'} \cdot C^{-1}_{\text{PB}}(k',k_2)_{\ell'}.\]

Similarly to the power spectrum, we can perform a partial marginalization over some bias parameters. We now have $\tilde{b} = \{\tilde{b}_{\text{NG}}, \tilde{b}_{\text{G}}\}$, with $\tilde{b}_{\text{G}} = \{b_3, c_{\ell,1}, c_{\ell,2}, c_{\ell,3}, c_{\ell,4}\}$ and $\tilde{b}_{\text{NG}} = \{b_1, b_2, b_4, c_{\ell,1}\}$. Writing, in analogy to what we did for the case of the power spectrum
only: 

\[ P_{\ell}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) = \sum_i b_{G,i} P_{\ell,\text{lin},i}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) + P_{\ell,\text{const}}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}), \quad (D.4) \]

\[ B_0^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}\}) = \sum_i b_{G,i} B_{0,\text{lin},i}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) + B_{0,\text{const}}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}), \quad (D.5) \]

we get:

\[ \mathcal{L}_{\text{PB}}(d|\{\bar{\Omega}, \bar{b}_G, \bar{b}_N\}) = \exp \left[ -\frac{1}{2} \sum_{ij} b_{G,i} G_{2,ij}(\{\bar{\Omega}, \bar{b}_G\}) b_{G,j} + \sum_i b_{G,i} G_{1,i}(\{\bar{\Omega}, \bar{b}_G\}) + G_0(\{\bar{\Omega}, \bar{b}_G\}) \right], \quad (D.6) \]

where

\[ G_{2,ij} = 2P_{\ell,\text{lin},i}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) \cdot C_{\text{PB,W}}^{-1}(k, k') \cdot \sum_{k} B_{0,\text{lin},j}^{(\text{EFT})}(k', \{\bar{\Omega}, \bar{b}_G\}), \quad (D.7) \]

\[ G_{1,i} = -P_{\ell,\text{const}}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) \cdot C_{\text{PB,W}}^{-1}(k, k') \cdot \sum_{k} B_{0,\text{const}}^{(\text{EFT})}(k', \{\bar{\Omega}, \bar{b}_G\}) - B_0^{(d)}(k'), \quad (D.8) \]

\[ G_0 = -P_{\ell,\text{const}}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) \cdot C_{\text{PB,W}}^{-1}(k, k') \cdot \sum_{k} B_{0,\text{const}}^{(\text{EFT})}(k', \{\bar{\Omega}, \bar{b}_G\}) - B_0^{(d)}(k'), \quad (D.9) \]

and

\[ \mathcal{L}_{\text{BB}}(d|\{\bar{\Omega}, \bar{b}_G, \bar{b}_N\}) = \exp \left[ -\frac{1}{2} \sum_{ij} b_{G,i} H_{2,ij}(\{\bar{\Omega}, \bar{b}_G\}) b_{G,j} + \sum_i b_{G,i} H_{1,i}(\{\bar{\Omega}, \bar{b}_G\}) + H_0(\{\bar{\Omega}, \bar{b}_G\}) \right], \quad (D.10) \]

where

\[ H_{2,ij}(\{\bar{\Omega}, \bar{b}_G\}) = B_{0,\text{lin},i}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) \cdot C_{\text{BB}}^{-1}(k, k') \cdot \sum_{k} B_{0,\text{lin},j}^{(\text{EFT})}(k', \{\bar{\Omega}, \bar{b}_G\}), \quad (D.11) \]

\[ H_{1,i}(\{\bar{\Omega}, \bar{b}_G\}) = -B_{0,\text{lin},i}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) \cdot C_{\text{BB}}^{-1}(k, k') \cdot (B_{0,\text{const}}^{(\text{EFT})}(k', \{\bar{\Omega}, \bar{b}_G\}) - B_0^{(d)}(k')), \quad (D.12) \]

\[ H_0(\{\bar{\Omega}, \bar{b}_G\}) = -\frac{1}{2} (B_{0,\text{const}}^{(\text{EFT})}(k, \{\bar{\Omega}, \bar{b}_G\}) - B_0^{(d)}(k)) \cdot C_{\text{BB}}^{-1}(k, k') \cdot (B_{0,\text{const}}^{(\text{EFT})}(k', \{\bar{\Omega}, \bar{b}_G\}) - B_0^{(d)}(k')), \quad (D.13) \]

Defining \( T_a = F_a + G_a + H_a \), the marginalized likelihood of the power spectrum plus bispectrum is given by:

\[ \mathcal{L}_{\text{FULL}}(d|\{\bar{\Omega}, \bar{b}_G\}) = \int dG \mathcal{L}_{\text{FULL}}(d|\{\bar{\Omega}, \bar{b}_G, \bar{b}_N\}) \]

\[ = \exp \left[ \frac{1}{2} T_{1,i}(\{\bar{\Omega}, \bar{b}_G\}) \cdot T_2(\{\bar{\Omega}, \bar{b}_G\})^{-1} \cdot T_{1,j}(\{\bar{\Omega}, \bar{b}_G\}) + T_6(\{\bar{\Omega}, \bar{b}_G\}) - \frac{1}{2} \log \det \left( T_2(\{\bar{\Omega}, \bar{b}_G\}) \right) \right], \quad (D.14) \]

where we dropped an irrelevant constant.
We finish by commenting on the application of the window function to the bispectrum monopole. As mentioned, unfortunately, we do not have at our disposal an exact formula for the application of the window function to the bispectrum, contrary to what we have in eq. (3.3) for the power spectrum. Therefore, we use an approximate procedure that was used in [26] and that we described above. It is unclear how accurate this procedure is, and, in order to limit the systematic error induced by this approximation, we start analyzing the data for the bispectrum at a minimum wavenumber \( k_{\text{min}} = 0.04h \text{Mpc}^{-1} \). It would be preferable to have a more accurate way to implement the window function. We leave this to future work.

E  Sound horizon at decoupling

The sound horizon \( r_d \) at decoupling at epoch \( z_d \) reads:

\[
 r_d = \int_{z_d}^{\infty} \frac{c_s(z)}{H(z)} \, dz, \tag{E.1}
\]

where \( c_s \) is the sound speed in the primordial photon-baryon plasma given by:

\[
 c_s^2(z) = \frac{c^2}{3} \left[ 1 + \frac{3 \rho_b(z)}{4 \rho_\gamma(z)} \right]^{-1}, \tag{E.2}
\]

where \( \rho_b \) and \( \rho_\gamma \) are the energy density of baryons and radiation respectively. The integral in (E.1) can be evaluated exactly as:

\[
 r_d = \frac{2c}{H_0 \sqrt{3R \Omega_m}} \log \left[ \frac{\sqrt{1 + z_d + R} + \sqrt{(1 + z_d)R \Omega_{\text{rad}} + R}}{\sqrt{1 + z_d} \left( 1 + \sqrt{R \Omega_{\text{rad}}} \right)} \right], \tag{E.3}
\]

where \( R = \frac{3 \Omega_b}{4 \Omega_\gamma} \), the normalized radiation density today is \( \Omega_{\text{rad}} = \left[ 1 + N_{\text{ur}} \left( \frac{4}{11} \right)^{4/3} \right] \Omega_\gamma h^2 \) with \( N_{\text{ur}} \) the number of ultra-relativistic species and \( \Omega_\gamma h^2 = 2.47282 \cdot 10^{-5} \), \( H_0 \) is, as usual, the Hubble constant at present time, and \( c \) is the speed of light.

When we use the Planck prior on \( r_d \), we use the following value from [84] Planck2018 TT,TE,EE+lowP+lensing (table 2) to analyze the CMASS sample:

\[
 r_d = 147.09 \pm 0.26 \text{ Mpc}, \quad z_d = 1059.94 \pm 0.30, \tag{E.4}
\]

while for the simulations we use instead \( r_d \) computed at their fiducial cosmology, table 1, and use the same error. In our analysis, we take \( z_d \) at its fixed Planck fiducial value considering that for the cosmologies under probe, \( r_d(z_d + \sigma(z_d)) - r_d(z_d - \sigma(z_d)) < 0.2 \sigma(r_d) \), and so the effect is marginal.

F  Correlation matrices

In tables 6 and 7 we provide the correlation matrices for the parameters entering the likelihood obtained by evaluating the non-marginalized posterior of the power spectrum. We present results at \( k_{\text{max}} = 0.2h \text{Mpc}^{-1} \) for Challenge Box D, and for the CMASS data. In figure 25 and 26 we plot the 2D posterior distributions for challenge D and for CMASS NGC and SGC.
Joint-EFT analysis of CMASS NGC \times SGC

For our main results, when analyzing the combination CMASS NGC and SGC, we conservatively kept two independent set of EFT parameters between the two skycuts in the fit. This is the same choice as done by the BOSS collaboration to account for different selection effects [25]. Here, we provide 1D marginalized posterior distributions of the cosmological parameters in figure 27 if we choose to fit with one set of EFT parameters but the shot noises for the whole CMASS sample. We find that the error bars on the cosmological parameters are roughly improved by 10% at \( k_{\text{max}} = 0.23 h \, \text{Mpc}^{-1} \) with respect to the former case, and the central values are shifting at most by about half a \( \sigma_{\text{stat}} \). The minimum \( \chi^2/\text{d.o.f.} \) is 1.05. From figure 26 we do not find evidence that the two samples NGC and SGC detect different EFT parameters, but we cautiously opt for the most conservative choice of two independent sets of parameters for our main results.

| Parameter | \( \ln(10^{10} A_s) \) | \( \Omega_m \) | \( h \) | \( b_1 \) | \( c_2 \) | \( b_3 \) | \( c_{\text{ct}} \) | \( c_{r,1} \) | \( c_{\epsilon,\text{quad}} \) |
|-----------|----------------|-------------|----------|--------|--------|--------|--------|--------|--------|
| \( \ln(10^{10} A_s) \) | 1.00 | -0.10 | -0.10 | -0.89 | -0.80 | -0.01 | 0.09 | -0.57 | 0.19 |
| \( \Omega_m \) | -0.10 | 1.00 | 0.34 | -0.14 | -0.02 | 0.40 | -0.10 | 0.05 | -0.25 |
| \( h \) | -0.10 | 0.34 | 1.00 | -0.29 | -0.15 | 0.50 | -0.06 | -0.02 | -0.28 |
| \( b_1 \) | -0.89 | -0.14 | -0.29 | 1.00 | 0.82 | -0.13 | -0.06 | 0.53 | -0.06 |
| \( c_2 \) | -0.80 | -0.02 | -0.15 | 0.82 | 1.00 | -0.14 | -0.33 | 0.16 | 0.05 |
| \( b_3 \) | -0.01 | 0.40 | 0.50 | -0.13 | -0.14 | 1.00 | 0.03 | 0.07 | -0.33 |
| \( c_{\text{ct}} \) | 0.09 | -0.10 | -0.06 | -0.06 | -0.33 | 0.03 | 1.00 | 0.25 | 0.15 |
| \( c_{r,1} \) | -0.57 | 0.05 | -0.02 | 0.53 | 0.16 | 0.07 | 0.25 | 1.00 | -0.69 |
| \( c_{\epsilon,\text{quad}} \) | 0.19 | -0.25 | -0.28 | -0.06 | 0.05 | -0.33 | 0.15 | -0.69 | 1.00 |

Table 6. Parameter Correlations Challenge D, \( k_{\text{max}} = 0.20 \).

| Parameter | \( \ln(10^{10} A_s) \) | \( \Omega_m \) | \( h \) | \( b_1 \) | \( c_2 \) | \( b_3 \) | \( c_{\text{ct}} \) | \( c_{r,1} \) | \( c_{\epsilon,1/\bar{n}_g} \) | \( c_{\epsilon,\text{quad}} \) |
|-----------|----------------|-------------|----------|--------|--------|--------|--------|--------|-------------|--------|
| \( \ln(10^{10} A_s) \) | 1.00 | 0.08 | -0.23 | -0.90 | -0.46 | -0.18 | 0.12 | -0.48 | 0.10 | 0.30 |
| \( \Omega_m \) | 0.08 | 1.00 | -0.10 | -0.13 | -0.18 | 0.28 | 0.05 | 0.04 | -0.05 | -0.01 |
| \( h \) | -0.23 | -0.10 | 1.00 | -0.16 | -0.13 | 0.40 | -0.07 | -0.11 | -0.10 | 0.04 |
| \( b_1 \) | -0.90 | -0.13 | -0.16 | 1.00 | 0.57 | 0.04 | -0.10 | 0.49 | -0.10 | -0.31 |
| \( c_2 \) | -0.46 | -0.18 | -0.13 | 0.57 | 1.00 | 0.14 | -0.56 | -0.21 | -0.34 | -0.11 |
| \( b_3 \) | -0.18 | 0.28 | 0.40 | 0.04 | 0.14 | 1.00 | 0.02 | -0.10 | -0.01 | -0.06 |
| \( c_{\text{ct}} \) | 0.12 | 0.05 | -0.07 | -0.10 | -0.56 | 0.02 | 1.00 | 0.21 | -0.08 | 0.12 |
| \( c_{r,1} \) | -0.48 | 0.04 | -0.11 | 0.49 | -0.21 | -0.10 | 0.21 | 1.00 | 0.30 | -0.67 |
| \( c_{\epsilon,1/\bar{n}_g} \) | 0.10 | -0.05 | -0.10 | -0.10 | -0.34 | -0.01 | -0.08 | 0.30 | 1.00 | 0.02 |
| \( c_{\epsilon,\text{quad}} \) | 0.30 | -0.01 | 0.04 | -0.31 | -0.11 | -0.06 | 0.12 | -0.67 | 0.02 | 1.00 |

Table 7. Parameter Correlations CMASS NGC, \( k_{\text{max}} = 0.2h \, \text{Mpc}^{-1} \).
Figure 25. 2D posterior distributions of all cosmological and EFT parameters, obtained from fitting the power spectrum up to $k_{\text{max}} = 0.20h\,\text{Mpc}^{-1}$ of the challenge box D with the non-marginalized likelihood.
Figure 26. 2D posterior distributions of all cosmological and EFT parameters, obtained from fitting the CMASS NGC and SGC power spectra independently up to $k_{\text{max}} = 0.23 h \text{Mpc}^{-1}$ with the non-marginalized likelihood.
Figure 27. Marginalized posterior distributions of the cosmological parameters obtained from fitting the power spectrum of CMASS NGC and SGC power spectra with a same set of parameters but for the shot noises for the two skycuts.

References

[1] D. Baumann, A. Nicolis, L. Senatore and M. Zaldarriaga, Cosmological non-linearities as an effective fluid, JCAP 07 (2012) 051 [arXiv:1004.2488] [SPIRE].

[2] J.J.M. Carrasco, M.P. Hertzberg and L. Senatore, The effective field theory of cosmological large scale structures, JHEP 09 (2012) 082 [arXiv:1206.2926] [SPIRE].

[3] R.A. Porto, L. Senatore and M. Zaldarriaga, The Lagrangian-space effective field theory of large scale structures, JCAP 05 (2014) 022 [arXiv:1311.2168] [SPIRE].

[4] J.J.M. Carrasco, S. Foreman, D. Green and L. Senatore, The 2-loop matter power spectrum and the IR-safe integrand, JCAP 07 (2014) 056 [arXiv:1304.4946] [SPIRE].

[5] L. Mercolli and E. Pajer, On the velocity in the effective field theory of large scale structures, JCAP 03 (2014) 006 [arXiv:1307.3220] [SPIRE].

[6] M. Lewandowski, A. Perko and L. Senatore, Analytic prediction of baryonic effects from the EFT of large scale structures, JCAP 05 (2015) 019 [arXiv:1412.5049] [SPIRE].

[7] L. Senatore and M. Zaldarriaga, The effective field theory of large-scale structure in the presence of massive neutrinos, arXiv:1707.04698 [SPIRE].

[8] M. Lewandowski, A. Maleknejad and L. Senatore, An effective description of dark matter and dark energy in the mildly non-linear regime, JCAP 05 (2017) 038 [arXiv:1611.07966] [SPIRE].

[9] L. Senatore, Bias in the effective field theory of large scale structures, JCAP 11 (2015) 007 [arXiv:1406.7843] [SPIRE].

[10] L. Senatore and M. Zaldarriaga, Redshift space distortions in the effective field theory of large scale structures, arXiv:1409.1225 [SPIRE].

[11] Ya. B. Zeldovich, Gravitational instability: an approximate theory for large density perturbations, Astron. Astrophys. 5 (1970) 84 [SPIRE].
[12] S. Foreman and L. Senatore, The EFT of large scale structures at all redshifts: analytical predictions for lensing, JCAP 04 (2016) 033 [arXiv:1503.01775] [inSPIRE].

[13] T. Baldauf, M. Mirbabayi, M. Simonović and M. Zaldarriaga, LSS constraints with controlled theoretical uncertainties, arXiv:1602.00674 [inSPIRE].

[14] M.H. Goroff, B. Grinstein, S.J. Rey and M.B. Wise, Coupling of modes of cosmological mass density fluctuations, Astrophys. J. 311 (1986) 6 [inSPIRE].

[15] F. Bernardeau, S. Colombi, E. Gaztanaga and R. Scoccimarro, Large scale structure of the universe and cosmological perturbation theory, Phys. Rept. 367 (2002) 1 [astro-ph/0112551] [inSPIRE].

[16] T. Matsubara, Nonlinear perturbation theory with halo bias and redshift-space distortions via the Lagrangian picture, Phys. Rev. D 78 (2008) 083519 [Erratum ibid. D 78 (2008) 109901] [arXiv:0807.1733] [inSPIRE].

[17] M. Crocce and R. Scoccimarro, Renormalized cosmological perturbation theory, Phys. Rev. D 73 (2006) 063519 [astro-ph/0509418] [inSPIRE].

[18] F. Bernardeau, M. Croce and R. Scoccimarro, Multi-point propagators in cosmological gravitational instability, Phys. Rev. D 78 (2008) 103521 [arXiv:0806.2334] [inSPIRE].

[19] M. Croce, R. Scoccimarro and F. Bernardeau, MPTbreeze: a fast renormalized perturbative scheme, Mon. Not. Roy. Astron. Soc. 427 (2012) 2537 [arXiv:1207.1465] [inSPIRE].

[20] F. Bernardeau, M. Croce and R. Scoccimarro, Constructing regularized cosmic propagators, Phys. Rev. D 85 (2012) 123519 [arXiv:1112.3895] [inSPIRE].

[21] J. Carlson, B. Reid and M. White, Convolution Lagrangian perturbation theory for biased tracers, Mon. Not. Roy. Astron. Soc. 429 (2013) 1674 [arXiv:1209.0780] [inSPIRE].

[22] N. Hand, U. Seljak, F. Beutler and Z. Vlah, Extending the modeling of the anisotropic galaxy power spectrum to $k = 0.4 \, h\text{Mpc}^{-1}$, JCAP 10 (2017) 009 [arXiv:1706.02362] [inSPIRE].

[23] BOSS collaboration, The clustering of galaxies in the completed SDSS-III Baryon Oscillation Spectroscopic Survey: cosmological analysis of the DR12 galaxy sample, Mon. Not. Roy. Astron. Soc. 470 (2017) 2617 [arXiv:1607.03155] [inSPIRE].

[24] BOSS collaboration, The clustering of galaxies in the completed SDSS-III Baryon Oscillation Spectroscopic Survey: baryon acoustic oscillations in the Fourier space, Mon. Not. Roy. Astron. Soc. 464 (2017) 3409 [arXiv:1607.03149] [inSPIRE].

[25] BOSS collaboration, The clustering of galaxies in the completed SDSS-III Baryon Oscillation Spectroscopic Survey: anisotropic galaxy clustering in Fourier-space, Mon. Not. Roy. Astron. Soc. 466 (2017) 2242 [arXiv:1607.03150] [inSPIRE].

[26] H. Gil-Marín et al., The power spectrum and bispectrum of SDSS DR11 BOSS galaxies — I. Bias and gravity, Mon. Not. Roy. Astron. Soc. 451 (2015) 539 [arXiv:1407.5668] [inSPIRE].

[27] J.J.M. Carrasco, S. Foreman, D. Green and L. Senatore, The effective field theory of large scale structures at two loops, JCAP 07 (2014) 057 [arXiv:1310.0464] [inSPIRE].

[28] S.M. Carroll, S. Leichenauer and J. Pollack, Consistent effective theory of long-wavelength cosmological perturbations, Phys. Rev. D 90 (2014) 023518 [arXiv:1310.2920] [inSPIRE].

[29] L. Senatore and M. Zaldarriaga, The IR-resummed effective field theory of large scale structures, JCAP 02 (2015) 013 [arXiv:1404.5954] [inSPIRE].

[30] T. Baldauf, E. Schaan and M. Zaldarriaga, On the reach of perturbative methods for dark matter density fields, JCAP 03 (2016) 007 [arXiv:1507.02255] [inSPIRE].

[31] S. Foreman, H. Perrier and L. Senatore, Precision comparison of the power spectrum in the EFTofLSS with simulations, JCAP 05 (2016) 027 [arXiv:1507.05326] [inSPIRE].
[32] T. Baldauf, L. Mercolli and M. Zaldarriaga, Effective field theory of large scale structure at two loops: the apparent scale dependence of the speed of sound, *Phys. Rev. D* 92 (2015) 123007 [arXiv:1507.02256] [inSPIRE].

[33] M. Cataneo, S. Foreman and L. Senatore, Efficient exploration of cosmology dependence in the EFT of LSS, *JCAP* 04 (2017) 026 [arXiv:1606.03633] [inSPIRE].

[34] M. Lewandowski and L. Senatore, IR-safe and UV-safe integrands in the EFT of LSS with exact time dependence, *JCAP* 08 (2017) 037 [arXiv:1701.07012] [inSPIRE].

[35] T. Konstandin, R.A. Porto and H. Rubira, The effective field theory of large scale structure at three loops, *arXiv:1906.00997* [inSPIRE].

[36] E. Pajer and M. Zaldarriaga, On the renormalization of the effective field theory of large scale structures, *JCAP* 08 (2013) 037 [arXiv:1301.7182] [inSPIRE].

[37] A.A. Abolhasani, M. Mirbabayi and E. Pajer, Systematic renormalization of the effective theory of large scale structure, *JCAP* 05 (2016) 063 [arXiv:1509.07886] [inSPIRE].

[38] M. McQuinn and M. White, Cosmological perturbation theory in $1 + 1$ dimensions, *JCAP* 01 (2016) 043 [arXiv:1502.07389] [inSPIRE].

[39] T. Baldauf, M. Mirbabayi, M. Simonović and M. Zaldarriaga, Equivalence principle and the baryon acoustic peak, *Phys. Rev. D* 92 (2015) 043514 [arXiv:1504.04366] [inSPIRE].

[40] L. Senatore and G. Trevisan, On the IR-resummation in the EFT of LSS, *JCAP* 05 (2018) 019 [arXiv:1710.02178] [inSPIRE].

[41] M. Lewandowski and L. Senatore, An analytic implementation of the IR-resummation for the BAO peak, *JCAP* 03 (2020) 018 [arXiv:1810.11855] [inSPIRE].

[42] D. Blas, M. Garny, M.M. Ivanov and S. Sibiryakov, Time-sliced perturbation theory II: baryon acoustic oscillations and infrared resummation, *JCAP* 07 (2016) 028 [arXiv:1605.02149] [inSPIRE].

[43] R.E. Angulo, S. Foreman, M. Schmittfull and L. Senatore, The one-loop matter bispectrum in the effective field theory of large scale structures, *JCAP* 10 (2015) 039 [arXiv:1406.4143] [inSPIRE].

[44] T. Baldauf, L. Mercolli, M. Mirbabayi and E. Pajer, The bispectrum in the effective field theory of large scale structure, *JCAP* 05 (2015) 007 [arXiv:1406.4135] [inSPIRE].

[45] D. Bertolini, K. Schutz, M.P. Solon and K.M. Zurek, The trispectrum in the effective field theory of large scale structure, *JCAP* 06 (2016) 052 [arXiv:1604.01770] [inSPIRE].

[46] T. Baldauf, E. Schaan and M. Zaldarriaga, On the reach of perturbative descriptions for dark matter displacement fields, *JCAP* 03 (2016) 017 [arXiv:1505.07098] [inSPIRE].

[47] M. Mirbabayi, F. Schmidt and M. Zaldarriaga, Biased tracers and time evolution, *JCAP* 07 (2015) 030 [arXiv:1412.5169] [inSPIRE].

[48] R. Angulo, M. Fasiello, L. Senatore and Z. Vlah, On the statistics of biased tracers in the effective field theory of large scale structures, *JCAP* 09 (2015) 029 [arXiv:1503.08826] [inSPIRE].

[49] T. Fujita, V. Mauerhofer, L. Senatore, Z. Vlah and R. Angulo, Very massive tracers and higher derivative biases, *JCAP* 01 (2020) 009 [arXiv:1609.00717] [inSPIRE].

[50] A. Perko, L. Senatore, E. Jennings and R.H. Wechsler, Biased tracers in redshift space in the EFT of large-scale structure, *arXiv:1610.09321* [inSPIRE].

[51] E.O. Nadler, A. Perko and L. Senatore, On the bispectra of very massive tracers in the effective field theory of large-scale structure, *JCAP* 02 (2018) 058 [arXiv:1710.10308] [inSPIRE].
[52] P. McDonald and A. Roy, *Clustering of dark matter tracers: generalizing bias for the coming era of precision LSS*, *JCAP* **08** (2009) 020 [arXiv:0902.0991] [INSPIRE].

[53] M. Lewandowski, L. Senatore, F. Prada, C. Zhao and C.-H. Chuang, *EFT of large scale structures in redshift space*, *Phys. Rev. D* **97** (2018) 063526 [arXiv:1512.06831] [INSPIRE].

[54] G. Cusin, M. Lewandowski and F. Vernizzi, *Dark energy and modified gravity in the effective field theory of large-scale structure*, *JCAP* **04** (2018) 005 [arXiv:1712.02783] [INSPIRE].

[55] B. Bose, K. Koyama, M. Lewandowski, F. Vernizzi and H.A. Winther, *Towards precision constraints on gravity with the effective field theory of large-scale structure*, *JCAP* **04** (2018) 063 [arXiv:1802.01566] [INSPIRE].

[56] V. Assassi, D. Baumann, E. Pajer, Y. Welling and D. van der Woude, *Effective theory of large-scale structure with primordial non-Gaussianity*, *JCAP* **11** (2015) 024 [arXiv:1505.06668] [INSPIRE].

[57] V. Assassi, D. Baumann, F. Schmidt, *Galaxy bias and primordial non-Gaussianity*, *JCAP* **12** (2015) 043 [arXiv:1610.03723] [INSPIRE].

[58] D. Bertolini, K. Schutz, M.P. Solon, J.R. Walsh and K.M. Zurek, *Non-Gaussian covariance of the matter power spectrum in the effective field theory of large scale structure*, *Phys. Rev. D* **93** (2016) 123505 [arXiv:1512.07630] [INSPIRE].

[59] D. Bertolini and M.P. Solon, *Principal shapes and squeezed limits in the effective field theory of large scale structure*, *JCAP* **11** (2016) 030 [arXiv:1608.01310] [INSPIRE].

[60] R. de Belsunce and L. Senatore, *Tree-level bispectrum in the effective field theory of large-scale structure extended to massive neutrinos*, *JCAP* **02** (2019) 038 [arXiv:1804.06849] [INSPIRE].

[61] M. Simonović, T. Baldauf, M. Zaldarriaga, J.J. Carrasco and J.A. Kollmeier, *Cosmological perturbation theory using the FFTLog: formalism and connection to QFT loop integrals*, *JCAP* **04** (2018) 030 [arXiv:1608.08130] [INSPIRE].

[62] C. Alcock and B. Paczynski, *An evolution free test for non-zero cosmological constant*, *Nature* **281** (1979) 358 [INSPIRE].

[63] H. Gil-Marín et al., *The clustering of galaxies in the SDSS-III Baryon Oscillation Spectroscopic Survey: RSD measurement from the power spectrum and bispectrum of the DR12 BOSS galaxies*, *Mon. Not. Roy. Astron. Soc.* **465** (2017) 1757 [arXiv:1606.00439] [INSPIRE].

[64] W.E. Ballinger, J.A. Peacock and A.F. Heavens, *Measuring the cosmological constant with redshift surveys*, *Mon. Not. Roy. Astron. Soc.* **282** (1996) 877 [astro-ph/9605017] [INSPIRE].

[65] Y.-S. Song, A. Taruya and A. Oka, *Cosmology with anisotropic galaxy clustering from the combination of power spectrum and bispectrum*, *JCAP* **08** (2015) 007 [arXiv:1502.03099] [INSPIRE].

[66] M.J. Wilson, J.A. Peacock, A.N. Taylor and S. de la Torre, *Rapid modelling of the redshift-space power spectrum multipoles for a masked density field*, *Mon. Not. Roy. Astron. Soc.* **464** (2017) 3121 [arXiv:1511.07799] [INSPIRE].

[67] F. Beutler, E. Castorina and P. Zhang, *Interpreting measurements of the anisotropic galaxy power spectrum*, *JCAP* **03** (2019) 040 [arXiv:1810.05051] [INSPIRE].

[68] E. Castorina et al., *Redshift-weighted constraints on primordial non-Gaussianity from the clustering of the eBOSS DR4 quasars in Fourier space*, *JCAP* **09** (2019) 010 [arXiv:1904.08859] [INSPIRE].

[69] C. Hahn, R. Scoccimarro, M.R. Blanton, J.L. Tinker and S.A. Rodriguez-Torres, *The effect of fibre collisions on the galaxy power spectrum multipoles*, *Mon. Not. Roy. Astron. Soc.* **467** (2017) 1940 [arXiv:1609.01714] [INSPIRE].
H. Gil-Marín et al., *The clustering of galaxies in the SDSS-III Baryon Oscillation Spectroscopic Survey: RSD measurement from the LOS-dependent power spectrum of DR12 BOSS galaxies*, Mon. Not. Roy. Astron. Soc. **460** (2016) 4188 [arXiv:1509.06386] [inSPIRE].

D. Foreman-Mackey, D.W. Hogg, D. Lang and J. Goodman, *emcee: the MCMC hammer*, Publ. Astron. Soc. Pac. **125** (2013) 306 [arXiv:1202.3665] [inSPIRE].

A. Gelman and D.B. Rubin, *Inference from iterative simulation using multiple sequences*, Statist. Sci. **7** (1992) 457.

T. Colas, G. D'amico, L. Senatore, P. Zhang and F. Beutler, *Efficient cosmological analysis of the SDSS/BOSS data from the effective field theory of large-scale structure*, arXiv:1909.07951 [inSPIRE].

A. Schneider et al., *Matter power spectrum and the challenge of percent accuracy*, JCAP **04** (2016) 047 [arXiv:1503.05920] [inSPIRE].

V. Springel, *The cosmological simulation code GADGET-2*, Mon. Not. Roy. Astron. Soc. **364** (2005) 1105 [astro-ph/0505010] [inSPIRE].

V.F. Mukhanov, *CMB-slow, or how to estimate cosmological parameters by hand*, Int. J. Theor. Phys. **43** (2004) 623 [astro-ph/0303072] [inSPIRE].

A.J.S. Hamilton, *Formulae for growth factors in expanding universes containing matter and a cosmological constant*, Mon. Not. Roy. Astron. Soc. **322** (2001) 419 [astro-ph/0006089] [inSPIRE].

P. Lemos, E. Lee, G. Efstathiou and S. Gratton, *Model independent H(z) reconstruction using the cosmic inverse distance ladder*, Mon. Not. Roy. Astron. Soc. **483** (2019) 4803 [arXiv:1806.06781] [inSPIRE].

---

[70] H. Gil-Marín et al., *The clustering of galaxies in the SDSS-III Baryon Oscillation Spectroscopic Survey: RSD measurement from the LOS-dependent power spectrum of DR12 BOSS galaxies*, Mon. Not. Roy. Astron. Soc. **460** (2016) 4188 [arXiv:1509.06386] [inSPIRE].

[71] D. Foreman-Mackey, D.W. Hogg, D. Lang and J. Goodman, *emcee: the MCMC hammer*, Publ. Astron. Soc. Pac. **125** (2013) 306 [arXiv:1202.3665] [inSPIRE].

[72] A. Gelman and D.B. Rubin, *Inference from iterative simulation using multiple sequences*, Statist. Sci. **7** (1992) 457.

[73] T. Colas, G. D'amico, L. Senatore, P. Zhang and F. Beutler, *Efficient cosmological analysis of the SDSS/BOSS data from the effective field theory of large-scale structure*, arXiv:1909.07951 [inSPIRE].

[74] A. Schneider et al., *Matter power spectrum and the challenge of percent accuracy*, JCAP **04** (2016) 047 [arXiv:1503.05920] [inSPIRE].

[75] K. Riebe et al., *The MultiDark database: release of the Bolshoi and MultiDark cosmological simulations*, arXiv:1109.0003 [inSPIRE].

[76] BOSS collaboration, *The clustering of galaxies in the completed SDSS-III Baryon Oscillation Spectroscopic Survey: cosmological implications of the Fourier space wedges of the final sample*, Mon. Not. Roy. Astron. Soc. **467** (2017) 2085 [arXiv:1607.03143] [inSPIRE].

[77] F.-S. Kitaura et al., *The clustering of galaxies in the SDSS-III Baryon Oscillation Spectroscopic Survey: mock galaxy catalogues for the BOSS final data release*, Mon. Not. Roy. Astron. Soc. **456** (2016) 4156 [arXiv:1509.06400] [inSPIRE].

[78] A. Klypin, G. Yepes, S. Gottlober, F. Prada and S. Hess, *MultiDark simulations: the story of dark matter halo concentrations and density profiles*, Mon. Not. Roy. Astron. Soc. **457** (2016) 4340 [arXiv:1411.4001] [inSPIRE].

[79] V. Springel, *The cosmological simulation code GADGET-2*, Mon. Not. Roy. Astron. Soc. **364** (2005) 1105 [astro-ph/0505010] [inSPIRE].

[80] Particle Data Group collaboration, *Review of particle physics*, Phys. Rev. D **98** (2018) 030001 [inSPIRE].

[81] SDSS collaboration, *Detection of the baryon acoustic peak in the large-scale correlation function of SDSS luminous red galaxies*, Astrophys. J. **633** (2005) 560 [astro-ph/0501171] [inSPIRE].

[82] W.J. Percival et al., *Measuring the Baryon Acoustic Oscillation scale using the SDSS and 2dFGRS*, Mon. Not. Roy. Astron. Soc. **381** (2007) 1053 [arXiv:0705.3323] [inSPIRE].

[83] 2dFGRS Team collaboration, *Parameter constraints for flat cosmologies from CMB and 2dFGRS power spectra*, Mon. Not. Roy. Astron. Soc. **337** (2002) 1068 [astro-ph/0206256] [inSPIRE].

[84] Planck collaboration, *Planck 2018 results. VI. Cosmological parameters*, arXiv:1807.06209 [inSPIRE].

[85] V.F. Mukhanov, *CMB-slow, or how to estimate cosmological parameters by hand*, Int. J. Theor. Phys. **43** (2004) 623 [astro-ph/0303072] [inSPIRE].

[86] A.J.S. Hamilton, *Formulae for growth factors in expanding universes containing matter and a cosmological constant*, Mon. Not. Roy. Astron. Soc. **322** (2001) 419 [astro-ph/0006089] [inSPIRE].

[87] P. Lemos, E. Lee, G. Efstathiou and S. Gratton, *Model independent H(z) reconstruction using the cosmic inverse distance ladder*, Mon. Not. Roy. Astron. Soc. **483** (2019) 4803 [arXiv:1806.06781] [inSPIRE].
[88] DES collaboration, Dark Energy Survey year 1 results: cosmological constraints from galaxy clustering and weak lensing, Phys. Rev. D 98 (2018) 043526 [arXiv:1708.01530] [nSPIRE].

[89] WMAP collaboration, Nine-year Wilkinson Microwave Anisotropy Probe (WMAP) observations: cosmological parameter results, Astrophys. J. Suppl. 208 (2013) 19 [arXiv:1212.5226] [nSPIRE].

[90] K.C. Wong et al., H0LiCOW XIII. A 2.4% measurement of H0 from lensed quasars: 5.3σ tension between early and late-universe probes, arXiv:1907.04869 [nSPIRE].

[91] W. Yuan, A.G. Riess, L.M. Macri, S. Casertano and D. Scolnic, Consistent calibration of the tip of the red giant branch in the Large Magellanic Cloud on the Hubble space telescope photometric system and a re-determination of the Hubble constant, Astrophys. J. 886 (2019) 61 [arXiv:1908.00993] [nSPIRE].

[92] C.-T. Chiang, M. LoVerde and F. Villaescusa-Navarro, First detection of scale-dependent linear halo bias in N-body simulations with massive neutrinos, Phys. Rev. Lett. 122 (2019) 041302 [arXiv:1811.12412] [nSPIRE].

[93] BOSS collaboration, The clustering of galaxies in the SDSS-III Baryon Oscillation Spectroscopic Survey: testing gravity with redshift-space distortions using the power spectrum multipoles, Mon. Not. Roy. Astron. Soc. 443 (2014) 1065 [arXiv:1312.4611] [nSPIRE].

[94] T. Baldauf, U. Seljak, L. Senatore and M. Zaldarriaga, Galaxy bias and non-linear structure formation in general relativity, JCAP 10 (2011) 031 [arXiv:1106.5507] [nSPIRE].

[95] T. Baldauf, U. Seljak, L. Senatore and M. Zaldarriaga, Linear response to long wavelength fluctuations using curvature simulations, JCAP 09 (2016) 007 [arXiv:1511.01465] [nSPIRE].

[96] T. Lazeyras, C. Wagner, T. Baldauf and F. Schmidt, Precision measurement of the local bias of dark matter halos, JCAP 02 (2016) 018 [arXiv:1511.01096] [nSPIRE].

[97] C.-T. Chiang, Y. Li, W. Hu and M. LoVerde, Quintessential scale dependence from separate universe simulations, Phys. Rev. D 94 (2016) 123526 [arXiv:1710.01310] [nSPIRE].

[98] C.-T. Chiang, W. Hu, Y. Li and M. Loverde, Scale-dependent bias and bispectrum in neutrino separate universe simulations, Phys. Rev. D 97 (2018) 123 0526 [arXiv:1710.01310] [nSPIRE].

[99] A. Cooray and R.K. Sheth, Halo models of large scale structure, Phys. Rept. 372 (2002) 1 [astro-ph/0206508] [nSPIRE].

[100] L. Senatore, Field theory, geometry and data in cosmology, seminar video, https://www.youtube.com/watch?v=0S3Gsj2jCOQ (2019).

[101] L. Senatore, The effective field theory of large-scale structure applied to BOSS data, conference slides, https://indico.ipmu.jp/event/159/timetable/ (2019).

[102] L. Senatore, Some aspects of theoretical cosmology, conference slides, https://indico.cern.ch/event/577856/contributions/339657/ (2019).

[103] B. Bose, A. Pourtsidou, K. Marković and F. Beutler, Assessing non-linear models for galaxy clustering II: model validation and forecasts for stage IV surveys, arXiv:1905.05122 [nSPIRE].

[104] D. Blas, J. Lesgourgues and T. Tram, The Cosmic Linear Anisotropy Solving System (CLASS) II: approximation schemes, JCAP 07 (2011) 034 [arXiv:1104.2933] [nSPIRE].

[105] S.R. Hinton, ChainConsumer, J. Open Source Softw. 1 (2016) 45.

[106] T. Hahn, CUBA: a library for multidimensional numerical integration, Comput. Phys. Commun. 168 (2005) 78 [hep-ph/0404043] [nSPIRE].

[107] G. Guennebaud et al., Eigen v3, http://eigen.tuxfamily.org, (2010).

[108] D.J. Eisenstein and W. Hu, Baryonic features in the matter transfer function, Astrophys. J. 496 (1998) 605 [astro-ph/9709112] [nSPIRE].