Cosmology with the Redshift-Space Galaxy Bispectrum Monopole at One-Loop Order

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ABSTRACT: We study the cosmological information content of the redshift-space galaxy bispectrum monopole at one-loop order in perturbation theory. We incorporate all effects necessary for comparison to data: fourth-order galaxy bias, infrared resummation (accounting for the non-linear evolution of baryon acoustic oscillations), ultraviolet counterterms, non-linear redshift-space distortions, stochastic contributions, projection, and binning effects. The model is implemented using FFTLog, and validated with the PT Challenge suite of N-body simulations, whose large volume allows for high-precision tests. Focusing on the mass fluctuation amplitude, $\sigma_8$, and galaxy bias parameters, we find that including one-loop corrections allow us to significantly extend the range of scales over which the bispectrum can be modeled, and greatly tightens constraints on bias parameters. However, this does not lead to noticeable improvements in the $\sigma_8$ errorbar due to the necessary marginalization over a large number of nuisance parameters with conservative priors. Analyzing a BOSS-volume likelihood, we find that the addition of the one-loop bispectrum may lead to improvements on primordial non-Gaussianity constraints by $\lesssim 30\%$ and on $\sigma_8$ by $\approx 10\%$, though we caution that this requires pushing the analysis to short-scales where the galaxy bias parameters may not be correctly recovered; this may lead to biases in the recovered parameter values. We conclude that restrictive priors from simulations or higher-order statistics such as the bispectrum multipoles will be needed in order to realize the full information content of the galaxy bispectrum.
1 Introduction

The distribution of matter in the Universe is not Gaussian. As such, any optimal analysis of large-scale structure surveys must involve non-Gaussian statistics. Historically, the distribution of galaxies has been analyzed by way of the power spectrum, which is a complete statistic only in the Gaussian (large-scale) limit. Such analyses have seen particular successes, both in measuring the Universe’s growth rate and expansion history via baryon acoustic oscillation signatures [e.g., 1–8], and in the direct determination of cosmological parameters using full-shape methods [e.g., 9–27]. However, if we wish to extract more information from the observational data-sets (and thus enhance their utility at fixed experimental cost), we must additionally include higher-order statistics, of which the simplest are the bispectrum and three-point correlation function.

Much as the power spectrum describes the distribution of pairs of points, the bispectrum encodes the statistics of triplets [28]. Despite being a subject of significant early work [e.g., 29–37], the bispectrum has been rarely used in modern cosmological analyses (though see [38–40]), owing to difficulties in its measurement [c.f., 41–43], modeling, and interpretation. On the other hand, its potential utility has been demonstrated a number of times, with claims of tight constraints on a number of cosmological parameters, particularly those encoding extensions to the ΛCDM paradigm, such as neutrino masses, primordial non-Gaussianity and the breaking of the equivalence principle [44–65]. To fully utilize this information, we require theoretical models capable of predicting the shape of the bispectrum and its dependence on the parameters of interest. This has been a subject of significant work, starting from the matter bispectrum [66–73], then incorporating the effects of redshift-space distortions [37, 74–76] and galaxy bias [62, 77–81], most successfully using the Effective Field Theory of Large Scale Structure (hereafter EFTofLSS, [82, 83], see [84] for a recent review).

Whilst the above references have been pivotal to the development of a bispectrum model, few contain all the necessary ingredients to allow for robust comparison of theory and observation. In particular, one must account for the backreaction of short-scale physics on the large-scale bispectrum [66, 67, 85], long-wavelength displacements [86–93], and survey geometry [94, 95], all of which can lead to biases in derived parameters if not properly accounted for. In [80] a complete model for the tree-level (leading-order) bispectrum of galaxies in redshift-space was presented and validated, including all the above effects (see also [19]). This allows for precise modelling of the angle-averaged bispectrum monopole, and has facilitated a number of analyses constraining ΛCDM parameters [15] and primordial non-Gaussianity [96, 97]. However, this model was restricted to relatively large scales.
If we wish to further exploit the constraining power of the bispectrum, we must push to smaller scales, by extending the perturbation theory to next order. Whilst [98] has recently demonstrated some work in this direction, a full model for the one-loop bispectrum (including all relevant phenomena such as projection effects) has not yet been presented and validated with simulations.

In this work, we present a complete and systematic computation of the redshift-space galaxy bispectrum monopole at one-loop order. This includes all effects necessary to compare with observational data: deterministic contributions, counterterms, bias renormalization, stochasticity, bin-averaging, and coordinate distortions. This involves the galaxy density at fourth-order: we systematically account for all bias operators (following [79]), and include full treatment of all necessary redshift-space counterterms, ensuring a convergent Taylor series. Our model necessarily depends on a number of free parameters: these account for the unknown complexities of ultraviolet physics (such as galaxy formation physics and feedback), and ensure physical robustness. Efficient computation of the one-loop bispectrum is non-trivial; as such, we devote a significant portion of this work to discussing its practical computation with the FFTLog algorithm [99]. We compare the theoretical predictions to real- and redshift-space bispectra obtained from the PT Challenge simulations [100], which serve both to validate the approach and to assess the information content of the one-loop bispectrum model. Though we restrict to the measurement of $\sigma_8$ and primordial non-Gaussianity parameters, one can constrain a variety of other phenomena with the bispectrum, and, further still, our methodology can be extended to other correlators including the bispectrum multipoles [65, 101] and the recently-detected trispectrum [102, 103].

The remainder of this paper is structured as follows. The theoretical model is presented in §2, before its implementation is outlined in §3. In §4 we give details of the data and analysis choices used to validate the model, before presenting the results of likelihood analyses using the real- and redshift-space galaxy bispectrum in §5 & 6 respectively. §7 comments on the method’s applicability to current datasets, with a summary and discussion given in §8. Finally, various technical details are presented in the Appendices, including: A the perturbation theory kernels, B details of the bispectrum integration routines, C discussion of the redshift-space counterterms, and D derivation of the stochastic bispectrum components. Appendix E is devoted to prior volume effects. The key plots of this work are Fig. 1, showing the one-loop bispectrum components, and Fig. 4, displaying the utility of the bispectrum for a BOSS-like survey.

2 Theoretical Model for the One-Loop Bispectrum

In this work, we analyze the power spectrum and bispectrum of biased tracers (i.e. galaxies) in redshift space at one-loop order. Whilst the one-loop power spectrum and tree-level bispectrum have been described in detail before [e.g. 19, 80], a complete model for the one-loop bispectrum has not been presented before (though some aspects can be found in [98]) and will be discussed below, with additional technical details found in the appendices. Here,
we will restrict to Gaussian initial conditions; extension to primordial non-Gaussianity is discussed in §5.2.

In the EFTofLSS, the bispectrum is comprised of the following terms at one-loop order [e.g., 66, 85, 104]:

\[ B_{1\text{-loop}}(k_1, k_2, k_3) = B_{211} + \left[ B_{222} + B_{321}^{I} + B_{321}^{II} + B_{411} \right] + B_{\text{ct}} + B_{\text{stoch}}, \]  

where the first and second terms give the tree-level and one-loop bispectrum in Eulerian perturbation theory, \( B_{\text{ct}} \) is the derivative and counterterm contribution, and \( B_{\text{stoch}} \) encodes stochasticity. This is strictly a function of five variables: three lengths, \( \{k_1, k_2, k_3\} \) and two angles, \( \{\mu_1, \mu_2\} \), for \( \mu_i \equiv \hat{k}_i \cdot \hat{n} \) with line-of-sight \( \hat{n} \) (hereafter LoS), noting that \( k_1 \mu_1 + k_2 \mu_2 + k_3 \mu_3 = 0 \). In real-space, this reduces to just three variables: \( \{k_1, k_2, k_3\} \).

### 2.1 Bias Expansion

To compute the bispectrum within Eulerian perturbation theory, our first step is to express real-space galaxy density field, \( \delta_g \), in terms of a basis of bias operators, i.e. all combinations of the density and velocity fields \( (\delta \text{ and } \theta) \) consistent with the relevant symmetries up to a given order in perturbation theory [79, 105–110]. For the one-loop bispectrum, we require terms up to fourth-order \( (\delta^4) \), and here use the basis of Galileon operators proposed in [79]:

\[
\delta_g = \left\{ b_1 \delta \right\} + \left\{ \frac{b_2}{2} \delta^2 + \gamma_2 \mathcal{G}_2(\Phi_v) \right\} \\
+ \left\{ \frac{b_3}{6} \delta^3 + \gamma_2^2 \delta \mathcal{G}_2(\Phi_v) + \gamma_3 \mathcal{G}_3(\Phi_v) + \gamma_{21} \mathcal{G}_2(\varphi_2, \varphi_1) \right\} \\
+ \left\{ \gamma_{21}^2 \delta \mathcal{G}_2(\varphi_2, \varphi_1) + \gamma_{211} \mathcal{G}_3(\varphi_2, \varphi_1, \varphi_1) + \gamma_{22} \mathcal{G}_2(\varphi_2, \varphi_2) + \gamma_{31} \mathcal{G}_2(\varphi_3, \varphi_1) \right\} + \mathcal{O}(\delta^5),
\]  

where curly brackets separate operators of different order and the bias parameters are marked in color. In (2.2), we drop any terms that do not appear in the one-loop bispectrum; these are all composite local evolution operators such as \( \delta^4 \) and \( \delta^2 \mathcal{G}_2(\Phi_v) \). Here we have ignored both higher-derivative operators (which we return to below) and bias renormalization, which is discussed in Appendix B. The Galileon operators are defined by

\[
\mathcal{G}_2(\Phi_v) \equiv \nabla_i \nabla_j \Phi_v \nabla^i \nabla^j \Phi_v - (\nabla^2 \Phi_v)^2, \\
\mathcal{G}_3(\Phi_v) \equiv 2 \nabla_i \nabla_j \Phi_v \nabla^i \nabla^j \Phi_v \nabla^k \Phi_v - 3 \nabla_i \nabla_j \Phi_v \nabla^i \nabla^j \Phi_v \nabla^2 \Phi_v + (\nabla^2 \Phi_v)^3,
\]  

where \( \Phi_v \equiv \nabla^{-2} \theta \) is the velocity potential, equal to the Newtonian potential \( \Phi \equiv \nabla^{-2} \delta \) at leading order. These can be simply generalized to functions of multiple potentials, with (2.2) involving the LPT potentials \( \varphi_{1,2} \), satisfying

\[
\nabla^2 \varphi_1 = -\delta, \quad \nabla^2 \varphi_2 = -\mathcal{G}_2(\varphi_1).
\]  

Up to third order, this is equivalent to the bias expansion used in [106] and previous works [e.g., 10, 80], with the relations\(^1\)

\[
\gamma_2 \equiv b_{\mathcal{G}_2}, \quad \gamma_{21} \equiv -\frac{4}{3} (b_{\mathcal{G}_2} + b_{\Gamma_3}).
\]  

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\(^1\)This is obtained by noting that \( \mathcal{G}_2(\Phi_v) \) is just the usual tidal tensor \( \mathcal{G}_2 \equiv -\mathcal{G}_2(\delta) \) at first order, and that \( \mathcal{G}_2(\varphi_2, \varphi_1) \) is a combination of the \( \mathcal{G}_2 \) and \( \Gamma_3 \) operators of [106].
Utilizing (2.2), and expanding each operator in terms of the linear density field $\delta^{(1)} = \delta_L$, we can define the $n$-th order contributions to the galaxy density field:

$$\delta_g^{(n)}(k) = \int_{q_1...q_n} (2\pi)^3 \delta_D(q_1 + ... + q_n - k) K_n(q_1, ..., q_n) \delta_L(q_1) ... \delta_L(q_n), \quad (2.6)$$

where the real-space kernels $K_n$ are given in Appendix A.1 and depend on the bias parameters given above. Furthermore, this generalizes to the redshift-space density field, $\delta_s(k)$, using the well-known mapping [e.g., 111]

$$\delta_s(k) = \delta_g(k) + \int d\vec{x} e^{-i\vec{k} \cdot \vec{x}} \left[ e^{-i\mu \cdot \vec{n} \cdot \vec{x}} - 1 \right] (1 + \delta_g(\vec{x})), \quad (2.7)$$

where $f$ is the logarithmic growth rate, $u_z(\vec{q}) = (i\mu \cdot \vec{q})\theta(\vec{q})$ is the Fourier-space LoS velocity field, and $\mu_q \equiv \vec{q} \cdot \hat{n}$, for LoS vector $\hat{n}$. The associated kernels, analogous to (2.6), are labelled as $Z_n$ and defined for $n \leq 4$ in Appendix A.2.

### 2.2 Deterministic Contributions

Utilizing the redshift-space kernels of Appendix A.2, the tree-level bispectrum, $B_{211} \equiv \left< \delta_g^{(2)} \delta_g^{(1)} \delta_g^{(1)} \right>$, can be written

$$B_{211}(k_1, k_2, k_3) = 2Z_2(k_1, k_2)Z_1(k_1)Z_1(k_2)P_L(k_1)P_L(k_2) + 2 \text{ cyc.}, \quad (2.8)$$

where $P_L(k)$ is the linear power spectrum (though see the below discussion on infrared resummation). This depends on the bias parameters $\{b_1, b_2, \gamma_2\}$, as well as the growth rate, $f(z)$. The one-loop terms can be written as loop integrals over the linear power spectrum, and come in four flavors [e.g., 66, 104]:

$$B_{222}(k_1, k_2, k_3) = 8 \int_q Z_2(k_1 + q, -q)Z_2(k_1 + q, k_2 - q)Z_2(k_2 - q, q) \times P_L(q)P_L(|k_1 + q|)P_L(|k_2 - q|), \quad (2.9)$$

$$B_{321}^I(k_1, k_2, k_3) = 6Z_1(k_1)P_L(k_1)\int_q Z_3(-q, q - k_2, -k_1)Z_2(q, k_2 - q) \times P_L(q)P_L(|k_2 - q|) + 5 \text{ perm.}, \quad (2.10)$$

$$B_{321}^{II}(k_1, k_2, k_3) = 6Z_2(k_1, k_2)Z_1(k_2)P_L(k_1)P_L(k_2)\int_q Z_3(k_1, q, -q)P_L(q) + 5 \text{ perm.}, \quad (2.11)$$

$$B_{411}(k_1, k_2, k_3) = 12Z_1(k_1)Z_1(k_2)P_L(k_1)P_L(k_2)\int_q Z_4(k_1, k_2, q, -q)P_L(q) + 2 \text{ cyc.}, \quad (2.12)$$

where the $B_{321}^I$ spectrum is similar to the $P_{13}(k_1)$ contribution to the one-loop power spectrum. Computation of the loop integrals can be performed via explicit numerical integration or with the FFTLog method [99]; we discuss the latter in §3, with details presented in Appendix B. As well as the tree-level biases, these spectra involve the higher-order parameters $\{b_3, \gamma_3, \gamma_21, \gamma_22, \gamma_31\}$, of which only $\gamma_21$ appears in the one-loop power spectrum.
2.3 Counterterms

To ensure a self-consistent theoretical model, we require a set of counterterms, which account for non-idealities in fluid equations (via the viscous stress tensor), and absorb the unknown ultraviolet (UV, \( q \gg k \)) behavior of the loop integrals in (2.9) [e.g., 66, 67, 85]. For the one-loop bispectrum in real-space, these operators are degenerate with derivative operators in the bias expansion, such as \( \nabla^2 \delta \). Furthermore, the redshift-space bispectrum contains additional counterterms that appear after the renormalization of contract operators in the perturbative mapping of (2.7); these are discussed in detail in Appendix C.

The overall bispectrum counterterm contribution can be written as

\[
B_{\text{ctr}}(k_1, k_2, k_3) = 2Z_1(k_1)Z_1(k_2)(F_{2,\text{ctr}}^2 + Z_{2,\text{ctr}}^2)(k_1, k_2)P_L(k_1)P_L(k_2) + \text{cyclic perms.},
\]

(2.10)

where \( F_{2,\text{ctr}}^2(k_1, k_2) \) is the real-space counterterm kernel [79]:

\[
F_{2,\text{ctr}}^2(k_1, k_2) = - \left\{ \left[ \beta_{B,a}(k_1^2 + k_2^2) + \beta_{B,b}k_1^2 \right] F_2(k_1, k_2) \right. 
\]

\[
\left. + \left[ \beta_{B,c}(k_1^2 + k_2^2) + \beta_{B,d}k_1^2 \right] \kappa(k_1, k_2) + \beta_{B,e} k_1 \cdot k_2 \right\} \frac{1}{k_{\text{NL}}^2},
\]

(2.11)

where \( \kappa(k_1, k_2) = (\hat{k}_1 \cdot \hat{k}_2) - 1 \) (cf. A.2), and we choose the non-linear scale \( k_{\text{NL}} = 0.45h \text{ Mpc}^{-1} \) [44, 112, 113]. (2.10) additionally involves the \( \mu \)-dependent redshift-space kernel \( Z_{2,\text{ctr}}^2 \) defined by

\[
Z_{2,\text{ctr}}^2(k_1, k_2) = \frac{1}{k_{\text{NL}}^2} \left( C_1 k_{3z}^2 F_2(k_1, k_2) + C_2 k_{3z}^2 \left( \mu_3^2 - \frac{1}{3} \right) F_2(k_1, k_2) + C_3 k_{3z}^2 \right)
\]

\[
+ C_4 k_{3z}^2 \left( \frac{(k_1 \cdot k_2)\mu_1\mu_2}{k_1 k_2} - \frac{\mu_1^2 + \mu_2^2}{3} + \frac{1}{9} \right) + C_5 f k_{3z}^2 \mu_3^2 G_2(k_1, k_2)
\]

\[
+ C_6 f k_{3z}^2 \left[ \frac{\mu_1}{k_1} \left( \mu_2 - \frac{1}{3} \right) + \frac{\mu_2}{k_2} \left( \mu_1 - \frac{1}{3} \right) \right] + C_7 f k_{3z}^2 \left[ \frac{\mu_1}{k_1} + \frac{\mu_2}{k_2} \right]
\]

\[
+ C_8 f^2 k_{3z}^2 \left( \frac{\mu_1}{k_1^2} \mu_2 + C_9 k_{3z}^2 \left( \frac{(k_1 \cdot k_2)^2}{k_1^2 k_2^2} - \frac{1}{3} \right) \right),
\]

(2.12)

as derived in Appendix C, with \( k_{3z} \equiv k_3 \mu_3 \). In principle, two combinations of \( C_1, C_2 \) and \( C_5 \) are constrained by the power spectrum, so only one parameter out of three is independent here. In practice, however, we did not find any difference between imposing the power spectrum constraints on \( C_1, C_2, C_3 \) or treating them as free parameters. This is why we proceed with keeping them free in what follows. In total, the one-loop bispectrum counterterm depends on 14 free parameters, \( \{ \beta_{B,i} \} \), and \( \{ C_i \} \) in addition to the one-loop power spectrum counterterms.\(^2\)

Notably, many of the counterterms appearing in (2.12) are degenerate at the bispectrum monopole level; nevertheless, we prefer to keep all of them in the model, and marginalize over them within physically motivated priors. This is done for two main reasons. First, terms with different powers of \( \mu \) can, in principle, be distinguished even at the

\(^2\)Some of the bispectrum counterterms are time integrals of the power spectrum counterterms, but since we do not know the time-dependence, we prefer to keep all the counterterms free.
bispectrum monopole level thanks to the Alcock-Paczynski projection effect [114], which is described below. Second, the degeneracy between these terms can be broken with higher order angular multipole moments of the bispectrum [37, 41], which we will analyze in the future.

2.4 Stochasticity

Contributions to the bispectrum are also sourced by the non-deterministic part of the density field $\epsilon$ [106–110], i.e. that uncorrelated with $\delta_L$. At tree-level, this gives two terms, $\propto 1/\bar{n}$, $P(k)/\bar{n}$ (arising from Poissonian shot-noise with sample density $\bar{n}$), whilst at one-loop order, we must keep contributions suppressed by $(k/k_{\text{halo}})^2$, where $k_{\text{halo}}^{-1}$ is some characteristic halo size. From [79], we have the following form at next-to-leading order in real-space:

$$B_{\text{stoch}}(k_1, k_2, k_3)|_{\text{real-space}} = \frac{A_{\text{shot},0}}{\bar{n}^2} + \frac{A_{\text{shot},1}}{\bar{n}^2} (k_1^2 + k_2^2 + k_3^2) \tag{2.13}$$

$$+ \left[ B_{\text{shot}} + S_1 \frac{k_1^2}{k_{NL}^2} + S_0 \frac{k_2^2 + k_3^2}{k_{NL}^2} \right] \frac{P_L(k_1)}{\bar{n}} + 2 \text{ cyc.},$$

depending on another five free parameters $\{A_{\text{shot},0}, B_{\text{shot}}\}$ and $\{A_{\text{shot},1}, S_0, S_1\}$, which cannot be constrained with the one-loop power spectrum. In the Poisson limit, $A_{\text{shot},0} = B_{\text{shot}} = 1$, with all higher-order terms (arising, for example, from halo exclusion) vanish.

In redshift-space, significantly more dependencies arise. A systematic derivation of these is presented in Appendix D and yields the following expression:

$$B_{\text{stoch}} = B^{(P\bar{n}^{-1})}_{\text{stoch}} + B^{(\bar{n}^{-2})}_{\text{stoch}} + B^{(k^2 P\bar{n}^{-1})}_{\text{stoch}} + B^{(k^2 \bar{n}^{-2})}_{\text{stoch}},$$

where

$$B^{(P\bar{n}^{-1})}_{\text{stoch}}(k_1, k_2, k_3) = \left[ b_1 B_{\text{shot}} + f \mu^2 (1 + P_{\text{shot}}) \right] \frac{Z_1(k_1) P_L(k_1)}{\bar{n}} + \text{cyc.},$$

$$B^{(\bar{n}^{-2})}_{\text{stoch}}(k_1, k_2, k_3) = \frac{A_{\text{shot},0}}{\bar{n}^2},$$

$$B^{(k^2 P\bar{n}^{-1})}_{\text{stoch}}(k_1, k_2, k_3) = \sum_{n=1,2,3,4} S_n k_1^2 \mu_1^{2(n-1)} + S_5 Z_1(k_1) (\mu_2^2 + \mu_3^2)$$

$$+ S_6 Z_1(k_1) \mu_1^2 (\mu_2 k_2^2 + \mu_3 k_3^2) + S_7 Z_1(k_1) \frac{k_1^2}{k_{NL}^2} (k_2^2 + k_3^2) + S_0 Z_1(k_1) (k_2^2 + k_3^2) \frac{P_L(k_1)}{k_{NL}^2 \bar{n}} + \text{cyc.},$$

$$B^{(k^2 \bar{n}^{-2})}_{\text{stoch}}(k_1, k_2, k_3) = \left[ A_{\text{shot},1} (k_1^2 + k_2^2 + k_3^2) + A_{\text{shot},2} (k_2^2 + k_1^2 + k_3^2) \right] \frac{1}{k_{NL}^2 \bar{n}^2}. \tag{2.15}$$

This expression shares the parameter $P_{\text{shot}}$ with the power spectrum, but includes an additional 12 nuisance coefficients: $\{S_n\}, \{A_{\text{shot},n}\}, B_{\text{shot}}$. $P_{\text{shot}}$ is defined as a constant rescaling of the stochastic power spectrum [80]:

$$\langle \epsilon(k) \epsilon(k') \rangle = (2\pi)^3 \delta_D^{(3)}(k + k') \frac{1 + P_{\text{shot}}}{\bar{n}}. \tag{2.16}$$

Note that in the absence of projection effects the counterterms $A_{\text{shot},1}$ and $A_{\text{shot},2}$ are fully degenerate. Therefore, for the purposes of this study we will set $A_{\text{shot},2} = 0$. 

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### 2.5 Infrared Resummation

An additional complication arises from the effects of long-wavelength displacements, which can be consistently treated using “infrared resummation”. A rigorous derivation of this was presented in [86, 87] in the context of time-sliced perturbation theory [115], and, at tree-level order, can be implemented by replacing the linear power spectrum $P_L$ with its IR-resummed equivalent, i.e.

$$P_L(k) \rightarrow P_{nw}(k) + e^{-\Sigma^2 k^2 (1 + f \mu^2 (2 + f))} P_w(k)$$

(2.17)

where $P_w$ and $P_{nw}$ are the wiggly and smooth parts of the power spectrum respectively. This has the effect of damping the oscillatory component by a $k$- and $\mu$-dependent factor. The damping scales are given in terms of the broadband power spectrum as

$$\Sigma^2 = \frac{1}{6 \pi^2} \int_0^{k_S} dq P_{nw}(q) [1 - j_0(q r_{BAO}) + 2j_2(q r_{BAO})]$$

(2.18)

$$\delta \Sigma^2 = \frac{1}{2 \pi^2} \int_0^{k_S} dq P_{nw}(q) j_2(q r_{BAO})$$

where $r_{BAO}$ is the sound-horizon scale and $k_S \sim 0.1 h \text{Mpc}^{-1}$. At one-loop order, the IR-resummed bispectrum can be written schematically as

$$B_{1\text{-loop}} = B_{211}[P_{nw} + (1 + k^2 \Sigma^2_{\text{tot}}) e^{-\Sigma^2_{\text{tot}} k^2 P_w}]$$

+ \left( B_{3321}^I + B_{321}^{II} + B_{411} \right) [P_{nw} + e^{-\Sigma^2_{\text{tot}} k^2 P_w}]$$

(2.19)

where $B[P]$ indicates that the bispectrum should be evaluated using the power spectrum $P$ and we have dropped the counterterms and stochasticity [86]. In this case, the loop corrections become more complex, since the damping factor, $\Sigma^2_{\text{tot}}$ is a function of the redshift-space angles $\mu$. To allow for efficient computation via the FFTLog procedure (Appendix B), we here adopt the isotropic approximation for the one-loop terms, dropping any $\mu$-dependence in $\Sigma^2_{\text{tot}}$ inside the integral. This is expected to be a good approximation in practice, and is exact for the real-space case. Note that we keep the full redshift-dependent damping function in the tree-level expressions, i.e. the isotropic templates are used only for the computations of the one-loop corrections.

### 2.6 Coordinate Rescalings

To compare to observations, we must distill the full bispectrum $B(k_1, k_2, k_3)$ to a lower-dimensional form. In this work, we consider the angle-averaged bispectrum monopole (as in [15, 80, 96]), defined via

$$B_0(k_1, k_2, k_3) = \int_{-1}^{1} \frac{d\mu_1}{2} \int_0^{2\pi} \frac{d\phi}{2\pi} B(k_1, k_2, k_3, \mu_1, \mu_2, \phi)$$

(2.20)

Note that the general expressions for the IR resummed functions in terms of differential operators presented in [86, 87] can be applied to the whole power spectrum, i.e. in principle the procedure of these works does not require a wiggly-smooth decomposition. We prefer to use this decomposition because it results in a significant time gain.
in terms of the angles $\mu$ and $\phi$, with $\mu_1 = \mu$, $\mu_2 = \mu \cos \alpha - \sqrt{1 - \mu^2} \sin \alpha \cos \phi$, for $\cos \alpha \equiv \hat{k}_1 \cdot \hat{k}_2$. In practice, we must consider also the impact of the fiducial cosmology used to convert angles and redshifts into Cartesian coordinates: this modifies (2.20) to

$$B_0(k_1, k_2, k_3) = \frac{1}{\alpha_\parallel^2 \alpha_\perp^2} \int_{-1}^1 \frac{d\mu}{2} \int_0^{2\pi} \frac{d\phi}{2\pi}$$

$$\times B\left[q_1[k_1, \mu_1], q_2[k_2, \mu_2], q_3[k_3, \mu_3], \nu_1[\mu], \nu_2[\mu_2(\mu, \phi)]\right],$$

with $q = k \sqrt{\mu^2/\alpha_\parallel^2 + (1 - \mu^2)/\alpha_\perp^2}$ and $\nu = \mu/\left[\alpha_\parallel \sqrt{\mu^2/\alpha_\parallel^2 + (1 - \mu^2)/\alpha_\perp^2}\right]$ for conventional rescaling parameters $\alpha_\parallel$, $\alpha_\perp$ depending on cosmological parameters [80]. The angular integrals can be performed numerically via Gauss-Legendre quadrature in practice. We further incorporate integration over finite $k$-bins as well as discreteness effects; both effects are discussed below.

3 Practical Implementation

3.1 Loop Integrals

We now discuss how to compute the one-loop bispectrum. The most difficult part of this is evaluating the loop integrals appearing in (2.9); in this work, these are computed via the FFTLog procedure [99], the subtleties of which are described in Appendix B. In essence, the real-space computation proceeds by first writing the integration kernels (products of $\delta$ functions) as a sum over complex polynomials, as an example:

$$B^{(i,j,k)}(x, y),$$

following the above tricks, the bispectrum takes the following schematic form, again taking $B_{222}$ as an example:

$$B_{222}(k_1, k_2, k_3, \mu_1, \mu_2) \sim \sum_{i} \theta_i \sum_{0 \leq j + k \leq 12} \mu^j \chi^k B_{222}^{(i,j,k)}(k_1, x, y),$$

where the $i$ index runs over all combinations of bias and $f(z)$, denoted $\theta_i$. Additionally, we have expanded in terms of the redshift-space angles $\{\mu, \chi \equiv \sqrt{1 - \mu^2} \cos \phi\}$ (of which there are 47 non-trivial combinations); these are related to the $\mu_i$ angles via:

$$\mu_1 \equiv \mu, \quad \mu_2 \equiv \mu \cos \alpha - \chi \sin \alpha, \quad k_3 \mu_3 = -k_1 \mu_1 - k_2 \mu_2.$$
We adopt this basis rather than the more familiar choice of \( \{ \mu_1, \mu_2 \} \), since it avoids pathologies for flattened triangles (whence \( k_1 \approx k_3 + k_3, \) and \( \mu_1 \approx -\mu_2 \)). The underlying shapes, \( B^{(i,j,k)} \), appearing in (3.2) are independent of both redshift-space angles and bias parameters, and depend only on the form of the linear power spectrum, \( k_1, x = k_3^2/k_1^2 \) and \( y = k_3^2/k_1^2 \), assuming \( k_1 \geq k_2 \geq k_3 \).

Two options arise for using the bispectrum templates \( B^{(i,j,k)} \) in Monte Carlo Markov Chain (MCMC) analyses: (a) they may be computed once for a fixed linear power spectrum, (b) they may be computed as a tensor multiplication (cf. 3.1) at each step in the MCMC chain, feeding in the relevant linear power spectrum (and thus \( c_m \) coefficients) each iteration. Whilst (b) is the approach usually adopted for the one-loop power spectrum, we will here adopt (a) for the one-loop bispectrum. This has the effect of fixing cosmology in the bispectrum loops (except for \( \sigma_8 \), which acts as a global rescaling, modulo a small effect concerning the IR resummation amplitude, which we ignore in this work), and is chosen on computational grounds, since the size of the necessary FFTLog matrices becomes very large.\(^5\) Explicitly, we compute the bispectrum templates, \( B^{(i,j,k)} \), for a grid of values of \( \{ x, y, k_1 \} \) (treating flattened triangles with \( \sqrt{x} + \sqrt{y} = 1 \) separately to avoid divergences), then use these to construct a three-dimensional linear interpolator for each shape. The resulting bispectra have been compared to results from explicit (and computationally intensive) numerical integration for a range of values of bias and triangle shapes and found to be in excellent agreement. Full details of the above steps are given in Appendix B.3. We additionally publicly release all our analysis code: this can be found at GitHub.com/OliverPhilcox/OneLoopBispectrum.

### 3.2 Bin Integration

To robustly compare theory and data, we must integrate the model across some set of bins. Following [80], this is achieved via the integral:

\[
\hat{B}_{0, \text{int}} = \frac{V^2}{(2\pi)^6 N_{123}} \prod_{i=1}^{3} \int_{k_i - \Delta k/2}^{k_i + \Delta k/2} dq_i \left( \int_{k_1 - \Delta k/2}^{k_1 + \Delta k/2} dq_j \left( \int_{k_2 - \Delta k/2}^{k_2 + \Delta k/2} dq_k \right) B_0(q_i, q_j, q_k) \right), \tag{3.4}
\]

where \( B_0 \) is the bispectrum monopole of (2.21) and \( N_{123} = 8\pi k_1 k_2 k_3 (\Delta k)^3 V^2/(2\pi)^6 \) for bin center \((k_1, k_2, k_3)\) and width \( \Delta k \). As in [80], this is strictly exact only in the narrow-bin limit, and can be corrected by “discreteness weights” as in the former work. In practice, we compute the set of bispectrum templates \( B^{(i,j,k)}(k_1, x, y) \) for a range of values of \( k_1, x, y \) (see Appendix B.3) then perform the bin-averaging by linearly interpolating these values, dropping any triangles that do not satisfy the triangle conditions \( |q_1 - q_2| \leq q_3 \leq q_1 + q_2 \). The integration is performed using Gauss-Legendre quadrature, as for the angular integrals. Finally, we note that we can perform bin integration either within the MCMC chains or as a pre-processing step (allowing us to use bin-averaged templates in the later analysis). We use the latter option for the purposes of this paper.

\(^5\)To see this, note that the matrix in (3.1) has size \( N_{\text{freq}}^3 \), for \( N_{\text{freq}} \) FFTLog frequencies. Taking \( N_{\text{freq}} = 64 \), with 47 angular combinations, \( O(50) \) bias parameter combinations, and computing the matrix for 10 choices of each of \( x \) and \( y \) (noting that \( k \) scales out), we find \( \sim 5 \times 10^{10} \) elements, or \( \sim 50 \text{ GB} \) in (complex) single precision.
3.3 Free Parameters

Our full model for the one-loop galaxy power spectrum and bispectrum depends on the following 44 free parameters (i.e. Wilson coefficients):

\[
\{ b_1, b_2, b_2^G, b_3, \gamma_2^S, b_{\Gamma_3}, \gamma_{211}, \gamma_{22}, \gamma_{31} \} \times \{ c_0, c_2, c_4, \tilde{c}, \beta_{B,a}, \beta_{B,b}, \beta_{B,c}, \beta_{B,d}, \beta_{B,e}, C_i[i = 1...9] \} \times \{ P_{\text{shot}}, a_0, a_2, B_{\text{shot}}, A_{\text{shot},0}, A_{\text{shot},1}, S_{i = 0...7} \},
\]

where parameters appearing only in the power spectrum (following the definitions of [80]), only in the bispectrum, and in both spectra, are shown in blue, black and purple respectively. The three lines give bias parameters, UV counterterms, and stochasticity parameters respectively. Note that here we switch to the power spectrum biases \( b_{G_2} \) and \( b_{\Gamma_3} \) instead of \( \gamma_2 \) and \( \gamma_{21} \) to ease the comparison with previous works [80, 116]; these are related via (2.5). Whilst performing an MCMC analysis in this high-dimensional space may seem a formidable task, we note that all parameters except \( \{ b_1, b_2, b_2^G \} \) enter the theory model linearly, and can thus be analytically marginalized, following [117]. This is exact, and will be applied to all analyses presented in this work, significantly reducing computational cost. Since the parameter \( b_{\Gamma_3} \) is of physical interest in power spectrum analyses, we opt to marginalize over this explicitly, alongside the quadratic biases.

For the purposes of the analytic marginalization, we assume the following priors on the bispectrum nuisance parameters: all means are zeros, and the expectation values given by 10 for all bias parameters, 10 for all real-space counterterms and one-loop stochastic contributions, 20 for redshift-space counterterms (in order to account for enhancements caused by short-scale non-linear redshift-space distortions, known as fingers-of-God [118]), and 20 for redshift-space one-loop stochastic contributions. For the tree-level stochastic counterterms, following [80] we assume standard deviations of 5 for the dimensionless \( B_{\text{stoch}}, P_{\text{shot}}, \) and \( A_{\text{shot}} \) parameters. The power spectrum nuisance priors match [80, 116]. Note that our nuisance parameters are normalized in such a way that their physical values are expected to be \( \mathcal{O}(1) \) numbers from the naturalness arguments. In this sense our physically-motivated choice of nuisance parameter is conservative, as we allow them to be as large as \( \mathcal{O}(10) \).

3.4 Numerical Results

Before proceeding to use the one-loop bispectra to perform parameter inference, we first consider the form of the spectra themselves. Plotting the bispectrum is a challenge itself: the monopole exists in the three-dimensional simplex of \( \{ k_1, k_2, k_3 \} \), and we have contributions from a wide variety of nuisance parameter combinations. For the purpose of visualization, we will fix the bias parameters to simple local-in-Lagrangian space predictions, based on [119]: assuming the bias to be described only by linear and quadratic terms...
$b_1^L, b_2^L$ in Lagrangian space, this gives
\begin{equation}
\begin{aligned}
    b_1 &= b_1^L + 1, \quad b_2 = b_2^L, \quad b_3 = -3b_2, \quad \gamma_2 = -\frac{2}{7}b_1^L, \quad \gamma_3 = -\frac{1}{9}b_1^L - \gamma_2, \\
    \gamma_{21} &= \frac{2}{21}b_1^L + \frac{6}{7}\gamma_2, \quad \gamma_{2}^x = -\frac{2}{7}b_2, \quad \gamma_{22} = -\frac{6}{339}b_1^L - \frac{9}{49}\gamma_2 \\
    \gamma_{21}^x &= \frac{2}{21}b_2 + \frac{2}{7}\gamma_2, \quad \gamma_{31} = -\frac{4}{11}b_1^L - 6\gamma_2, \quad \gamma_{211} = \frac{5}{77}b_1^L + \frac{15}{14}\gamma_2 + \gamma_{21} - \frac{9}{7}\gamma_3.
\end{aligned}
\end{equation}

In Fig. 1 we plot the deterministic (Eulerian PT) bispectrum contributions assuming the above bias relations with $b_1^L = 1$, $b_2^L = 0.3$ and $f(z) = 0.7$, as well as distortion parameters $\alpha_\parallel = \alpha_\perp = 1$ and the best-fit PT Challenge input power spectrum (cf. §4). For both equilateral and squeezed triangles we observe a similar form: the one-loop corrections are suppressed on large scales (by $k/k_{N\!L}$) but become large as $k$ increases, with the $B_{221}^L$ piece exceeding tree-level theory by $k \sim 0.1 h \text{Mpc}^{-1}$. We find significant cancellation between the various one-loop components (which all depend on the same biases), which is expected from the IR cancellation of loop integrals. Note that the high-$k$ behavior is further modified by the counterterms (scaling as $k^2 P_2^L(k)$) and stochasticity (scaling as $k^3$ and $P_L(k)$ at leading-order). The individual shapes of the bispectrum components are generally non-trivial, with oscillatory signatures seen in $B_{411}$ and, to a lesser extent, $B_{321}^{\ell \Gamma}$. The smooth nature of $B_{222}$ (expected since the three power spectra are all inside the $q$ integral) implies that a smaller number of FFTLog frequencies can likely be used in its computation, which may expedite the template computation, and suggests that this has only weak cosmology dependence. From the deterministic contributions alone, it is clear that the one-loop bispectrum is a significant fraction of $B_{\text{tree}}$ for all $k \gtrsim 0.1 h \text{Mpc}^{-1}$, and thus its inclusion is necessary if we wish to model the bispectrum beyond the softest modes.

### 4 Data and Analysis Details

The dataset used in this paper is the PT Challenge suite [100], comprising high-resolution $N$-body simulations at $z = 0.61$ with a total volume of $566 h^{-3} \text{Gpc}^3$. Galaxies are allocated via a BOSS-like halo occupation prescription, and various summary statistics computed using a fiducial cosmology with $\Omega_m = 0.3$. In all our analyses, we use the redshift-space power spectrum multipoles, $P_L(k)$, and the real-space power spectrum proxy $Q_0$, both of which were studied in detail in [116]. In this work, we additionally add the bispectra in both real- and redshift-space, with the comparison allowing us to assess the relative importance of redshift-space distortions in the one-loop bispectrum.

The relevant bispectra are computed as described in [80], which studied the tree-level bispectrum likelihood. We bin the bispectrum data in wavenumber bins of width $\Delta k = 0.01 h \text{Mpc}^{-1}$, and use only triangles whose bin centers satisfy momentum conservation.\footnote{As shown in [80], the remaining triangles do not carry significant cosmological information.} For $k_{\text{max}} = 0.15, 0.2, 0.3$, we find a total of 372, 825, and 2600 independent triangle configurations, respectively, and note that, unlike [80], we do not include the very first bin in the analysis, i.e. we fix $k_{\text{min}} = 0.01 h \text{Mpc}^{-1}$ for the bispectrum. This matches
the analyses of the actual surveys like BOSS, where the very first bin is often affected by systematics including stellar contamination [2, 15].

Our theory model for the power spectrum matches that of [80, 116], and we make use of the publicly available code class-pt [112] to compute the power spectrum models.\(^7\) Similarly, our theoretical model for the bispectrum is discussed in detail in Section 2, and implemented using the FFTLog prescription using MATHEMATICA – we refer the reader to Appendix B for technical details.

An important part of the likelihoods are the covariance matrices, encoding both errors and correlations. As in previous works, we here adopt the Gaussian tree-level approximation for the analytic covariance matrices of power spectra and bispectra, neglecting any cross-correlation between the two statistics. For sufficiently large scales these assumptions are well justified [24, 63, 65, 80]; at smaller-scales, and in the presence of non-uniform survey geometry, a mock-based approach will probably be needed, such as in [15], most likely in combination with some compression scheme [e.g., 117].

The mock galaxy clustering data from the PT Challenge simulations are analyzed within the Bayesian framework. Here, we perform a global MCMC analysis using the

\(^7\)GitHub.com/michalychforever/CLASS-PT.
publicly available sampler Montepython [120, 121] varying the clustering amplitude $\sigma_8$, $f_{\text{NL}}^{\text{equil}}$ (the amplitude of equilateral primordial non-Gaussianity [96]) and the EFT nuisance parameters. Since the true value of $\sigma_8$ in the simulations remains blinded, we will show results only for the fractional error on $\sigma_8$. As noted above, we will marginalize over all physical nuisance parameters given in §3.3. This is in contrast with some bispectrum studies that aim to fix certain nuisance parameters, such as asserting coevolution relations for Lagrangian biases [81]. Indeed, for some particular purposes, i.e. fits of $\sigma_8$, it may be sufficient to keep fewer parameters in the fit. However, such approximations are unwarranted – their validity can break down for other types of analyses. Therefore, we prefer to explicitly vary all physical nuisance parameters in the fit. By virtue of analytic marginalization [117], this is done at no computational cost.

5 Results: Real-space

We now present results from the above analyses, focusing first on the combination of the redshift-space power spectrum and real-space bispectrum. Though not quite matching observational setups (where the power spectrum and bispectrum are both observed in redshift-space), this analysis will allow us to understand the impact of redshift-space distortions.

To obtain the real space model, we set $f = 0$ in all calculations and retain only EFT operators that do not depend on the LoS angles, giving a one-loop model fully equivalent to that used in [79, 81]. Our data vector contains the power spectrum multipoles, the real-space analog and the bispectrum monopole, i.e. $[P_\ell(k),Q_0(k),B(k_1,k_2,k_3)]$ and we restrict to the $z = 0.61$ snapshot of the PT Challenge simulations. In most analyses we use $P_\ell(k)$ up to $k_{\text{max}} = 0.16h\ Mpc^{-1}$, and $Q_0$ in the range $0.16h\ Mpc^{-1} \leq k < 0.40h\ Mpc^{-1}$, as validated in [80, 100, 116]. We explore the impact of varying the bispectrum $k_{\text{max}}$ below.

5.1 Clustering amplitude and bias parameters

We first focus on measuring the mass clustering amplitude $\sigma_8$ and leading galaxy bias parameters $\{b_1, b_2, b_G, b_\Gamma\}$. These appear both in the one-loop power spectrum and bispectrum models, and hence can be tightly constrained by the data. Unlike $\sigma_8$, the true values of bias parameters in the simulations are unknown. As such, we take their best-fit values at a certain $k_{\text{max}}$ (where the one-loop model can be trusted) as a proxy for their true values. This $k_{\text{max}}$ is measured as in [100] (see also [113]) by determining at what scale cut posteriors for at least one parameter become biased w.r.t. analyses with lower $k_{\text{max}}$. We take best-fit values of bias parameters at the last stable $k_{\text{max}}$ as ground truth, $p_{\text{true}}$. Following this, our parameter measurements are quoted as $\Delta p = p - p_{\text{true}}$, to avoid unblinding the results.

We fit the real-space bispectrum data for the following choices of scale-cut:

$$k_{\text{max}}^B/(h\ Mpc^{-1}) = 0.15, 0.17, 0.19, 0.21, 0.23, 0.25.$$

The resulting posterior contours are shown in Fig. 2, with one-dimensional marginalized limits on the amplitude and some nuisance parameters shown in Tab. 1. From the figure,
we find that the posterior on $\sigma_8$ remains unbiased up to $k_{\text{max}}^{B} = 0.21\ h \text{Mpc}^{-1}$, with a shift of $1 - 2\%$ observed for $k_{\text{max}}^{B} = 0.23\ h \text{Mpc}^{-1}$ and $k_{\text{max}}^{B} = 0.25\ h \text{Mpc}^{-1}$, which becomes significant relative to the PT Challenge error-bars.

However, at $k_{\text{max}}^{B} \geq 0.23\ h \text{Mpc}^{-1}$, we observe that the nuisance parameters become biased w.r.t. measurements at low scale cuts, for example, we find a visible tension between the $b_2$ posterior at $k_{\text{max}}^{B} = 0.23\ h \text{Mpc}^{-1}$ and $k_{\text{max}}^{B} = 0.19\ h \text{Mpc}^{-1}$. In addition, we see that parameters the $b_{p2}$ and $b_{T3}$ become biased. The optimal values of these parameters scale

**Figure 2.** Posterior distributions for the clustering amplitude, $\sigma_8$, and certain nuisance parameters extracted from MCMC analyses of the power spectrum multipoles and the one-loop real-space bispectrum. The power spectrum likelihood is the same for all cases, whilst we vary $k_{\text{max}}^{B}$ for the bispectrum, as indicated in the caption. Corresponding marginalized parameter contours for $k_{\text{max}} = 0.21\ h \text{Mpc}^{-1}$ are given in Tab. 1.
along the degeneracy direction $b_{G_2} + 0.34b_{\Gamma_3} \approx \text{const}$, which closely matches the degeneracy combination imposed by the power spectrum, $b_{G_2} + 0.4b_{\Gamma_3}$ [11].

In contrast with the $k_{\max}^B \geq 0.21 \, h \, \text{Mpc}^{-1}$ picture, the results at all choices of $k_{\max}^B \leq 0.21h \, \text{Mpc}^{-1}$ are fully consistent, implying that $k_{\max}^B = 0.21h \, \text{Mpc}^{-1}$ should be chosen as a baseline scale cut. This is somewhat larger than the one-loop power spectrum scale-cut of $k_{\max} = 0.16h \, \text{Mpc}^{-1}$; whilst this might appear unusual, we note that the power spectrum contains significantly higher signal-to-noise, and is subject to redshift-space complexities, both of which decrease $k_{\max}^P$. We use best-fit values of nuisance parameters from the baseline $P_\ell + Q_0 + B(k_{\max}^B = 0.21h \, \text{Mpc}^{-1})$ analysis as ground truth values in the below.

| Parameter | $P_\ell$ 68% limits |
|-----------|----------------------|
| $\Delta \sigma_8/\sigma_8$ | $0.0080 \pm 0.0064$ |
| $\Delta (b_1\sigma_8)/(b_1\sigma_8)$ | $-0.0063^{+0.0042}_{-0.0038}$ |
| $\Delta b_2$ | $-0.04^{+0.44}_{-0.56}$ |
| $\Delta b_{G_2}$ | $0.08^{+0.25}_{-0.34}$ |
| $\Delta b_{\Gamma_3}$ | $-$ |

| Parameter | $P_\ell+Q_0+B_{\text{real}}$ 68% limits |
|-----------|----------------------|
| $\Delta \sigma_8/\sigma_8$ | $0.0005 \pm 0.0049$ |
| $\Delta (b_1\sigma_8)/(b_1\sigma_8)$ | $-0.0003 \pm 0.0013$ |
| $\Delta b_2$ | $-0.005 \pm 0.037$ |
| $\Delta b_{G_2}$ | $-0.003 \pm 0.017$ |
| $\Delta b_{\Gamma_3}$ | $0.005 \pm 0.050$ |

**Table 1.** One-dimensional marginalized contraints on low-order bias parameters and the clustering amplitude $\sigma_8$ extracted from the PT Challenge data-set. We display results obtained using only the power spectrum multipoles $P_\ell$ (left panel, cf. [100]), and those including the power spectrum, $Q_0$ and the one-loop real-space bispectrum likelihood with $k_{\max}^B = 0.21h \, \text{Mpc}^{-1}$ (right panel). The one-loop bispectrum is the main new feature of this work. Most parameters are normalized to their true values, to avoid unblinding the simulation. In real-space, the addition of the bispectrum significantly tightens posteriors on bias parameters (by at least an order of magnitude), and gives $\approx 20\%$ improvement on $\sigma_8$. Further details are given in the main text, with corresponding results for the redshift-space bispectrum shown in Tab. 2.

It is instructive to compare the parameter constraints extracted using the one-loop bispectrum to those from the power spectrum multipoles alone, i.e. $P_\ell(k)$ at the baseline $k_{\max} = 0.16h \, \text{Mpc}^{-1}$. These are shown in the left panel of Tab. 1. We find an improvement of $31\%$ in $\sigma_8$, whilst the error-bars on bias parameters tighten by an order of magnitude in some cases. Despite the noticeable increase in signal-to-noise of the data-set, we find a modest improvement in $\sigma_8$: this is linked to the proliferation of bias, counterterm, and stochasticity parameters needed to describe the one-loop bispectrum in an unbiased manner.

It is also useful to compare our one-loop bispectrum results with those from the tree-level bispectrum. We cannot directly use results from [80] since the former work also varied other cosmological parameters such as $H_0$ and $\Omega_m$. To obtain a cleaner comparison, we repeat the tree-level analysis of [80] with the same analysis settings as here, using $k_{\max}^B = 0.08h \, \text{Mpc}^{-1}$ for the tree-level bispectrum. We find $\Delta \sigma_8/\sigma_8 = 0.002 \pm 0.0053$, i.e. an $21\%$ improvement over the power spectrum only result. Comparing this the present
analysis, we see that the addition of the one-loop bispectrum likelihood yields an extra 10% improvement over the tree-level bispectrum likelihood.

Finally, we compare our results with those of [81]. Unlike our work, [81] used the power spectrum of halos and galaxies in real space, leading to the notorious $b_1 - \sigma_8$ degeneracy being largely unbroken. This explains why our results on $\sigma_8$ are much better – most of the constraining power comes from redshift-space distortions omitted in [81]. Despite this difference, our analysis does confirm a general trend pointed out in [81] – the returns from the one-loop bispectrum are limited by the large number of nuisance parameters. As such, it will be important to obtain better priors on them in the future, for example using hydrodynamical simulations.

5.2 Primordial non-Gaussianity

It is interesting to study to what extent the one-loop bispectrum model can help improve constraints on primordial non-Gaussianity (PNG), following constraints from the tree-level bispectrum in [96, 97]. We consider here the case of equilateral PNG, which induces the following three-point correlation of the linear density field (see [96] for further details),

$$\langle \delta^{(1)}_{k_1} \delta^{(1)}_{k_2} \delta^{(1)}_{k_3} \rangle = (2\pi)^3 \delta^{(3)}_D(\mathbf{k}_{123}) B_{\text{equil}}(k_1, k_2, k_3),$$

$$B_{\text{equil}} = \prod_{a=1}^3 T(k_a) \left[ 18 J_{\text{NL}} \Delta_\zeta S_{\text{equil}}(k_1, k_2, k_3) \right] \frac{1}{k_1^2 k_2^2 k_3^2},$$  

where $\zeta$ is the primordial curvature fluctuation with dimensionless amplitude $\Delta_\zeta$, and we have introduced the transfer function $T(k) = (P_{11}(k)/P_\zeta(k))^{1/2}$.

Non-Gaussianity in the initial conditions generates three main effects [96]: (1) an additional contribution $B_{111}$ to the tree-level bispectrum, (2) an extra one-loop power spectrum correction $P_{12}$, and (3) further contributions in the galaxy bias expansion, which modifies the tree-level expressions by introducing the so-called “scale-dependent” bias. The latter stems from the following expression,

$$\delta_g \subset b_\zeta \nabla^2 \zeta + ...,$$

where “…” denote non-linear PNG corrections which can be ignored for the purposes of this paper. The term $\nabla^2 \zeta$ generates tree-level “scale-dependent” bias corrections to the power spectrum. Note that these corrections are suppressed in the equilateral case w.r.t. the case of local PNG, where the scale-dependent bias is a leading effect on the galaxy power spectrum, and thus the power spectrum dominates the constraining power on $f_{\text{NL}}$.

As shown in [96], for the one-loop power spectrum and tree-level bispectrum, we must include all three of the $P_{12}$, $B_{111}$ and $b_\zeta$-related terms in our model. In this paper, we consider the one-loop bispectrum, which technically requires additional non-linear $f_{\text{NL}}$ corrections to the galaxy bispectrum, such as $B_{113}^{(I)}, B_{113}^{(II)}, B_{122}^{(I)}, B_{133}^{(II)}$ [110, 122]. However, given that $f_{\text{NL}}^\alpha \Delta_\zeta$ is small, these will be suppressed, thus we leave their systematic calculation for future work, focussing only on the leading terms, similar to [96, 98].
$f_{\text{equil}}^{\text{NL}} \lesssim 500$, the next-to-leading order contributions are subdominant to two-loop matter corrections for $k \lesssim 0.2h\text{Mpc}^{-1}$.

Concerning scale-cuts, we find that use of the $P_t$ and $Q_0$ statistics at high $k_{\text{max}}$ can lead to biases in the recovered values of $f_{\text{equil}}^{\text{NL}}$. This is consistent with the estimates of [96], which showed that the two-loop corrections can actually be larger than the non-Gaussian $P_{12}$ contribution at small scales. Thus, we choose $k_{\text{max}} = 0.2h\text{Mpc}^{-1}$ for the $Q_0$ statistics and $k_{\text{max}} = 0.14h\text{Mpc}^{-1}$ for $P_t$ in the PNG analysis of this section. For $B_{\text{real}}$ we use the baseline data cut $k_{\text{max}}^B = 0.21h\text{Mpc}^{-1}$, motivated by the discussion above.

To perform the analyses including PNG, we fit the parameter $f_{\text{equil}}^{\text{NL}}$ in addition to $\sigma_8$ and nuisance parameters. Since the PT challenge simulations were run using purely Gaussian initial conditions, we expect to find $f_{\text{equil}}^{\text{NL}}$ consistent with zero. Indeed, our nominal constraint on the amplitude of the equilateral shape is given by

$$1\text{-loop } (k_{\text{max}}^B = 0.21h\text{Mpc}^{-1}): \quad f_{\text{equil}}^{\text{NL}} = 72 \pm 80 \text{ (95\% CL)}.$$  \hfill (5.3)

We stress that these results are obtained without any external priors on $\sigma_8$ or the non-linear bias coefficients. This constraint can be compared with that obtained from the tree-level real-space bispectrum likelihood,

$$\text{tree } (k_{\text{max}}^B = 0.08h\text{Mpc}^{-1}): \quad f_{\text{equil}}^{\text{NL}} = 53 \pm 114 \text{ (95\% CL)}.$$  \hfill (5.4)

At face value, our results imply that the addition of the one-loop bispectrum can improve constraints on $f_{\text{equil}}^{\text{NL}}$ by $\sim 30\%$.

### 6 Results: Redshift-Space

In this section we present the analysis of the data-vector $[P_t(k), Q_0(k), B_0(k_1, k_2, k_3)]$, where all statistics are in redshift space and include projection and coordinate-distortion effects. This set-up thus fully matches an analysis of a realistic galaxy survey such as BOSS [2]. As for the power spectrum, we expect that the addition of redshift-space distortions (particularly the fingers-of-God effect [118], hereafter FoG), will reduce the non-linear scale, thus it is likely that $k_{\text{max}}^B$, and the constraining power of the bispectrum monopole, will decrease.

#### 6.1 Clustering amplitude and bias parameters

Let us discuss the recovery of the mass clustering amplitude $\sigma_8$ and bias parameters. As a point of comparison, we fix the fiducial bias parameters to those extracted from the real space analysis before with $k_{\text{max}}^B = 0.21h\text{Mpc}^{-1}$. We fit the redshift-space bispectrum data for the following choices of scale-cut:

$$k_{\text{max}}^B/(h\text{Mpc}^{-1}) = 0.15, \ 0.17, \ 0.20, \ 0.22.$$  \hfill (6.1)

The resulting posterior contours are shown in Fig. 3, and we observe that all parameters of interest are unbiased at $k_{\text{max}}^B = 0.15h\text{Mpc}^{-1}$. For comparison, we also show the baseline real space bispectrum results of §5. Note that if we use $k_{\text{max}}^B < 0.15h\text{Mpc}^{-1}$,
Figure 3. Posterior distributions of the clustering amplitude and certain nuisance parameters obtained from MCMC analyses of the power spectra and one-loop redshift-space bispectrum monopole $B_0$. The power spectrum likelihood is the same for all cases. We show results for different values of the bispectrum data cut $k_{B_{\text{max}}}^B$, as indicated by the caption. This is analogous to Fig. 2 (whose optimal constraint is shown by the purple curve), but utilizes the redshift-space bispectrum. Corresponding marginalized parameter constraints with $k_{B_{\text{max}}}^B = 0.15\, h\text{Mpc}^{-1}$ are given in Tab. 2.

The bispectrum data is not sufficient to constrain all the nuisance parameters entering the theory model. This gives rise to significant marginalization projection effects, which can be naively interpreted as a bias in our model. We study these effects in Section E and show that the measurements at $k_{B_{\text{max}}}^B < 0.15\, h\text{Mpc}^{-1}$ are consistent with our baseline choice $k_{B_{\text{max}}}^B = 0.15\, h\text{Mpc}^{-1}$ once projection effects are taken into account.
Table 2. One-dimensional marginalized constraints on low-order bias parameters and the clustering amplitude $\sigma_8$ extracted from the PT Challenge data-set. We show the fit from the combined likelihood including power spectrum multipoles, $Q_0$, the tree-level redshift-space bispectrum at $k^B_{\text{max}} = 0.06h\text{Mpc}^{-1}$ (left table) and the one-loop redshift-space bispectrum at $k^B_{\text{max}} = 0.15h\text{Mpc}^{-1}$ (right table). The parameters are normalized relative to their true values. Whilst we find significant enhancements in the bias parameter constraints compared to the power spectrum alone (cf. Tab. 1), the constraint on $\sigma_8$ does not improve appreciably.

At $k^B_{\text{max}} = 0.17h\text{Mpc}^{-1}$ the clustering amplitude $\sigma_8$, the rescaled linear bias $b_1\sigma_8$, $b_2$ and $b_{\Gamma_3}$ become biased w.r.t. their optimal values coming from the real-space bispectrum analysis. These biases are accompanied with a significant increase in the $\chi^2$ statistics. Thus, we conclude that the two-loop bispectrum corrections are not negligible at this scale. This is further supported by the bias growing with $k^B_{\text{max}}$: in particular, at $k^B_{\text{max}} = 0.22h\text{Mpc}^{-1}$ the bias on $\sigma_8$ reaches 2%, which is significant in the context of the PT Challenge simulation volume. In conclusion, we find that the one-loop galaxy bispectrum model in redshift space works well up to $k^B_{\text{max}} = 0.15h\text{Mpc}^{-1}$ for the precision that corresponds to the total volume of the PT Challenge simulation (which we recall is significantly larger than current and forthcoming datasets). The optimal values of cosmological and bias parameters for this choice are presented in Tab. 2. For comparison, we also show the results from the tree-level bispectrum analysis akin to [80].\footnote{For this analysis we use $k^B_{\text{max}} = 0.06h\text{Mpc}^{-1}$ and set the $c_1$ counterterm to zero for consistency with our one-loop model. This term served as a proxy for the one-loop corrections thus its inclusion is equivalent to partially accounting for one-loop information. To clearly compare the tree-level and one-loop results, we exclude it here.} That $k^B_{\text{max}}$ is lower in redshift-space than real-space is no surprise: this indicates that the characteristic scale of FoG effects ($\sigma_{\text{FoG}}$) is smaller than that of non-linearities ($k^{-1}_{\text{NL}}$). For the power spectrum in redshift-space, higher-order counterterms were important to model FoG, scaling as $k^4$. An analogous set of nuisance parameters may be included here, but we caution that their number is large due to the higher dimensionality of the bispectrum.

Considering the marginalized posteriors directly, we find that the one-loop bispectrum likelihood (at $k^B_{\text{max}} = 0.15h\text{Mpc}^{-1}$) yields only a 12% improvement on $\sigma_8$ compared to the power spectrum alone, though the constraints on bias parameters (and thus astrophysics) improve markedly. Comparing this with the tree-level case, we see that the inclusion of one-loop corrections actually lead to a somewhat worse result than for tree-level bispectrum,
which tightens the \( \sigma_8 \) constraint by 18\% for the analysis settings adopted in this work. This is consistent with previous studies considering the real-space bispectrum \([81]\), and arises primarily due to the large number of nuisance parameters appearing in the one-loop calculation, especially in redshift-space. A similar situation takes place in the context of the one-loop redshift-space power spectrum, whose information content is limited by marginalization over nuisance parameters \([24]\). We will discuss this issue in detail later.

7 Implications for the BOSS survey

In this section we estimate the potential performance of the one-loop bispectrum model applied to the BOSS survey data \([2]\), which is the largest publicly-available spectroscopic galaxy clustering dataset. This survey has significantly smaller volume than our mock
simulation data, so one can expect that the analysis can be pushed to smaller scales [16, 80]. Indeed, the relevant parameter in this problem is the ratio of the theory systematic bias in a certain parameter to the statistical error on that parameter. For the BOSS volume the statistical errors are significantly larger than the PT Challenge simulation volume, due to a ratio of volumes of \( \approx 100 \). As the theoretical errors do not depend on the volume, the ratio between the theoretical error and statistical errors thus becomes smaller, and hence any residual theoretical systematics becomes less sizable in relative terms.

To demonstrate this, we repeat the likelihood analysis above for the redshift-space data vector \([P_\ell, Q_0, B_0]\), but rescale the covariance to match the BOSS volume \(V_{\text{BOSS}} = \frac{V_{\text{PT Challenge}}}{100} \approx 6 \, (h^{-1}\text{Gpc})^3\). We select \(k_{\text{max}} = 0.20 h \, \text{Mpc}^{-1}\) for power spectrum multipoles \(P_\ell\) and \(k_{\text{max}}^B = 0.20 h \, \text{Mpc}^{-1}\) for the bispectrum monopole \(B_0\); significantly larger than that found in §6. The results of this analysis are presented in Fig. 4 and in Tab. 3, showing also the \(P_\ell + Q_0\) and \(P_\ell + Q_0 + B_0^{\text{tree}}\) \((k_{\text{max}}^B = 0.08 h \, \text{Mpc}^{-1}\) following [80]) analyses for comparison.

| \(P_\ell + Q_0\), BOSS-like | \(P_\ell + Q_0 + B_0^{\text{tree}}\), BOSS-like |
|----------------------------|---------------------------------------------|
| Parameter | \(68\%\) limits | Parameter | \(68\%\) limits |
| \(\Delta \sigma_8/\sigma_8\) | \(0.019 \pm 0.042\) | \(\Delta \sigma_8/\sigma_8\) | \(-0.023^{+0.035}_{-0.040}\) |
| \(\Delta (b_1 \sigma_8)/(b_1 \sigma_8)\) | \(-0.013 \pm 0.013\) | \(\Delta (b_1 \sigma_8)/(b_1 \sigma_8)\) | \(0.0021 \pm 0.0089\) |
| \(\Delta b_2\) | \(-0.37^{+0.50}_{-0.68}\) | \(\Delta b_2\) | \(-0.02^{+0.34}_{-0.40}\) |
| \(\Delta b_{g_2}\) | \(0.37 \pm 0.33\) | \(\Delta b_{g_2}\) | \(0.07 \pm 0.19\) |

Table 3. One-dimensional marginalized constraints on low-order bias parameters and the clustering amplitude \(\sigma_8\) extracted from the PT Challenge dataset, with covariance adjusted to match the volume of the BOSS survey. We show results for the \(P_\ell + Q_0\) only analysis (top left), the tree-level \(P_\ell + Q_0 + B_0\) likelihood (top right), and the one-loop \(P_\ell + Q_0 + B_0\) likelihood (bottom). The inclusion of the bispectrum sharpens constraints on \(\sigma_8\) by \(\approx 24\%\), with some \(\approx 10\%\) improvement arising from the addition of the one-loop contributions.

We see that in the context of BOSS, the addition of the one-loop bispectrum yields an \(\approx 24\%\) improvement over the power spectrum-only result and an \(\approx 10\%\) improvement over the tree-level bispectrum likelihood result. However, this leads to a noticeable shift in nuisance parameters, with \(b_{g_2}\) approximately \(1.7\sigma\) from its fiducial value. This could simply be a prior-volume effect however (since the effect of the priors becomes more important at lower simulation volume), especially given that \(b_{g_2}\) departs from its fiducial value at \(1.1\sigma\) already for the power-spectrum alone. The tree-level bispectrum analysis, however, results in an unbiased recovery of all nuisance and cosmological parameters.
It is also instructive to study whether the one-loop bispectrum can improve constraints on equilateral PNG. Incorporating this parameter in the analysis as before (varying both $f_{\text{equil}}^{\text{NL}}$ and $\sigma_8$) we find $f_{\text{equil}}^{\text{NL}} = 197 \pm 350$. For the clustering amplitude we find $\Delta \sigma_8/\sigma_8 = -0.026\pm0.035$, with a slight 0.6σ shift w.r.t. the ground truth. For comparison, we have also run an analysis using the tree-level bispectrum at $k_{\text{max}}^B = 0.08\, h\, \text{Mpc}^{-1}$ instead of the one-loop bispectrum, and found $f_{\text{equil}}^{\text{NL}} = 420\pm440$, $\Delta \sigma_8/\sigma_8 = -0.025\pm0.040$. First, we see some bias in $f_{\text{equil}}^{\text{NL}}$, which can be attributed to prior volume effects and somewhat more optimistic data cuts for the power spectrum that we use in our analysis here. Indeed, in [96] it was shown that the tree-level model yields unbiased results on $f_{\text{equil}}^{\text{NL}}$ for $k_{\text{max}}^B = 0.08\, h\, \text{Mpc}^{-1}$ and $k_{\text{max}}^{\text{PT}} = 0.17\, h\, \text{Mpc}^{-1}$. Second, we notice that the bound on $f_{\text{equil}}^{\text{NL}}$ in the one-loop case is 30% better than that of the tree-level analysis. The improvement is quite modest as a consequence of the fact that the one-loop model introduces many nuisance parameters, which cannot be constrained by the data. In our analysis we use highly conservative but still physically-motivated priors; if more aggressive priors on nuisance parameters are used, the constraints are likely to improve further.

In conclusion, we note that the addition of the one-loop bispectrum may yield some ≈ 30% improvement on the amplitude of equilateral PNG. We stress, however, that this comes with two important caveats. First, the $k_{\text{max}}^B$ used for this study (0.2\,h\,\text{Mpc}^{-1}) results in noticeable biases on the nuisance parameters, suggesting that the errorbar on $f_{\text{equil}}^{\text{NL}}$ may be underestimated due to over-fitting. Whether this induces a bias on $f_{\text{equil}}^{\text{NL}}$ is unclear; such an error would likely show up only in the analysis of simulations containing $f_{\text{equil}}^{\text{NL}} \neq 0$. Secondly, we have neglected the PNG-induced one-loop corrections to the bispectrum (as in [98]), which can be marginally important for the scales of interest (particularly in the tails of the $f_{\text{equil}}^{\text{NL}}$ posterior), as can be easily estimated with the scaling universe approximation outlined in [122].

Finally, we note that our analysis indicates a more modest improvement on $f_{\text{equil}}^{\text{NL}}$ than that reported in [98]. [98] suggest that the one-loop bispectrum improves $f_{\text{equil}}^{\text{NL}}$ constraints over the tree-level result by a factor of few. This follows from a comparison with the tree-level bispectrum analysis of [96]. This comparison is misleading, however, since the baseline analysis of [96] varies cosmology whilst [98] always keeps cosmological parameters fixed. We have checked that this accounts for most of the difference between [96] and [98]. A more detailed comparison with [98] is not currently possible because the former work has not yet presented sufficient details about their analysis and theory model. It will be interesting to compare our results with their analysis in the future.

8 Conclusions and Discussion

In this work, we have presented and validated a complete calculation of the galaxy bispectrum monopole in redshift space at one-loop order in effective field theory. Our model includes one-loop corrections due to mode coupling, as well as the full set of EFT counterterms that are needed to regulate the UV behavior of loop integrals and capture the physical effects of backreaction of short scales onto the large-scale modes. Furthermore, we incorporate a bias expansion up to fourth order (noting that many operators van-
ish after renormalization of the power spectrum and bispectrum) as well as fourth order redshift-space distortions. In addition, our calculation includes IR resummation to capture the non-linear evolution of baryon acoustic oscillations (both for the power spectrum and bispectrum), as well as projection and binning effects. In short, we include all relevant ingredients needed to compare theory with observational galaxy clustering data.

We have studied the performance of the one-loop bispectrum model in terms of cosmological parameter constraints, focusing primarily on the mass fluctuation amplitude $\sigma_8$. To validate our model we use the PT Challenge simulation suite [100], which are equivalent to a BOSS-like survey with a hundred times larger volume, thus allowing for high-precision tests. We analyze a data vector that consists of the standard redshift-space power spectrum multipoles, the real-space power spectrum proxy $Q_0$ [116], and the redshift-space bispectrum monopole. In this setup, we have found that the inclusion of the one-loop corrections allows us to extend the agreement between bispectrum theory and data up to $k_{\text{max}} = 0.15 h \, \text{Mpc}^{-1}$, or $k_{\text{max}} = 0.21 h \, \text{Mpc}^{-1}$ in real-space. This can be contrasted with the tree-level model bispectrum model, which works only up to $k_{\text{max}} = 0.08 h \, \text{Mpc}^{-1}$ [80].

We caution that these scale-cuts depend on both the survey volume and galaxy type: for BOSS, we can use $k_{\text{max}} = 0.20 h \, \text{Mpc}^{-1}$, and it is likely that the wavenumber reach is larger for emission line galaxies, which boast smaller FoG effects [12, 64]. Further, one might hope to extend the $k$-reach by specializing to some real-space bispectrum analog (similar to $Q_0$) at high-$k$: this will be considered in future work.

Despite a significant extension of the $k$-space reach, we have not found the bispectrum to lead to noticeable improvements in the $\sigma_8$ constraints compared to those with obtained from tree-level theory when applied to the PT Challenge simulations. This is a consequence of the large number of the EFT nuisance parameters that appear in the one-loop calculation (particularly in redshift-space), and must be marginalized over in our analysis. For a BOSS-volume survey (and accompanying systematic error thresholds), we find greater utility, with the one-loop bispectrum improving constraints by $\sim 10\%$ over the tree-level case, though it remains to be seen whether any accompanying shifts in nuisance parameters are real (and malignant) or just prior volume effects.

We have additionally studied whether the one-loop bispectrum can help constrain equilateral primordial non-Gaussianity (and thus single-field inflation), finding that, for the BOSS survey, the one-loop bispectrum may improve constraints on the non-Gaussianity parameter $f_{\text{NL}}^{\text{equil}}$ by $\approx 30\%$ compared to the tree-level theory. Achieving this, however, requires pushing the bispectrum analysis to $k_{\text{max}}^{B} = 0.2 h \, \text{Mpc}^{-1}$, where the shifts in the bias parameters become evident. It remains to be seen if this problem can be alleviated with better priors on nuisance parameters or with one-loop PNG-induced corrections to the bispectrum, which were omitted in this study. If one is interested in astrophysics, the one-loop bispectrum is much more useful: we find a significant tightening in the posteriors of parameters such as linear and tidal bias compared to those with only tree-level theory.

An important conclusion from our study is that we need better knowledge of the EFT nuisance parameters if we wish to extract more cosmological information from the bispectrum. This can be done in several ways. First, one can include data from the higher-order angular moments of the redshift-space bispectrum [37]. Since these moments...
depend on the same set of parameters, their inclusion should tighten the EFT nuisance parameters posteriors, aiding determination of the cosmological parameters of interest. Second, one can constrain the EFT nuisance parameters with higher order statistics, such as the trispectrum, see e.g. [102, 123] for work in this direction. Finally, one can obtain better priors on the extra nuisance parameters using high fidelity N-body or hydrodynamical simulations. A powerful route by which to achieve this involves EFT field level techniques, see e.g. [85, 124–132]. We plan to investigate these options in future work.

Though the one-loop bispectrum analysis of this work was limited only to two cosmological parameters, the mass fluctuation amplitude and the equilateral non–Gaussianity parameter, it may be similarly extended to other parameters such as local primordial non-Gaussianity, or the neutrino mass. The improvement on other parameters, especially those beyond the minimal $\Lambda$CDM model, could be significantly larger, particularly when some new feature is introduced that is not degenerate with the smooth loop corrections. If the parameter of interest enters the theoretical model linearly, the analysis can proceed as above; if this is not the case, one would require an optimization of our one-loop bispectrum pipeline, since the FFTLog-based approach does not currently allow for a fast re-calculation of the theoretical template as the power spectrum is varied. If only the $\alpha_\parallel$, $\alpha_\perp$ parameters are varied however, the templates do not need to be recomputed, only rebinned (via 2.21). Analyses including such effects will be natural next steps in our research program.

Finally, we note that the bispectrum data offers novel probes of new physics. In particular, constructing the bispectrum from different tracers will allow one to probe the equivalence principle [53–59]. Such an analysis is complicated if one considers only the power spectrum since the effects sensitive to the equivalence principle appear there only at the one-loop order. In contrast, the cross-bispectrum of different kinds of tracers can be a sensible probe of the equivalence principle, whose violation would generate new bispectrum shapes that are not present in the $\Lambda$CDM model. This, in particular, will help one derive new constraints on the violation of Lorentz symmetry in the dark matter sector [133, 134]. We leave this and other tests of new physics with the bispectrum for future work.

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At fourth-order, the real-space perturbation theory kernels (cf. 2.6) are given by

\[ K_1(q_1) = b_1, \]
\[ K_2(q_1, q_2) = \{b_1 F_2(q_1, q_2)\} + \left\{ \frac{b_2}{2} + \frac{\gamma_2}{2} \kappa(q_1, q_2) \right\}, \]
\[ K_3(q_1, q_2, q_3) = \{b_1 F_3(q_1, q_2, q_3)\} + \left\{ b_2 F_2(q_1, q_2) + 2\gamma_2 \kappa(q_1, q_23) G_2(q_2, q_3) \right\} + \left\{ \frac{b_3}{6} + \frac{\gamma_2}{2} \kappa(q_1, q_2) + \gamma_3 L(q_1, q_2, q_3) + \gamma_21 \kappa(q_1, q_23) \kappa(q_2, q_3) \right\}, \]
\[ K_4(q_1, q_2, q_3, q_4) = \{b_1 F_4(q_1, q_2, q_3, q_4)\} \]
\[ + \left\{ \frac{b_2}{2} \left[ F_2(q_1, q_2) F_2(q_3, q_4) + 2 F_3(q_1, q_2, q_3) \right] \right\} + \gamma_2 \left[ \kappa(q_12, q_34) G_2(q_1, q_2) G_2(q_2, q_3) + 2 \kappa(q_12, q_4) G_3(q_1, q_2, q_3) \right] \]
\[ + \left\{ \frac{b_3}{2} F_2(q_1, q_2) + \frac{\gamma_2}{2} \left[ 2 \kappa(q_12, q_3) G_2(q_1, q_2) + \kappa(q_3, q_4) F_2(q_1, q_2) \right] \right\} \]
\[ + 3 \gamma_3 L(q_1, q_2, q_34) G_2(q_34) \]
\[ + \gamma_21 \left[ \kappa(q_12, q_34) \kappa(q_1, q_2) F_2(q_34, q_4) + 2 \kappa(q_123, q_4) \kappa(q_12, q_3) F_2(q_1, q_2) \right] \]
\[ + \left\{ \gamma_21 \kappa(q_1, q_23) \kappa(q_2, q_3) + \gamma_21 L(q_1, q_2, q_34) \kappa(q_3, q_4) \right\} \]
\[ + \gamma_22 \kappa(q_12, q_34) \kappa(q_1, q_2) \kappa(q_3, q_4) \]
\[ + \gamma_31 \left\{ \frac{1}{18} \kappa(q_1, q_34) \left( \frac{15}{4} \kappa(q_23, q_4) \kappa(q_2, q_3) - L(q_2, q_3, q_4) \right) \right\} \]
\[ + \frac{1}{14} \left( M(q_1, q_23, q_4, q_234) - M(q_1, q_234, q_3, q_4) \right) \kappa(q_2, q_3) \right\}, \]

where the bias parameters are shown in color, \( q_i \ldots j \equiv q_i + \cdots + q_j \) and we define the angles

\[ \kappa(q_1, q_2) = (q_1 \cdot q_2)^2 - 1, \quad (A.2) \]
\[ L(q_1, q_2, q_3) = \left[ (q_1 \cdot q_2)(q_2 \cdot q_3)(q_3 \cdot q_1) - (q_1 \cdot q_2)^2 - (q_2 \cdot q_3)^2 - (q_3 \cdot q_1)^2 + 1, \right. \]
\[ M(q_1, q_2, q_3, q_4) = \left. (q_1 \cdot q_2)(q_2 \cdot q_3)(q_3 \cdot q_4)(q_4 \cdot q_1) \right]. \]

(A.1) uses the standard Eulerian density and velocity kernels \( (F_n \text{ and } G_n) \) and we note that the kernels have not been symmetrized over their arguments. We additionally note that, despite appearing in \( K_1 \), \( b_3 \) does not contribute to the one-loop bispectrum after averaging over internal angles and performing bias renormalization. Furthermore, we drop any bias operators in \( K_4 \) that do not appear in the one-loop bispectrum, thus the above expression will not be sufficient for computation of the two-loop power spectrum.
A.2 Redshift-Space

The redshift-space kernels are obtained by expanding the RSD mapping of (2.7) and expanding all fields in terms of the linear density field. Following a lengthy computation, we find the following forms in terms of the real-space kernels:

\[
Z_1(q_1) = K_1 + f \mu_2^2,
\]

\[
Z_2(q_1, q_2) = K_2(q_1, q_2) + f \mu_2^2 G_2(q_1, q_2) + \frac{f \mu_1 q_1 \mu_2}{2} K_1 \left[ \frac{\mu_1}{q_1} + \frac{\mu_2}{q_2} \right] + \frac{(f \mu_1 q_1 \mu_2)^2}{2} \frac{\mu_1 \mu_2}{q_1 q_2},
\]

\[
Z_3(q_1, q_2, q_3) = K_3(q_1, q_2, q_3) + f \mu_2^2 G_3(q_1, q_2, q_3)
\]

\[
\begin{aligned}
&+ (f \mu_1 q_1 q_2) \left[ \frac{\mu_1}{q_1^2} K_1 G_2(q_1, q_2) + \frac{\mu_3}{q_3} K_2(q_1, q_2) \right] \\
&+ \frac{(f \mu_1 q_1 q_2)^2}{2} \left[ 2 \frac{\mu_1^2}{q_1^3} G_2(q_1, q_2) + \frac{\mu_1 \mu_2}{q_1 q_2} K_1 \right]
\end{aligned}
\]

\[
Z_4(q_1, q_2, q_3, q_4) = K_4(q_1, q_2, q_3, q_4) + f \mu_2^2 G_4(q_1, q_2, q_3, q_4)
\]

\[
\begin{aligned}
&+ (f \mu_1 q_1 q_2 q_3) \left[ \frac{\mu_1}{q_1^2} K_1 G_3(q_1, q_2, q_3) + \frac{\mu_4}{q_4} K_3(q_1, q_2, q_3) \right] \\
&+ \frac{(f \mu_1 q_1 q_2 q_3)^2}{2} \left[ 2 \frac{\mu_1^2}{q_1^3} G_3(q_1, q_2, q_3) + \frac{\mu_1 \mu_2}{q_1 q_2} G_2(q_1, q_2, q_3) \right]
\end{aligned}
\]

writing \( \mu_{i...j} \equiv \mu_{q_i + ... + q_j} \) and dropping the argument of \( K_1 \) for clarity.

B Computation of the One-Loop Bispectrum with FFTLog

In this appendix, we discuss practical computation of the loop integrals given in (2.9). Before considering the redshift-space case, we will first examine how to compute the real-space integrals, which follow a similar logic, but are significantly simpler. Our approach follows [99], but is extended to the case of biased tracers and redshift-space.

B.1 Real-Space

B.1.1 Formalism

As noted in §2, the first step in the bispectrum computation is the expansion of the perturbation theory kernels (Appendix A) as polynomials in \( q^2, |k_1 - q|^2 \) and \( |k_2 + q|^2 \) or their reciprocals (utilizing permutation symmetries). In practice, this results in a sum over many thousands of terms, once the relevant symmetries have been imposed, and is automated.
using mathematica. For \( B_{222} \), each term is proportional to
\[
\int P_L(q) P_L(|k_1 - q|) P_L(|k_2 + q|),
\]
(B.1)
for integer \( \alpha_i \), with a similar form found for the other loop integrals except with fewer factors of \( P_L \). Expanding the linear power spectrum as a sum of complex polynomials in \( k \), i.e. \( P_L(k) = \sum_m c_m k^{\nu + i \eta_m} \) for frequencies \( \eta_m \), coefficients \( c_m \), and (real) FFTLog bias \( \nu \) (which sets the eventual integral convergence properties), we can rewrite (B.1) in the form
\[
\sum_{m_1m_2m_3} c_{m_1}c_{m_2}c_{m_3} \int \frac{1}{q^{2\nu_1} |k_1 - q|^{2\nu_2} |k_2 + q|^{2\nu_3}} \]
(B.2)
for \( 2\nu_j = \alpha_j - \nu - j \eta_m \), where all the cosmology dependence (encoded in \( c_m \)) is now outside the integral. The remaining integral can be computed using path integral methods as
\[
\int \frac{1}{q^{2\nu_1} |k_1 - q|^{2\nu_2} |k_2 + q|^{2\nu_3}} \equiv k^{3-2(\nu_1+\nu_2+\nu_3)} J(\nu_1, \nu_2, \nu_3; x, y)
\]
(B.3)
where \( x^2 = k_3/k_1, y^2 = k_2/k_1 \) and \( J \) (with complex arguments \( \nu_j \)) can be expressed as a sum of hypergeometric functions and Gamma functions [99]. This reduces the computation of bispectrum templates to a set of matrix multiplications and function evaluations, as noted in §3. For \( B_{321}^I \) we find a similar form to (3.1), except with rank-two matrices, whilst \( B_{321}^{II} \) and \( B_{111} \) involve only a one-dimensional sum (and one set of \( c_m \) coefficients).

### B.1.2 Limiting Behavior

When computing spectra via FFTLog, it is important to verify whether the relevant loop integrals actually converge. This is achieved by taking the UV and IR limits of the integration kernels and assessing the dependence on the \( \nu_i \) parameters appearing in (B.2). As an example, we consider the contribution of three \( \delta \) operators to \( B_{222} \) (involving three copies of \( b_1 F_2(k - q, q) \)). This has the following limits in the equilateral configuration \( k_1 \sim k_2 \sim k_3 \sim k \):

\[
B_{222}^{UV}(k_1, k_2, k_3) \supset \frac{89}{2401} b_1^3 k^6 \int_q \frac{P_L^6(q)}{q^6},
\]
(B.4)
\[
B_{222}^{IR}(k_1, k_2, k_3) \supset \frac{2}{21} b_1^3 k^2 P_L^2(k) \int_q \frac{P_L(q)}{q^2}.
\]

For \( P_L(q) \sim q^\nu \), the integral is UV convergent for \( \nu < 1 \) and IR convergent for \( \nu > -1 \). By choosing the bias in this range, FFTLog will give accurate values for the integrals. In contrast, if \( \nu \) is chosen to be outside this range, we must add the relevant UV or IR limits by hand (taking care to include subleading divergences if necessary).
Considering all bias terms, the limits of $B_{222}$ take the following schematic form for equilateral triangles:

$$B_{222}^{\text{UV}} \sim \int_q P_L^2(q) \left\{ \frac{k}{q} b_2^3 + \frac{b_2^2}{q} \left[ b_1^2 f_1(b_1, \gamma_2) + \left( \frac{k}{q} \right)^4 b_2 f_2(b_1, \gamma_2) + \left( \frac{k}{q} \right)^6 f_2(b_1, \gamma_2) \right] \right\}$$

$$B_{222}^{\text{IR}} \sim P_L^2(k) \int_q P_L(q) \left\{ \frac{k^2}{q} \left[ b_1^3 f_1^2(b_2, \gamma_2) + b_1^2 f_2^2(b_2, \gamma_2) + f_2^2(b_2, \gamma_2) \right] \right\},$$

where $\{f_i\}$ are some polynomials, and we consider only the leading-order contribution for each bias parameter. Inserting $P_L(q) \sim q^n$ as before shows that a term containing $K$ powers of $b_2$ is UV convergent for $\nu < 1 - 2K/3$, implying $\nu < -1$ for $b_2^3$, significantly tighter than the $\nu < 1$ limit for matter (i.e. $b_1^3$). However, the UV limit of $b_2^3$ is fully degenerate with the bispectrum shot-noise ($\epsilon_0$ in 2.13), and should be subtracted off in practice, as for the $b_2^3$ contribution to $P_{22}$. If we adopt $\nu > -1$, this term will not be captured by the FFTLog formalism, thus the subtraction becomes implicit. In this case, we require $\nu < -1/3$ to avoid the $b_2^3$ divergence (and the second-order $b_3^2$ divergence, both of which are degenerate with the $\epsilon_2$ stochasticity in 2.13). In the IR, (B.5) shows that the integral is convergent for $\nu > -1$ for terms involving two or more powers of $b_1$, and $\nu > -3$ else. To satisfy all the conditions simultaneously, we may take $-1 < \nu < -1/3$, dropping the shot-noise piece.

For $B_{321}^{I}$, the limiting UV and IR form is given by

$$B_{321}^{I\text{UV}} \sim b_1 P_L(k) \int_q P_L^2(q) \left\{ \frac{1}{2} X^2 + \left( \frac{k}{q} \right)^2 X f_1(\cdots) + \left( \frac{k}{q} \right)^4 f_2(\cdots) \right\}$$

$$B_{321}^{I\text{IR}} \sim b_1 P_L^2(k) \int_q P_L(q) \left\{ \left( \frac{k}{q} \right)^2 \left[ b_1^2 f_3(\cdots) + b_1 f_4(\cdots) \right] + f_6(\cdots) \right\},$$

where $X \in \{b_2, b_3, \gamma_2\}$ and ellipses are taken to mean bias operators excluding $X$ (in the UV) or $b_1$ (in the IR). (B.6) implies that UV divergences can be avoided if we take $\nu < (1 - 2K)/2$ when the term involves $K$ powers of $b_2$, $b_3$, or $\gamma_2$; these are all the composite operators appearing at third order. Furthermore, as in $B_{222}$, the UV limits of the terms involving two powers of $b_2$, $b_3$ and $\gamma_2$ are proportional to shot-noise (this time of the $\eta_0$ variety in 2.13),\(^9\) and should be subtracted off in practice (or dropped implicitly by fixing $\nu > -3/2$). In the IR, divergences vanish for $\nu > -1$ for terms involving $b_1^3$ or $b_1^2$ and $\nu > -3$ else. Overall, we require a bias of $-1 < \nu < -1/2$ to satisfy all conditions, assuming subtraction of the $\eta_0$ shot-noise contributions.

For $B_{321}^{I}$, we require the UV and IR convergence properties of $\int_q Z_3(k, q, -q) P_L(q)$, which we label $\tilde{P}_{13}(k)$ by analogy with the galaxy power spectrum (2.9). This natively involves all bias operators in (2.2) up to third order; however, this set is reduced to just $\{\delta, G_2(\Phi_e), \mathcal{G}_2(\varphi_2, \varphi_1)\}$ when the renormalization conditions are applied. These conditions demand

$$\lim_{k_1 \to 0} \langle X(k) \delta_L(k_1) \rangle_{\text{loop}} = 0 \quad (B.7)$$

\(^9\)This occurs regardless of the triangle configuration, once permutations are taken into account.
for renormalized operator $X$ and linear density field $\delta_L(k)$, \textit{i.e.} there can be no loop contributions which do not decay in the UV limit \cite{106}. The contribution of all composite operators (\textit{e.g.}, $\delta^2$) to $\tilde{P}_{13}$ is exactly that of a non-decaying loop diagram (since there is no suppression by the $F_3$ kernel), thus must vanish when the operators are properly renormalized. This leaves only $G_3$, which evaluates to zero after averaging over the angular part of $q$. Following these redefinitions (which do not affect $B_{222}$ and $B_{321}^{f_1}$), we find the UV and IR limits:

\begin{equation}
\tilde{P}_{13}^{\text{UV}} \sim \int_q P_L(q) \left( \frac{k}{q} \right)^2 f_1(b_1, \gamma_2, \gamma_{21}) \tag{B.8}
\end{equation}

\begin{equation}
\tilde{P}_{13}^{\text{IR}} \sim \int_q P_L(q) \left\{ \left( \frac{k}{q} \right)^2 b_1 f_2 + f_3(\gamma_2, \gamma_{21}) \right\}.
\end{equation}

UV divergences occur unless $\nu < -1$, and IR divergences occur unless $\nu > -1$ (for $b_1$) or $\nu > -3$ (else). As for $P_{13}$ \cite{99}, there is no range of biases which satisfy all the conditions; in this case, we can choose $-1 < \nu < 1$ (satisfying the IR limits, and avoiding subleading UV divergences at $\nu > 1$), and correct the UV part by adding the relevant limit by hand, which takes the following explicit form in real-space:

\begin{equation}
\tilde{P}_{13}^{\text{UV}}(k) = -k^2 \left( \frac{61}{315} b_1 + \frac{64}{35} \gamma_2 - \frac{32}{15} \gamma_{21} \right) \int_0^\infty \frac{q^2 dq}{2\pi^2} P(q) q^2,
\end{equation}

proportional to the velocity divergence $\sigma_v^2$.

Finally, we consider $B_{411}$. This contains the fourth-order bias operators, and involves Wick contractions of linear density fields within the same operator, permitting simplification via the renormalization condition:

\begin{equation}
\lim_{k_1, k_2 \to 0} \langle X(k) \delta_L(k_1) \delta_L(k_2) \rangle_{\text{loop}} = 0,
\end{equation}

which is proportional to the UV limit of $B_{411}$. The first effect of this is to remove contributions from any fourth-order composite local evolution operator (such as $\delta^4$ or $\delta G_3$); these operators were already dropped from the bias expansion in (2.2). Secondly, this will remove a number of UV divergences in the below. Before bias renormalization, the UV and IR limits of the remaining terms take the form:

\begin{equation}
B_{411}^{\text{UV}} \sim b_1^2 P_L^2(k) \int_q P_L(q) \left\{ f_1(b_2, \gamma_2, \gamma_{21}) + \left( \frac{k}{q} \right)^2 f_2(\cdots) \right\} \tag{B.11}
\end{equation}

\begin{equation}
B_{411}^{\text{IR}} \sim b_1^2 P_L^2(k) \int_q P_L(q) \left\{ \left( \frac{k}{q} \right)^2 f_3(b_1, b_2, \gamma_2) + f_4(\cdots) \right\},
\end{equation}

where ellipses represent additional bias terms which impatience lead us to ignore. The first line is UV convergent for $\nu < -3$ (first term, involving composite operators) or $\nu < -1$ (second term, no composite operators). However, the first term possesses a UV limit that does not decay as an \textit{negative} integer power of $q^2$, violating the renormalization condition (B.10). The precise action of bias operator renormalization is to remove such terms (and
By evaluating the diagram with $\nu > -1$, such contributions will be avoided, i.e., the operators will be correctly renormalized. In the IR, we find that divergences can be avoided by setting $\nu > -1$ (for terms involving the first and second order operators proportional to $b_1$, $\gamma_2$ or $b_2$), or $\nu > -3$ (for the remaining terms). As for $\tilde{P}_{13}$, there is no single bias that will simultaneously remove all the UV and IR divergences in $B_{411}$, even after bias renormalization. Fixing $-1 < \nu < 1$, we may compute the full expression by manually adding the appropriate UV limit to the FFTLog result. These limits can be computed straightforwardly from the kernels in MATHEMATICA and are omitted from this publication to avoid unnecessary tedium.

**B.2 Redshift-Space**

In redshift-space the perturbation theory kernels depend not only on the lengths $q$, $|k_1 - q|$ and $|k_2 + q|$ but also the LoS angles $\mu_i \equiv \hat{k}_i \cdot \hat{n}$ and $\hat{q} \cdot \hat{n}$.

Although we are primarily interested only in the bispectrum monopole (i.e., that integrated over $\mu_1, \mu_2$, with a suitable Lebesque measure, as in 2.20), the full dependence on $\mu_i$ is necessary for accurate calculation of coordinate distortions (2.21), thus we cannot simply average over $\mu_i$ before computing the loop integrals; furthermore, this is difficult to perform analytically due to the presence of high powers of $\hat{q} \cdot \hat{n}$.

After expanding the kernels as polynomials, we will find loop integrals of the form:

$$\int_q \frac{(\hat{q} \cdot \hat{n})^n}{q^{2\nu_1} |k_1 - q|^{2\nu_2} |k_2 + q|^{2\nu_3}} \tag{B.12}$$

for $n \in \{0, 1, \cdots, 6\}$ (cf. 3.1), with prefactors depending on $\mu_i$, $k_i$, biases and $f(z)$. Below, we consider how to compute this utilizing the FFTLog procedure, generalizing the approach of [112] for the power spectrum.

First, we expand the $\hat{q} \cdot \hat{n}$ angles as Cartesian sums, i.e., $\sum_{i=1}^3 \hat{q}_i \hat{n}^i$, and pull out the LoS vectors from the integral. The remaining function is a fully symmetric rank-$n$ tensor, given by

$$F_{i_1 \cdots i_n} \equiv \int_q \frac{\hat{q}^{i_1} \cdots \hat{q}^{i_n}}{q^{2\nu_1} |k_1 - q|^{2\nu_2} |k_2 + q|^{2\nu_3}}. \tag{B.13}$$

This has dependence only on $k_1$ and $k_2$; as such, its tensorial dependence can be written in terms of the components of $k_1$, $k_2$, and any isotropic tensors of relevance, i.e., the Kronecker delta.

Explicitly, this takes the form:

$$F_{i_1 \cdots i_n} = \sum_k A_k O_k^{i_1 \cdots i_n}, \tag{B.14}$$

where $\{O_k\}$ is the set of all independent symmetric rank-$n$ combinations of $\hat{k}_1^i$, $\hat{k}_2^j$ and $\delta_k^{ij}$. As an example, the $n = 2$ operators are $\{\delta_k^{ij}, \hat{k}_1^i \hat{k}_1^j, \hat{k}_1^i \hat{k}_2^j + \hat{k}_2^i \hat{k}_1^j, \hat{k}_2^i \hat{k}_2^j\}$. We then define an

---

10 In the isotropic approximation of infrared resummation (§2) the angular dependence is purely polynomial.

11 Note that the Levi-Cevita tensor, which is relevant for rank-three tensors and above, does not contribute to the expansion of (B.13), since it is antisymmetric.
where the first set of parentheses contains a set of \( \hat{B} \).

Applying it as a simplification step before the loop integrals are computed as in Appendix B.1.

\( n_k \) triangles (with \( k \), giving the correlation between basis elements:

\[ \mathcal{I}_{kk'} = O^{k_1 \cdots k_n}_{k} O^{k'_1 \cdots k'_n}_{k} \quad (B.15) \]

(assuming Einstein summation conventions): this allows extraction of the \( A_k \) coefficients via \( A_k = [\mathcal{I}]_{kk}^{-1} O^{k_1 \cdots k_n}_{k_1 \cdots k_n} F^{k_1 \cdots k_n} \), where the second term is just the contraction of (B.14) with various powers of \( \hat{k}_1 \) and \( \hat{k}_2 \). Finally, we contract (B.14) with \( n \) copies of \( \hat{n} \) to yield

\[ \int_q q^{2\nu_1} |\hat{q} \cdot \hat{n}| k_1 - q |q|^{2\nu_2} |k_2 + q |^{2\nu_3} = \left( [\mathcal{I}]_{kk}^{-1} O^{k_1 \cdots k_n}_{k_1 \cdots k_n} F^{k_1 \cdots k_n} \right) \left( O^{k'_1 \cdots k'_n}_{k} \hat{n}_{i_1} \cdots \hat{n}_{i_n} \right), \quad (B.16) \]

where the first set of parentheses contains a set of \( \{ \hat{k}_{1,2} \cdot \hat{q} \) coefficients inside the \( q \) integral, and the second contains powers of \( \mu_{1,2} \). To make this explicit, we give the \( n = 1 \) case:

\[ \int_q q^{2\nu_1} |\hat{q} \cdot \hat{n}| = \frac{\mu_1 - \mu_2 \nu_{12}}{1 - \nu_{12}^2} \int_q q^{2\nu_1} |\hat{q} \cdot \hat{k}_1| 1 - q |q|^{2\nu_2} |k_2 + q |^{2\nu_3} \]

writing \( \nu_{12} = \hat{k}_1 \cdot \hat{k}_2 \). In this manner, the FFTLog integral can be performed for arbitrarily large \( n \). We adopt this method to compute the bispectrum templates in redshift space, applying it as a simplification step before the loop integrals are computed as in Appendix B.1.

Notably, the above decomposition breaks down in the limit of \( \hat{k}_1 \cdot \hat{k}_2 \rightarrow -1 \), i.e. for \( k_1 \parallel k_2 \), whence there is only one angle in the problem. The corresponds to flattened triangles (with \( k_1 = k_2 + k_3 \) or \( \sqrt{x} + \sqrt{y} = 1 \)), which contain the divergence \( 1 - \nu_{12}^2 \rightarrow 0 \). Strictly speaking, this divergence is cancelled by the numerators, once \( \hat{k}_1 = -\hat{k}_2 \) is identified; however, if one separately computes the loop integral coefficients proportional to powers of \( \mu_1 \) and \( \mu_2 \), numerical issues will arise. In this limit, we adopt a different angular decomposition, noting that (B.13) can depend only on \( k_1 \) and the Kronecker delta. The basis tensors are much simpler in this case, for example, with \( \{ \delta_0^{k_1}, \hat{k}_1 \hat{k}_1 \} \) for \( n = 2 \), and facilitate computation in an analogous manner to the above. For \( n = 1 \), (B.17) becomes

\[ \int_q q^{2\nu_1} |\hat{q} \cdot \hat{n}| = \mu_1 \int_q q^{2\nu_1} |\hat{q} \cdot \hat{k}_1| \quad (B.18) \]

which does not diverge. This divergence also illustrates the importance of expanding the bispectrum templates in the \( \{ \mu, \chi \} \) basis (with \( \mu \equiv \mu_1, \chi \equiv \sqrt{1 - \mu^2 \cos \phi} \)) rather than \( \{ \mu_1, \mu_2 \} \) (cf. §3): the former is undefined for flattened triangles, whence \( \mu_1 = -\mu_2 \), whilst the latter simply has dependence only on \( \mu \) in this limit (noting that \( \mu_2 = \mu \nu_{12} - \nu_{12} \chi \rightarrow -\mu \) as \( \nu_{12} \rightarrow -1 \)).

**B.3 Implementation**

The above tricks allow us to efficiently compute the one-loop bispectrum in redshift space.

A rough overview of the computation is the following:
1. Expand the relevant (symmetrized) perturbation theory kernels as polynomials in $q$, $|k_i \pm q|$, $\mu_i$, and $\hat{q} \cdot \hat{n}$.

2. Replace powers of $\hat{q} \cdot \hat{n}$ with (B.16) and expand.

3. Create lookup tables giving each $\{\nu_1, \nu_2, \nu_3\}$ triplet of FFTLog frequencies and the associated coefficient (depending on biases and $\mu_i$).

4. Switch from the native $\{\mu_1, \mu_2, \mu_3\}$ basis to the (divergence-free) $\{\mu, \chi\}$ angular basis.

5. Compute the bispectrum templates for each of the 47 combinations of $\mu^\chi j$ and the relevant combinations of bias parameters using the FFTLog algorithm. This is performed for a grid of values of $k_1, x, y$, with flattened templates (obeying $\sqrt{x} + \sqrt{y} = 1$) computed separately, using the alternate angular decomposition given in Appendix B.2, and involving only 7 non-trivial powers of $\mu$.

6. Create a three-dimensional linear interpolator for each template using the precomputed bispectrum shapes (combining full and flattened configurations).

7. Apply binning and angular integration using (3.4), including the Alcock-Paczynski parameters if necessary.

8. Compute the full bispectrum as a sum over templates, weighted by the bias configurations and any necessary discreteness weights.

Notably, only steps (5) and beyond depend on the power spectrum template, and thus the cosmological survey in question. We note one further subtlety: computing the bispectrum templates near (but not at) the flattened limit of $\sqrt{x} + \sqrt{y} = 1$ can lead to numerical issues due to large values of $1/(1 - \nu_2^2)$, which appear in the angular decompositions of $(\hat{q} \cdot \hat{n})$ raised to the $n$-th power (cf. B.17). To counter this, when the templates are being computed and the condition $\sqrt{x} + \sqrt{y} < 1.1$ is met, we replace the FFTLog prefactor by its Taylor series in $(1 + \nu_2^2)$, artificially removing the divergent terms (which are present only due to numerical inaccuracies). Mathematica and Python code implementing all of the above steps is publicly available at GitHub.com/OliverPhilcox/OneLoopBispectrum.

Following initial testing of the FFTLog routines against explicit numerical integration for a small number of bins, we use the following choices of FFTLog bias: $\nu = -0.6$ for $B_{222}$ and $B_{321}^{II}$, $\nu = -0.3$ for the matter terms in $B_{321}^{I}$ (i.e. those involving only $b_1$ and $f$) and $B_{411}$, and $\nu = -0.8$ for the remaining terms in $B_{321}^{I}$. We additionally use 64 FFTLog frequencies for $B_{222}$, 96 frequencies for $B_{321}^{I}$, and 128 frequencies for $B_{321}^{II}$ and $B_{411}$, which reduces computational costs, whilst keeping good resolution on the terms with enhanced oscillatory behavior. Additionally, we manually add the UV limits of $B_{321}^{II}$ and $B_{411}$ (as in Appendix B.1), and choose the $k$-range for the FFTLog decomposition to lie in $[10^{-5}, 10] h^{-1}$ Mpc, computing all shapes on an interpolation grid of 64 $k_1$-values logarithmically spaced in $[10^{-5}, 0.5] h^{-1}$ Mpc, and 41 equally-spaced values of each of $\sqrt{x}$ and $\sqrt{y}$, subject to the triangle conditions. This is sufficient to ensure that the spectra are subpercent accurate in the regime of interest; the results are largely unchanged if the number of FFTLog frequencies is reduced by a factor of two. The computation requires $\sim$
$10^4$ CPU-hours to compute all templates (entirely performed within MATHEMATICA), with the majority of time devoted to $B_{222}$, and could certainly be optimized further. Calculations have been compared against explicit numerical integration of the (unsimplified) bispectrum kernels, and we find excellent sub-percent agreement in all cases.

## C Counterterms from Redshift-Space Distortions

The RSD mapping to $O(\delta_z^4)$ can be obtained by expanding (2.7) to fourth order:

$$\delta_k = \delta_k - ik_z[(1 + \delta_g)u_z]_k + \frac{i^2}{2} k_z^2[(1 + \delta_g)u_z^2]_k$$  

(C.1)

where $A_z \equiv A \cdot \hat{n} \equiv A_{\mu} A$. To facilitate renormalization, we must smooth this expansion with a low-pass filter of some size $R = \Lambda^{-1}$. Products of fields at the same point (contact terms) are sensitive to short-scaled modes and hence must to be smoothed and renormalized. We denote these operations by square brackets, $[...]_R$. Galilean symmetry implies the following schematic structure of the renormalized correlators (see [135–137] for the first order results),

- $u^i(1 + \delta_g) \equiv 1 + \delta(\mu) u^i + O_u^i$,
- $u^i u^j(1 + \delta_g) \equiv 1 + \delta(\mu) u^i u^j + u^i 4 O^{(1)}_{u^i u^j} + u^i 2 O^{(2)}_{u^i u^j}$,
- $u^i u^k(1 + \delta_g) \equiv 1 + \delta(\mu) u^i u^k + 3 u^i 2 O^{(1)}_{u^i u^k} + 3 u^i 2 O^{(2)}_{u^i u^k}$,
- $u^i u^j u^k(1 + \delta_g) \equiv 1 + \delta(\mu) u^i u^j u^k + 4 u^i 2 O^{(1)}_{u^i u^j u^k} + 4 u^i 2 O^{(2)}_{u^i u^j u^k} + O^{(3)}_{u^i u^j u^k}$.

(C.2)

where $u^i$, $\delta \ell$ are the smoothed long-wavelength velocity and density fields (for clarity, we will drop the subscript $\ell$ in the below). To preserve Galilean symmetry, the operators $O$ should not depend on the smoothed velocity field.

Note that the velocity field scales like $k^{-1} \delta_k$ at the linear order, i.e.

$$u_z = -\frac{\partial}{\Delta} \theta,$$

(C.3)

for Laplacian $\Delta$. Thus, in order to obtain the redshift-space bispectrum counterterms scaling like $k^2 \delta^2$, we need to retain four types of correction: (1) $O(k^3 \delta^2)$ arising from $u^i(1 + \delta_g)$, (2) $O(k^3 \delta^2)$ coming from $u^i u^j(1 + \delta_g)$, (3) $O(k^3 \delta^2)$ coming from $u^i u^j u^k(1 + \delta_g)$, and (4) $O(k^3 \delta^2)$ coming from $u^i u^j u^k u^l(1 + \delta_g)$.

We now discuss each type of composite operator. The renormalized operator $u^i(1 + \delta_g)$ with counterterms $O(k^3 \delta^2)$ is given by

$$[u^i(1 + \delta_g)]_R = u^i(1 + \delta) + \tilde{c}_1 \partial_i \delta + \tilde{c}_2 \partial_i(s_{kl} s_{kl}) + \tilde{c}_3 \partial_i \delta^2 + \tilde{c}_4 s_{kl} \partial_i s_{kl} + \tilde{c}_5 \partial_i \delta,$$

(C.4)

where $\delta$ is the matter density field and we have introduced the tidal tensor $s_{ij} = (\partial_i \partial_j \Phi - \delta_{ij}(\Delta \Phi / 3)) / (\frac{3}{2} \mathcal{H}^2)$. The renormalized operator $u^i u^j(1 + \delta_g)$ with counterterms $O(k^3 \delta^2)$ is given by

$$[u^i u^j(1 + \delta_g)]_R = u^i u^j(1 + \delta) + \tilde{c}_1 \partial_i \delta + \tilde{c}_2 s_{ij} + c_3 \delta^2 \delta_{ij} + \tilde{c}_4 s_{ij} s_{ij},$$

(C.5)
The renormalized operator \([u^i u^j u^k (1 + \delta_g)]_R\) with counterterms \(\mathcal{O}(k^{-1}\delta^2)\) is given by

\[
[u^i u^j u^k (1 + \delta_g)]_R = u^i u^j u^k (1 + \delta) + \tilde{b}_1 u(i\delta_{jk}) + \tilde{b}_2 u(i s_{jk}) + \tilde{b}_3 u(i s_{jk}), \tag{C.6}
\]

where \((...)\) denotes symmetrization over indices. Finally, the renormalized operator \([u^i u^j u^k u^m (1 + \delta_g)]_R\) with counterterms \(\mathcal{O}(k^{-2}\delta^2)\) is given by

\[
[u^i u^j u^k u^m (1 + \delta_g)]_R = u^i u^j u^k u^m (1 + \delta) + d_1 u(i u_j \delta_{kl}). \tag{C.7}
\]

In combination, once certain redundant operators are removed, we find the following redshift-space bispectrum counterterm kernel,

\[
Z^{ctr}_{22}(k_1, k_2) = \left(C_1 \mu^2 k_3^2 F_2(k_1, k_2) + C_2 \mu^2 \left(\mu_3^2 - \frac{1}{3}\right) F_2(k_1, k_2) + C_3 k_3^2 \right.

\begin{align*}
+ & C_4 k_3^2 \left(\frac{(k_1 \cdot k_2) \mu_1 \mu_2}{k_1 k_2} - \frac{\mu_1^2 + \mu_2^2}{3} + \frac{1}{9}\right) + C_5 f k_3^2 \mu_3^2 G_2(k_1, k_2) \\
+ & C_6 f k_3^2 \left[\frac{\mu_1}{k_1} \left(\mu_2 - \frac{1}{3}\right) + \frac{\mu_2}{k_2} \left(\mu_1 - \frac{1}{3}\right)\right] + C_7 f k_3^2 \left[\frac{\mu_1}{k_1} + \frac{\mu_2}{k_2}\right] + C_8 f^2 k_3^4 \frac{\mu_1^2}{k_1^2} \frac{\mu_2}{k_2} \\
+ & C_9 k_3^2 \left(\frac{(k_1 \cdot k_2)^2}{k_1 k_2} - \frac{1}{3}\right) \left\{\frac{1}{k_3^2 NL}\right. \\
\end{align*}
\]

involving nine Wilson coefficients, marked in color. The overall bispectrum counterterm is given by

\[
B^{ctr}_{22}(k_1, k_2, k_3) = 2 Z_1(k_1) Z_1(k_2) (F^{ctr}_{22} + Z^{ctr}_{22})(k_1, k_2) P_{11}(k_1) P_{11}(k_2) + \text{cyclic perms}, \tag{C.9}
\]

where \(F^{ctr}_{22}(k_1, k_2)\) is the real space, \(\mu\)-independent kernel. In principle, some of the redshift-space counterterms are the same as coefficients in front of the \(\mu\)-dependent \(k^2 P_{11}\) redshift-space counterterms in redshift space, and thus should be constrained by the power spectrum data. Indeed, at leading order we have [137]

\[
\delta^{CTR,LO}_2(k) = C_0 k^2 \delta + C_1 \mu^2 k^2 \delta + C_2 \mu^4 k^2 \delta, \tag{C.10}
\]

where

\[
C_1 = C_1 - \frac{C_2}{3}, \quad C_2 = C_2 + C_5 f. \tag{C.11}
\]

The relationship between our power spectrum counterterm coefficients (\(\propto k^2 P_L\)) and \(C_n\) is given by:

\[
\begin{align*}
c_0 &= b_1 \left(\frac{C_0 + C_1}{3} + \frac{C_2}{5}\right) + \frac{C_0 f}{3} + \frac{C_1 f}{5} + \frac{C_2 f}{7}, \\
c_2 &= \frac{1}{7f}(7b_1 C_1 + 6b_1 C_2 + 7C_0 f + 6C_1 f + 5C_2 f), \tag{C.12} \\
c_4 &= \frac{1}{11f^2}(11b_1 C_2 + 11C_1 f + 15C_2 f). 
\end{align*}
\]
D Stochastic terms

In this section we discuss stochastic contributions to the one-loop bispectrum.

D.1 Real space

The stochastic contributions to the galaxy density field in real space are given in terms of $\epsilon$ by (some operators are present in [137]):

$$
\delta_\epsilon = d_1 \epsilon + d_2 \epsilon \delta + \bar{d}_1 \left( \frac{k}{k_{NL}} \right)^2 \epsilon + d_2 \left( \frac{k}{k_{NL}} \right)^2 \left[ \epsilon \delta \right]_k + d_3 \left[ \delta^i \epsilon \delta^j \right]_k + d_4 \left[ \epsilon \Delta \delta \right]_k + d_5 \left[ \Delta \epsilon \delta \right]_k.
$$

(D.1)

Note that $\langle \epsilon \delta \rangle = 0$ by definition. There are two non-trivial possibilities to contract operators in $\delta_\epsilon$ to obtain the tree-level bispectrum contributions (with free coefficients shown in color):

$${\cal O}(P \bar{n}^{-1}): \langle b_1 \delta_{k_1} d_2 [\epsilon \delta]_{k_2 \epsilon_{k_3}} \rangle' = 2 d_2 \frac{b_1 P(k_1)}{\bar{n}} + \text{cyc.},
$$

(D.2)

$${\cal O}(\bar{n}^{-2}): \langle \bar{d}^3 \rangle' = \frac{d_1 \bar{A}_{\text{shot}}}{\bar{n}^2},
$$

where primes denote that we drop the Dirac delta function. These match the operators present in the tree-level bispectrum model [80]. At the one-loop order we find three distinct contractions:

$${\cal O}(k^2 \bar{n}^{-1}): \langle b_1 \delta_{k_1} \bar{d}_2 k_2^2 \epsilon_{k_2} \epsilon_{k_3} \rangle' = b_1 \bar{d}_2 \frac{(k_2^2 + k_3^2) P(k_1)}{\bar{n}} + \text{cyc.},
$$

$${\cal O}(k^2 P \bar{n}^{-1}): \langle b_{\gamma_2} \bar{d}_1^2 k_1^2 \epsilon_{k_1} \epsilon_{k_2} \epsilon_{k_3} \rangle' = b_{\gamma_2} \bar{d}_2 \frac{k_1^2 P(k_1)}{\bar{n}} + \text{cyc.},
$$

(D.3)

$${\cal O}(k^2 \bar{n}^{-2}): \langle \bar{d}_1 k_1^2 \epsilon_{k_1} \epsilon_{k_2} \epsilon_{k_3} \rangle' = \bar{d}_1 A_{\text{shot}} \frac{k_1^2}{\bar{n}^2} + \text{cyc.}.
$$

These reproduce the stochastic bispectrum contribution derived in [79].

D.2 Redshift space

In redshift space, the density field involves the following terms at tree-level:

$$
\delta^{(z)}(k) = Z_1(k) \delta^{(1)}(k) + [Z_2(\delta^{(1)})^2]_k + d_4 [\delta^{(1)} \epsilon]_k - i f k_z d_1 \left[ \epsilon \frac{z_i \partial_i}{\Delta^{(1)}} \right]_k,
$$

(D.4)

whilst at next-to-leading order (i.e. one-loop), we have the additional stochastic contributions,

$$
\delta^{(z^2 \mu^2)}_{\text{stoch}}(k) = (d_1 f \mu^2 k^2 \epsilon + k_z z^i (\epsilon^i + k_i \epsilon) + k_z^2 z_i z_j \epsilon^{ij}) (1 + \delta_g).
$$

(D.5)

At leading order we have the LoS-dependent $P \bar{n}^{-1}$ term derived in [80]. To obtain the one-loop expression, we must perform a renormalization of the contact operators involving...
the stochastic fields, as in Appendix C. The contact operator renormalization \( O(\epsilon, \epsilon \delta) \) at the leading order in derivatives, gives the following terms, schematically:

\[
[u_i(1 + \delta_y)]_R \supset u_i^{(1)} \epsilon + (\epsilon_i + k_i \epsilon)(1 + \delta_y) + c \partial_\delta \epsilon + \partial_\epsilon (\epsilon \delta),
\]

\[
[u_i u_j (1 + \delta_y)]_R \supset (\delta_{ij} \epsilon + \epsilon_{ij})(1 + \delta) + u_i^{(1)} \epsilon_j + u_i^{(1)} k_j \epsilon + s_{ij} \epsilon,
\]

\[
[u_i u_j u_k (1 + \delta_y)]_R \supset u_i^{(1)} \delta_{jk} \epsilon + u_i^{(1)} \epsilon_{jk}.
\]

These operators generate the following LoS-dependent correction to the stochastic component of the galaxy density field:

\[
\delta_{\text{stoch}} \supset k_z u_i^{(1)} \epsilon + k_z^2 (\epsilon_z + \partial_\epsilon \epsilon)(1 + \delta_y) + k_z^2 (\epsilon + \epsilon_{zz})(1 + \delta_y)
+ k_z^2 \epsilon_z + k_z^2 u_z^{(1)} \epsilon_z + k_z^3 u_z^{(1)} \epsilon_z + k_z^3 \epsilon_{zz} + k_z \epsilon_z^2 + k_z^4 \epsilon_{zz}.
\]

Focussing on the \( k^2 P(k)/\bar{n} \) corrections to the bispectrum that are generated by this density field (D.7), we find three possibilities:

1. Insertion of the redshift-space counterterms \( k^2 \delta \) and the real-space operator \( d_2 \delta \epsilon \), i.e. \( (k^2 \delta)(\epsilon \delta) \epsilon \):

\[
\langle k^2 \mu_1^{2a} \delta_{k_1} d_2[\epsilon \delta]_{k_2} \epsilon_{k_3} \rangle' = d_2 \frac{k^2 \mu_1^{2a} P_L(k_1)}{\bar{n}} + \text{cyc.},
\]

\[
\langle k^2 \mu_1^{2a} \delta_{k_1} d_2 f k_{2z} d_1 \left[ \epsilon \frac{\partial}{\Delta} \epsilon(1) \right]_{k_2} \epsilon_{k_3} \rangle' = -f d_1 \frac{k^2 \mu_1^{2a+2} P_L(k_1)}{\bar{n}} + \text{cyc.},
\]

where \( \alpha = 0, 1, 2 \).

2. Insertion of the redshift-space stochastic counterterm \( \sim k^2 \epsilon \) and the real-space operator \( d_2 \delta \epsilon \), i.e.

\[
\langle Z_1(k_1) \delta_{k_1} d_2[\epsilon \delta]_{k_2} \epsilon_{k_3} \rangle' = d_2 \frac{Z_1(k_1) P_L(k_1) (k^2_{2z} + k^3_{3z})}{\bar{n}} + \text{cyc.}.
\]

3. Insertion of the new redshift-dependent stochastic correction \( k^2 \delta \epsilon \), i.e.

\[
\langle Z_1(k_1) \delta_{k_1} k^2_{2z} \epsilon[\delta \epsilon]_{k_2} \epsilon_{k_3} \rangle' = \frac{Z_1(k_1) P_L(k_1) (k^2_{2z} + k^3_{3z})}{\bar{n}} + \text{cyc.},
\]

\[
\langle Z_1(k_1) \delta_{k_1} k^2_{2z} \left[ \epsilon \frac{\partial}{\Delta} \frac{\partial \epsilon}{\Delta} \right]_{k_2} \epsilon_{k_3} \rangle' = \frac{Z_1(k_1) P_L(k_1) (2 \mu_1 k^2_{2z} k^3_{3z})}{\bar{n}} + \text{cyc.},
\]

\[
\langle Z_1(k_1) \delta_{k_1} k^3_{2z} \left[ \epsilon \frac{\partial \epsilon}{\Delta} \right]_{k_2} \epsilon_{k_3} \rangle' = \left( \frac{Z_1(k_1) P_L(k_1) k^2_{2z}}{k^2_{1z}} (k^2_{2z} + k^3_{3z}) + \text{cyc.} \right) \frac{1}{\bar{n}},
\]

\[
\langle Z_1(k_1) \delta_{k_1} k^2_{2z} \left[ \epsilon \frac{\partial^2}{\Delta^2} \epsilon \right]_{k_2} \epsilon_{k_3} \rangle' = \left( Z_1(k_1) P_L(k_1) \left( \mu_1^2 - \frac{1}{3} \right) (k^2_{2z} + k^3_{3z}) + \text{cyc.}\right) \frac{1}{\bar{n}},
\]

\[
\langle Z_1(k_1) \delta_{k_1} k^2_{2z} \left[ \epsilon \frac{\partial^2}{\Delta^2} - \frac{1}{3} \delta \right]_{k_2} \epsilon_{k_3} \rangle' = \left( Z_1(k_1) P_L(k_1) \left( \mu_1^2 - \frac{1}{3} \right) (k^2_{2z} + k^3_{3z}) + \text{cyc.}\right) \frac{1}{\bar{n}}.
\]

(D.10)
Many of these terms happen to be linearly-dependent once we apply the bispectrum triangle condition. All in all, the $\mu$-dependent part of the stochastic bispectrum at order $k^2 P \bar{n}^{-1}$ is given by

$$B_{\text{stoch}}^{(k^2 P \bar{n}^{-1})}(k_1, k_2, k_3) = \sum_{n=1,2,3,4} S_n k_1^2 \mu_1^{2(n-1)} + S_5 Z_1(k_1)(\mu_2^2 k_2^2 + \mu_3^2 k_3^2)$$

$$+ S_6 Z_1(k_1) \mu_1^2 (\mu_2^2 k_2^2 + \mu_3^2 k_3^2) + S_7 Z_1(k_1) \frac{k_1}{k_1^2} (k_2^3 + k_3^3) \frac{P_L(k_1)}{k_5^2 \bar{n}} + \text{cyc.},$$

which involves additional nuisance coefficients shown in color.

In addition, there are purely stochastic terms that generate the bispectrum of the order $k^2 \bar{n}^{-2}$. These arise from the following combinations:

$$\langle \epsilon_{k_1} \epsilon_{k_2} \epsilon_{k_3} \rangle' = k_2^2 + \text{cyc.},$$

giving the bispectrum contribution

$$B_{\text{stoch}}^{(k^2 \bar{n}^{-2})}(k_1, k_2, k_3) = A_{\text{shot},2} \frac{k_2^2 + k_1^2 + k_3^2}{\bar{n}^2} .$$

After performing angular integration (in the absence of coordinate-distortion effects), this term takes the same form as the real-space term $\sim k^2 \bar{n}^{-2}$. Thus, it does not produce a new contribution to the bispectrum monopole, though is important if higher-order multipoles are also considered.

### E Prior volume effects

In this section we study the prior volume effects present in our posteriors when the one-loop bispectrum likelihood is analyzed with small data cuts, such as $k_B^{\text{max}} = 0.12 h \text{ Mpc}^{-1}$. At face value, the posterior distributions from this analysis are several $\sigma$ away from the true values. However, here we show that as much as half of this shift can be explained by prior volume (marginalization projection) effects. Indeed, such effects are expected to be present when the data volume is not sufficient to tightly constrain model parameters, which is the case for analyses with low $k_B^{\text{max}}$.

We performed the following test: rerunning our full analysis on the mock data generated by our fitting pipeline for the best-fit cosmology at $k_B^{\text{max}} = 0.15 h \text{ Mpc}^{-1}$. This mock data is simply a theory curve without any statistical scatter. In the absence of prior volume effects our pipeline must exactly recover the input parameters. However, when we fit this mock bispectrum data at $k_B^{\text{max}} = 0.12 h \text{ Mpc}^{-1}$, we find that the mean values recovered from our pipeline are shifted relative to the input values at the $(1-1.5)\sigma$ level, as shown in Fig. 5. This is evidence of prior volume effects. Furthermore, the shifts are in the directions of the apparent biases observed in the actual data ($\S 6$). Thus, if we subtract these shifts from the actual posteriors at $k_B^{\text{max}} = 0.12$, the mean posterior values would match the true input parameter values at least within the 99% CL. Finally, we note that the parameters
$b_2, b_{G2}$ and $b_{\Gamma 3}$ are highly correlated; this means that a shift in one would induce a shift in both.

As an additional check, we repeat our mock analysis for $k_{\text{max}}^B = 0.15 h \text{Mpc}^{-1}$. Overall, we find much improved agreement between the mock and actual analyses. The posteriors for $\sigma_8$ and $b_{\Gamma 3}$ are still shifted with respect to the ground truth by $\lesssim 1\sigma$ (which is smaller than 1.5$\sigma$ shifts in the $k_{\text{max}}^B = 0.12 h \text{Mpc}^{-1}$ case), but all other parameters are recovered without noticeable bias.

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Figure 5. Posterior distributions of the clustering amplitude and low-order nuisance parameters from MCMC analyses of the power spectrum and bispectrum likelihoods from the redshift-space analysis at $k_{\text{max}}^B = 0.12 \, \text{hMpc}^{-1}$ for the PT Challenge simulation data (in green) and for the mock bispectrum data vector (in gray) computed with our pipeline for the bestfit cosmology at $k_{\text{max}}^B = 0.15 \, \text{hMpc}^{-1}$. Dashed lines show the input values for the mock data-vector, whose discrepancies with the grey posteriors indicate clear evidence for prior volume effects.

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\[ P_l + Q_0 + B_0 (k_{\text{max}}^B = 0.15 h \text{Mpc}^{-1}) \]

**Figure 6.** As Fig. 5, but for \( k_{\text{max}}^B = 0.15 h \text{Mpc}^{-1} \).

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