

## Constraining $M_\nu$ with the Bispectrum I: Breaking Parameter Degeneracies

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### ABSTRACT

Massive neutrinos suppress the growth of structure below their free-streaming scale and leave an imprint on large-scale structure. Measuring this imprint allows us to constrain the sum of neutrino masses,  $M_\nu$ , a key ingredient for particle physics beyond the Standard Model. However, degeneracies with cosmological parameters limit the constraining power of standard two-point clustering statistics. In this work, we investigate whether we can break these parameter degeneracies and constrain  $M_\nu$  with the next higher-order correlation function, the bispectrum. We first examine the degeneracy between  $M_\nu$  and  $\sigma_8$  using 800  $N$ -body simulations from the HADES suite and demonstrate that the redshift-space halo bispectrum helps break this degeneracy. Next, we quantify the information content of the redshift-space halo bispectrum in a Fisher matrix forecast that includes  $M_\nu$  and cosmological parameters  $\Omega_m$ ,  $\Omega_b$ ,  $h$ ,  $n_s$ , and  $\sigma_8$ . This is the first  $N$ -body simulation based Fisher matrix forecast of cosmological parameters using the bispectrum. More importantly, for  $k_{\text{max}} = 0.2$  and  $0.5$  we find

*Keywords:* cosmology: —

### 1. INTRODUCTION

very brief intro on neutrinos

Brief intro on the impact of massive active neutrinos on the matter powerspectrum and how that's detectable with CMB and LSS

Quick summary of current constraints and where they come from. Talk about the CMB-LSS lever arm. The degeneracy between  $A_s$  and  $\tau$  and how that's a bottleneck short thing about how  $\tau$  is hard to constrain.

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Fortunately the imprint of neutrinos on the matter distribution leaves imprints on clustering. So with clustering measurements alone we can derive constraints on  $M_\nu$  and at the very least tighten constraints.

Brief summary of previous works that look at the powerspectrum. Then Discuss the shortcomings of the powerspectrum only analysis– Not good enough.

However, we don’t have to settle for just two point statistics, three-point statistics such as the bispectrum and 3PCF...

In Section blah

**CH:** List plans for paper 2

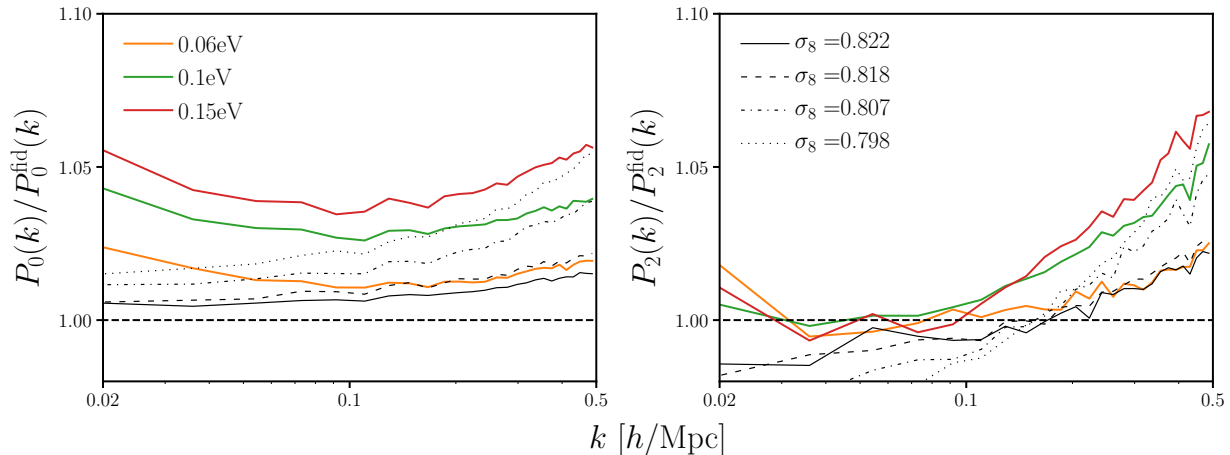
## 2. HADES AND QUIJOTE SIMULATION SUITES

We use a subset of the HADES<sup>1</sup> and Quijote simulation suites. Below, we briefly describe these simulations; a brief summary of the simulations can be found in Table 1. The HADES simulations start from Zel’dovich approximated initial conditions generated at  $z = 99$  using the Zennaro et al. (2017) rescaling method and follow the gravitational evolution of  $N_{\text{cdm}} = 512^3$  CDM, plus  $N_\nu = 512^3$  neutrino particles (for massive neutrino models), to  $z = 0$ . They are run using the GADGET-III TreePM+SPH code (Springel 2005) in a periodic  $(1h^{-1}\text{Gpc})^3$  box. All of the HADES simulations share the following cosmological parameter values, which are in good agreement with Planck constraints Ade et al. (2016):  $\Omega_{\text{m}}=0.3175$ ,  $\Omega_{\text{b}}=0.049$ ,  $\Omega_{\Lambda}=0.6825$ ,  $n_s=0.9624$ ,  $h=0.6711$ , and  $k_{\text{pivot}} = 0.05 \text{ hMpc}^{-1}$ .

The HADES suite includes models with degenerate massive neutrinos of different masses:  $M_\nu = 0.06, 0.10$ , and  $0.15 \text{ eV}$ . These massive neutrino models are run using the “particle method”, where neutrinos are described as a collisionless and pressureless fluid and therefore modeled as particles, same as CDM (Brandbyge et al. 2008; Viel et al. 2010). HADES also includes models with massless neutrino and different values of  $\sigma_8$  to examine the  $M_\nu - \sigma_8$  degeneracy. The  $\sigma_8$  values were chosen to match either  $\sigma_8^m$  or  $\sigma_8^c - \sigma_8$  computed with respect to total matter (CDM + baryons +  $\nu$ ) or CDM + baryons — of the massive neutrino models:  $\sigma_8 = 0.822, 0.818, 0.807$ , and  $0.798$ . Each model has 100 independent realizations and we focus on the snapshots saved at  $z = 0$ . Halos closely trace the CDM+baryon field rather than the total matter field and neutrinos have negligible contribution to halo masses (*e.g.* Ichiki & Takada 2012; Castorina et al. 2014; LoVerde 2014; Villaescusa-Navarro et al. 2014). Hence, dark matter halos are identified in each realization using the Friends-of-Friends algorithm (FoF; Davis et al. 1985) with linking length  $b = 0.2$  on the CDM + baryon distribution; only halos with masses  $> 3.2 \times 10^{13} h^{-1} M_\odot$  are included. For further details on the HADES simulations, we refer readers to Villaescusa-Navarro et al. (2018).

In addition to HADES, we use simulations from the Quijote simulation suite, a set of 23,000  $N$ -body simulations that in total contain more than 3.3 trillion ( $3.3 \times 10^{12}$ ) particles over a volume of  $23000(h^{-1}\text{Gpc})^3$ . These simulations were constructed to quantify the information content of different cosmological observables using Fisher matrix forecasting (*e.g.* Section 4.2). They are therefore designed to accurately calculate the covariance matrices of observables and the derivatives of observ-

<sup>1</sup> <https://franciscovillaescusa.github.io/hades.html>



**Figure 1.** Impact of  $M_\nu$  and  $\sigma_8$  on the redshift-space halo power spectrum monopole and quadrupole measured using the HADES simulation suite.  $M_\nu$  and  $\sigma_8$  produce almost identical effects on halo clustering on small scales ( $k > 0.1 h/\text{Mpc}$ ). This degeneracy can be partially broken through the quadrupole; however,  $M_\nu$  and  $\sigma_8$  produce almost the same effect on two-point clustering — within a few percent.

ables with respect to cosmological parameters. The suite considers 6 cosmological parameters:  $\Omega_m$ ,  $\Omega_b$ ,  $h$ ,  $n_s$ ,  $\sigma_8$ , and  $M_\nu$ .

To calculate covariance matrices, Quijote includes 15,000  $N$ -body simulations run at a fiducial cosmology ( $\Omega_m=0.3175$ ,  $\Omega_b=0.049$ ,  $h=0.6711$ ,  $n_s=0.9624$ ,  $\sigma_8=0.834$ , and  $M_\nu=0.0$  eV). It also includes sets of 500  $N$ -body simulations run at different cosmologies where only one parameter is varied from the fiducial cosmology at a time for the derivatives. Along  $\Omega_m$ ,  $\Omega_b$ ,  $h$ ,  $n_s$ , and  $\sigma_8$ , the fiducial cosmology is adjusted by either a small step above and below the fiducial value. Along  $M_\nu$ , because the derivative of certain observable with respect to  $M_\nu$  is noisy, Quijote includes sets of 500 simulations for  $M_\nu = 0.1$ ,  $0.2$ , and  $0.4$  eV. In Table 1, we list the cosmologies included in the Quijote suite.

The initial conditions for all Quijote simulations were generated at  $z = 127$  using 2LPT for simulations with massless neutrinos and the Zel’dovich approximation for massive neutrinos. Like HADES, the initial conditions of simulations with massive neutrinos take their scale-dependent growth factors/rates into account using the Zennaro et al. (2017) method. From the initial conditions, all of the simulations follow the gravitational evolution of  $512^3$  dark matter particles, and  $512^3$  neutrino particles (for massive neutrino models), to  $z = 0$  using GADGET-III TreePM+SPH code (same as HADES). The simulations run at the fiducial cosmology for covariance matrix estimation are standard  $N$ -body simulations. However, the rest are paired fixed simulations, which greatly reduce cosmic variance without introducing bias for a large set of statistics (Angulo & Pontzen 2016; Pontzen et al. 2016; Villaescusa-Navarro et al. 2018). We confirm that the paired fixed simulations do not introduce any bias for the redshift-space halo bispectrum (the observable we consider in this paper). For further details on the Quijote simulations, we refer readers to Villaescusa-Navarro et al. (in preparation).

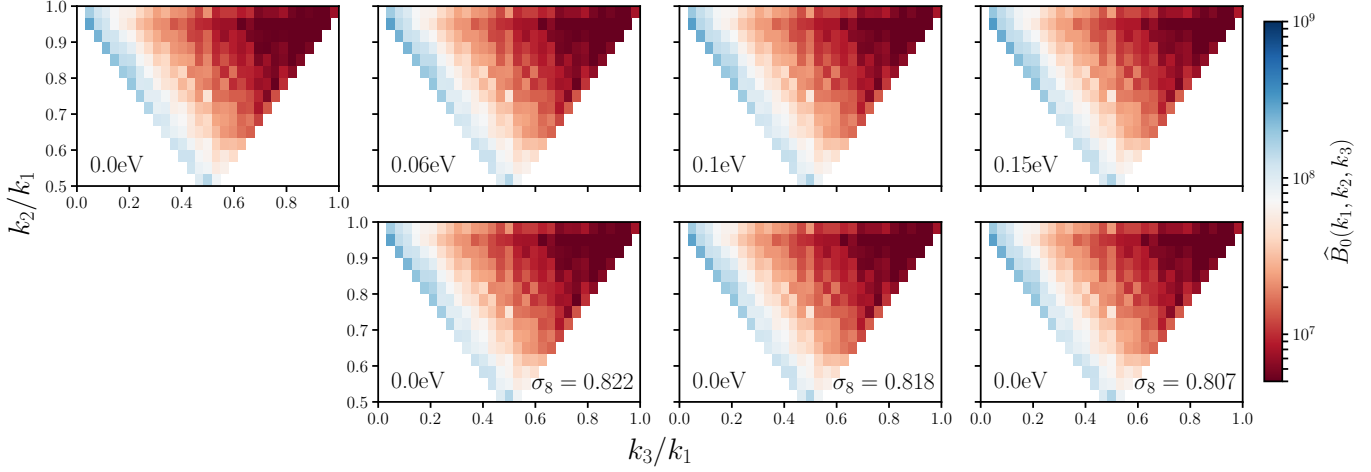
### 3. BISPECTRUM

**Table 1.** Specifications of the HADES and Quijote simulation suites.

Name	$M_\nu$ (eV)	$\Omega_m$	$\Omega_b$	$h$	$n_s$	$\sigma_8^m$ ( $10^{10}h^{-1}M_\odot$ )	$\sigma_8^c$ ( $10^{10}h^{-1}M_\odot$ )	realizations
HADES suite								
Fiducial	0.0	0.3175	0.049	0.6711	0.9624	0.833	0.833	100
	0.06	0.3175	0.049	0.6711	0.9624	0.819	0.822	100
	0.10	0.3175	0.049	0.6711	0.9624	0.809	0.815	100
	0.15	0.3175	0.049	0.6711	0.9624	0.798	0.806	100
	0.0	0.3175	0.049	0.6711	0.9624	0.822	0.822	100
	0.0	0.3175	0.049	0.6711	0.9624	0.818	0.818	100
	0.0	0.3175	0.049	0.6711	0.9624	0.807	0.807	100
	0.0	0.3175	0.049	0.6711	0.9624	0.798	0.798	100
Quijote suite								
Fiducial	0.0	0.3175	0.049	0.6711	0.9624	0.834	0.834	15,000
$M_\nu^+$	<u>0.1</u>	0.3175	0.049	0.6711	0.9624	0.834	0.834	500
$M_\nu^{++}$	<u>0.2</u>	0.3175	0.049	0.6711	0.9624	0.834	0.834	500
$M_\nu^{+++}$	<u>0.4</u>	0.3175	0.049	0.6711	0.9624	0.834	0.834	500
$\Omega_m^+$	0.0	<u>0.3275</u>	0.049	0.6711	0.9624	0.834	0.834	500
$\Omega_m^-$	0.0	<u>0.3075</u>	0.049	0.6711	0.9624	0.834	0.834	500
$\Omega_b^+$	0.0	0.3175	<u>0.050</u>	0.6711	0.9624	0.834	0.834	500
$\Omega_b^-$	0.0	0.3175	<u>0.048</u>	0.6711	0.9624	0.834	0.834	500
$h^+$	0.0	0.3175	0.049	<u>0.6911</u>	0.9624	0.834	0.834	500
$h^-$	0.0	0.3175	0.049	<u>0.6511</u>	0.9624	0.834	0.834	500
$n_s^+$	0.0	0.3175	0.049	0.6711	<u>0.9824</u>	0.834	0.834	500
$n_s^-$	0.0	0.3175	0.049	0.6711	<u>0.9424</u>	0.834	0.834	500
$\sigma_8^+$	0.0	0.3175	0.049	0.6711	0.9624	<u>0.849</u>	<u>0.849</u>	500
$\sigma_8^-$	0.0	0.3175	0.049	0.6711	0.9624	<u>0.819</u>	<u>0.819</u>	500

**Top:** The HADES suite includes sets of 100  $N$ -body simulations with degenerate massive neutrinos of  $M_\nu = 0.06, 0.10$ , and  $0.15$  eV as well as sets of simulations with massless neutrino and  $\sigma_8 = 0.822, 0.818, 0.807$ , and  $0.798$  to examine the  $M_\nu - \sigma_8$  degeneracy. **Bottom:** The Quijote suite includes 15,000  $N$ -body simulations at the fiducial cosmology to accurately estimate the covariance matrices. It also includes sets of 500 paired fixed simulations at different cosmologies, where only one parameter is varied from the fiducial value (underlined), to estimate derivatives of observables along the cosmological parameters.

We're interested in breaking parameter degeneracies that limit the constraining power on  $M_\nu$  of two-point clustering analyses using three-point clustering statistics — *i.e.* the bispectrum. In this section, we describe the bispectrum estimator used throughout the paper. We focus on the bispectrum monopole ( $\ell = 0$ ) and use an estimator that exploits Fast Fourier Transforms (FFTs).



**Figure 2.** The redshift-space halo bispectrum,  $\hat{B}_0(k_1, k_2, k_3)$ , as a function of triangle configuration shape for  $M_\nu = 0.0, 0.06, 0.10$ , and  $0.15$  eV (upper panels) and  $\sigma_8 = 0.822, 0.818$ , and  $0.807$  (lower panels). The HADES simulations of the top and bottom panels in the three right-most columns, have matching  $\sigma_8$  values (Section 2). We describe the triangle configuration shape by the ratio of the triangle sides:  $k_3/k_1$  and  $k_2/k_1$ . The upper left bin contains squeezed triangles ( $k_1 = k_2 \gg k_3$ ); the upper right bin contains equilateral triangles ( $k_1 = k_2 = k_3$ ); and the bottom center bin contains folded triangles ( $k_1 = 2k_2 = 2k_3$ ). We include all triangle configurations with  $k_1, k_2, k_3 \leq k_{\text{max}} = 0.5 h/\text{Mpc.}$  and use the  $\hat{B}_0$  estimator in Section 3.

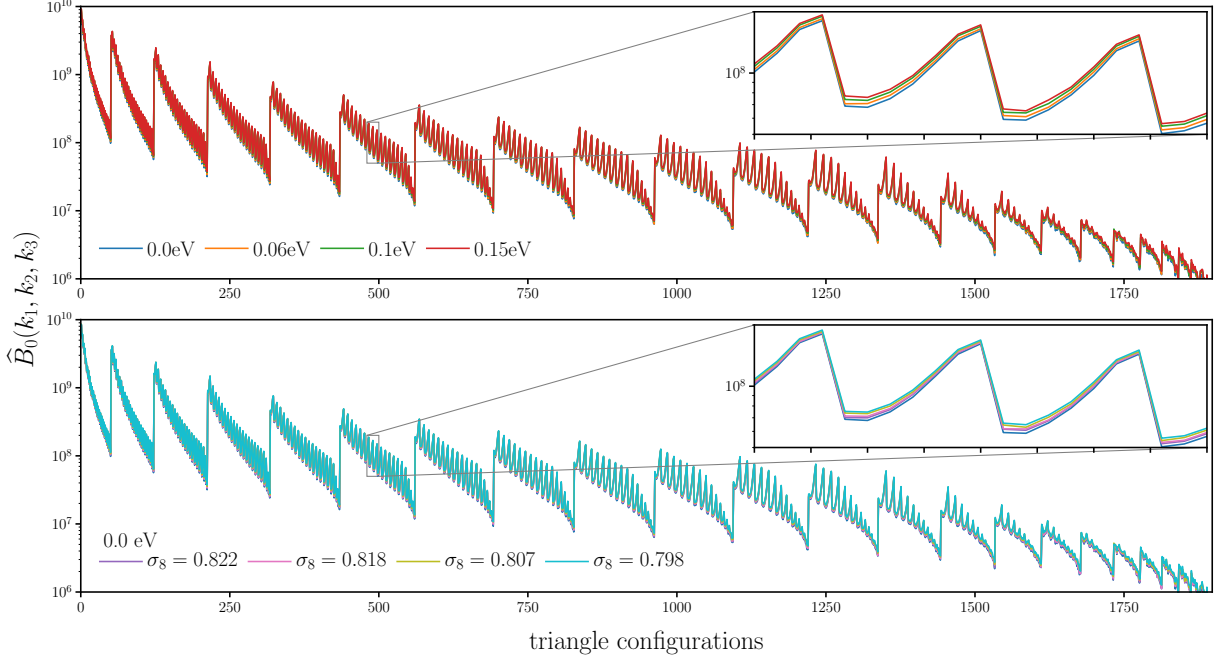
Our estimator is similar to the estimators described in Scoccimarro (2015); Sefusatti et al. (2016); we also follow their formalism in our description below. Although Sefusatti et al. (2016) and Scoccimarro (2015) respectively describe estimators in redshift- and real-space, since we focus on the bispectrum monopole, we note that there is no difference.

To measure the bispectrum of our halo catalogs, we begin by interpolating the halo positions to a grid,  $\delta(\mathbf{x})$  and Fourier transforming the grid to get  $\delta(\mathbf{k})$ . We use a fourth-order interpolation to interlaced grids, which has advantageous anti-aliasing properties (Hockney & Eastwood 1981; Sefusatti et al. 2016) that allow unbiased measurements up to the Nyquist frequency. Then using  $\delta(\mathbf{k})$ , we measure the bispectrum monopole as

$$\hat{B}_{\ell=0}(k_1, k_2, k_3) = \frac{1}{V_B} \int_{k_1} d^3 q_1 \int_{k_2} d^3 q_2 \int_{k_3} d^3 q_3 \delta_D(\mathbf{q}_{123}) \delta(\mathbf{q}_1) \delta(\mathbf{q}_2) \delta(\mathbf{q}_3) - B_{\ell=0}^{\text{SN}} \quad (1)$$

$\delta_D$  above is a Dirac delta function and hence  $\delta_D(\mathbf{q}_{123}) = \delta_D(\mathbf{q}_1 + \mathbf{q}_2 + \mathbf{q}_3)$  ensures that the  $\mathbf{q}_i$  triplet actually form a closed triangle. Each of the integrals above represent an integral over a spherical shell in  $k$ -space with radius  $\delta k$  centered at  $\mathbf{k}_i$  — *i.e.*

$$\int_{k_i} d^3 q \equiv \int_{k_i - \delta k/2}^{k_i + \delta k/2} dq q^2 \int d\Omega. \quad (2)$$



**Figure 3.** The redshift-space halo bispectrum,  $\hat{B}_0(k_1, k_2, k_3)$ , as a function of triangle configurations for  $M_\nu = 0.0, 0.06, 0.10$ , and  $0.15$  eV (top panel) and  $M_\nu = 0.0$  eV,  $\sigma_8 = 0.822, 0.818, 0.807$ , and  $0.798$  (lower panel). We include all possible triangle configurations with  $k_1, k_2, k_3 \leq k_{\max} = 0.5$   $h/\text{Mpc}$  where we order the configurations by looping through  $k_3$  in the inner most loop and  $k_1$  in the outer most loop satisfying  $k_1 \leq k_2 \leq k_3$ . In the insets of the panels we zoom into triangle configurations with  $k_1 = 0.113$ ,  $0.226 \leq k_2 \leq 0.283$ , and  $0.283 \leq k_3 \leq 0.377$   $h/\text{Mpc}$ .

$V_B$  is a normalization factor proportional to the number of triplets  $\mathbf{q}_1$ ,  $\mathbf{q}_2$ , and  $\mathbf{q}_3$  that can be found in the triangle bin defined by  $k_1$ ,  $k_2$ , and  $k_3$  with width  $\delta k$ :

$$V_B = \int_{k_1} d^3 q_1 \int_{k_2} d^3 q_2 \int_{k_3} d^3 q_3 \delta_D(\mathbf{q}_{123}) \quad (3)$$

Lastly,  $B_{\ell=0}^{\text{SN}}$  is the correction for the Poisson shot noise, which contributes due to the self-correlation of individual objects:

$$B_{\ell=0}^{\text{SN}}(k_1, k_2, k_3) = \frac{1}{\bar{n}} (P_0(k_1) + P_0(k_2) + P_0(k_3)) + \frac{1}{\bar{n}^2}. \quad (4)$$

$\bar{n}$  is the number density of objects (halos) and  $P_0$  is the powerspectrum monopole.

In order to evaluate the integrals in Eq. 1, we take advantage of the plane-wave representation of the Dirac delta function and rewrite the equation as

$$\hat{B}_{\ell=0}(k_1, k_2, k_3) = \frac{1}{V_B} \int \frac{d^3 x}{(2\pi)^3} \int_{k_1} d^3 q_1 \int_{k_2} d^3 q_2 \int_{k_3} d^3 q_3 \delta(\mathbf{q}_1) \delta(\mathbf{q}_2) \delta(\mathbf{q}_3) e^{i\mathbf{q}_{123} \cdot \mathbf{x}} - B_{\ell=0}^{\text{SN}} \quad (5)$$

$$= \frac{1}{V_B} \int \frac{d^3 x}{(2\pi)^3} \prod_{i=1}^3 I_{k_i}(\mathbf{x}) - B_{\ell=0}^{\text{SN}} \quad (6)$$

where

$$I_{k_i}(\mathbf{x}) = \int_k d^3q \delta(\mathbf{q}) e^{i\mathbf{q}\cdot\mathbf{x}}. \quad (7)$$

At this point, we measure  $\widehat{B}_{\ell=0}(k_1, k_2, k_3)$  by calculating the  $I_{k_i}$ s with inverse FFTs and summing over in real space.<sup>2</sup> For  $\widehat{B}_{\ell=0}$  measurements throughout the paper, we use  $\delta(\mathbf{x})$  grids with  $N_{\text{grid}} = 360$  and triangle configurations defined by  $k_1, k_2, k_3$  bins of width  $\Delta k = 3k_f = 0.01885 h/\text{Mpc}$ .

We present the redshift-space halo bispectrum of the HADE simulations measured using the estimator above in two ways: one that emphasizes the triangle shape dependence (Figure 2) and the other that emphasizes the amplitude (Figure 3). In Figure 2, we plot  $\widehat{B}_0(k_1, k_2, k_3)$  as a function of  $k_2/k_1$  and  $k_3/k_1$ , which describe the triangle configuration shape. In each panel, the colormap in each  $(k_2/k_1, k_3/k_1)$  bin is the weighted average  $\widehat{B}_0$  amplitude of all triangle configurations in the bin. The upper left bins contain squeezed triangles ( $k_1 = k_2 \gg k_3$ ); the upper right bins contain equilateral triangles ( $k_1 = k_2 = k_3$ ); and the bottom center bins contain folded triangles ( $k_1 = 2k_2 = 2k_3$ ). We include all possible triangle configurations with  $k_1, k_2, k_3 < k_{\text{max}} = 0.5 h/\text{Mpc}$ . The  $\widehat{B}_0$  in the upper panels are HADES models with  $M_\nu = 0.0$  (fiducial), 0.06, 0.10, and 0.15 eV;  $\widehat{B}_0$  in the lower panels are HADES models with  $M_\nu = 0.0$  eV and  $\sigma_8 = 0.822, 0.818$ , and 0.807. The top and bottom panels of the three right-most columns have matching  $\sigma_8$  values (Section 2).

Next, in Figure 3, we plot  $\widehat{B}_0(k_1, k_2, k_3)$  for all possible triangle configurations with  $k_1, k_2, k_3 < k_{\text{max}} = 0.5 h/\text{Mpc}$  where we order the configurations by looping through  $k_3$  in the inner most loop and  $k_1$  in the outer most loop with  $k_1 \leq k_2 \leq k_3$ . In the top panel, we present  $\widehat{B}_0$  of HADES models with  $M_\nu = 0.0, 0.06, 0.10$ , and 0.15 eV; in the lower panel, we present  $\widehat{B}_0$  of HADES models with  $M_\nu = 0.0$  eV and  $\sigma_8 = 0.822, 0.818$ , and 0.807. We zoom into triangle configurations with  $k_1 = 0.113, 0.226 \leq k_2 \leq 0.283$ , and  $0.283 \leq k_3 \leq 0.377 h/\text{Mpc}$  in the insets of the panels.

## 4. RESULTS

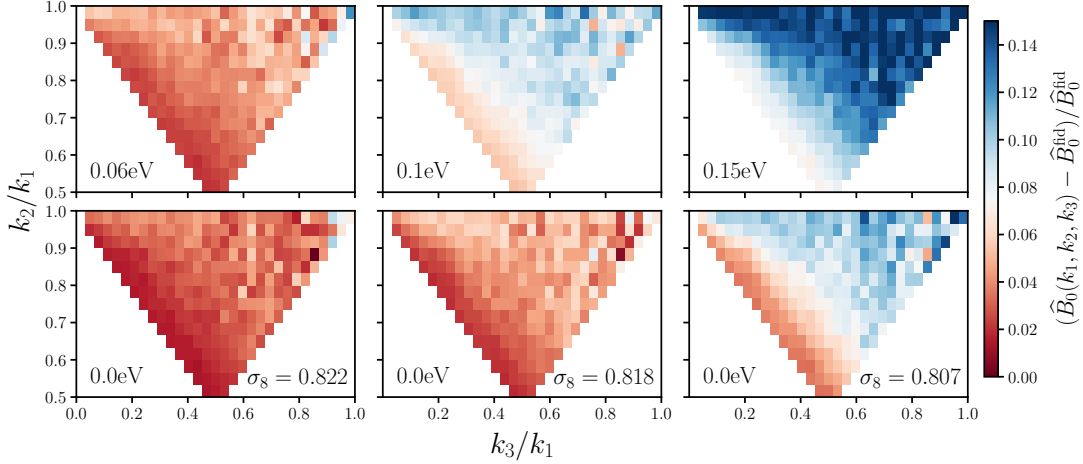
### 4.1. Breaking the $M_\nu - \sigma_8$ degeneracy

One major bottleneck of constraining  $M_\nu$  with the power spectrum alone is the strong  $M_\nu - \sigma_8$  degeneracy. The imprint of  $M_\nu$  and  $\sigma_8$  on the power spectrum are degenerate and for models with the same  $\sigma_8^c$ , the power spectrum only differ by  $< 1\%$  (see Figure 1 and Villaescusa-Navarro et al. 2018). The HADES suite, which has simulations with  $M_\nu = 0.0, 0.06, 0.10$ , and 0.15 eV as well as  $M_\nu = 0.0$  eV simulations with matching  $\sigma_8^c - \sigma_8 = 0.822, 0.818$ , and 0.807, provide an ideal set of simulations to separate the impact of  $M_\nu > 0.0$  eV and examine the degeneracy between  $M_\nu$  and  $\sigma_8$  (Section 2 and Table 1). Hence, by measuring bispectrum of these simulations (Figure 2 and 3), we can determine whether the bispectrum helps break the  $M_\nu - \sigma_8$  degeneracy. Below, we present our comparison of the HADES bispectrum and illustrate that the bispectrum can significantly improve  $M_\nu$  constraints by breaking the  $M_\nu - \sigma_8$  degeneracy.

We begin by examining the triangle shape dependent imprint of  $M_\nu$  on the redshift-space halo bispectrum versus  $\sigma_8$  alone. In Figure 4, we present the fractional residual,  $(\Delta \widehat{B}_0 = \widehat{B}_0 - \widehat{B}_0^{\text{fid}})/\widehat{B}_0^{\text{fid}}$ ,

<sup>2</sup> The code that we use to evaluate  $\widehat{B}_{\ell=0}$  is publicly available at <https://github.com/changhoonhahn/pySpectrum>



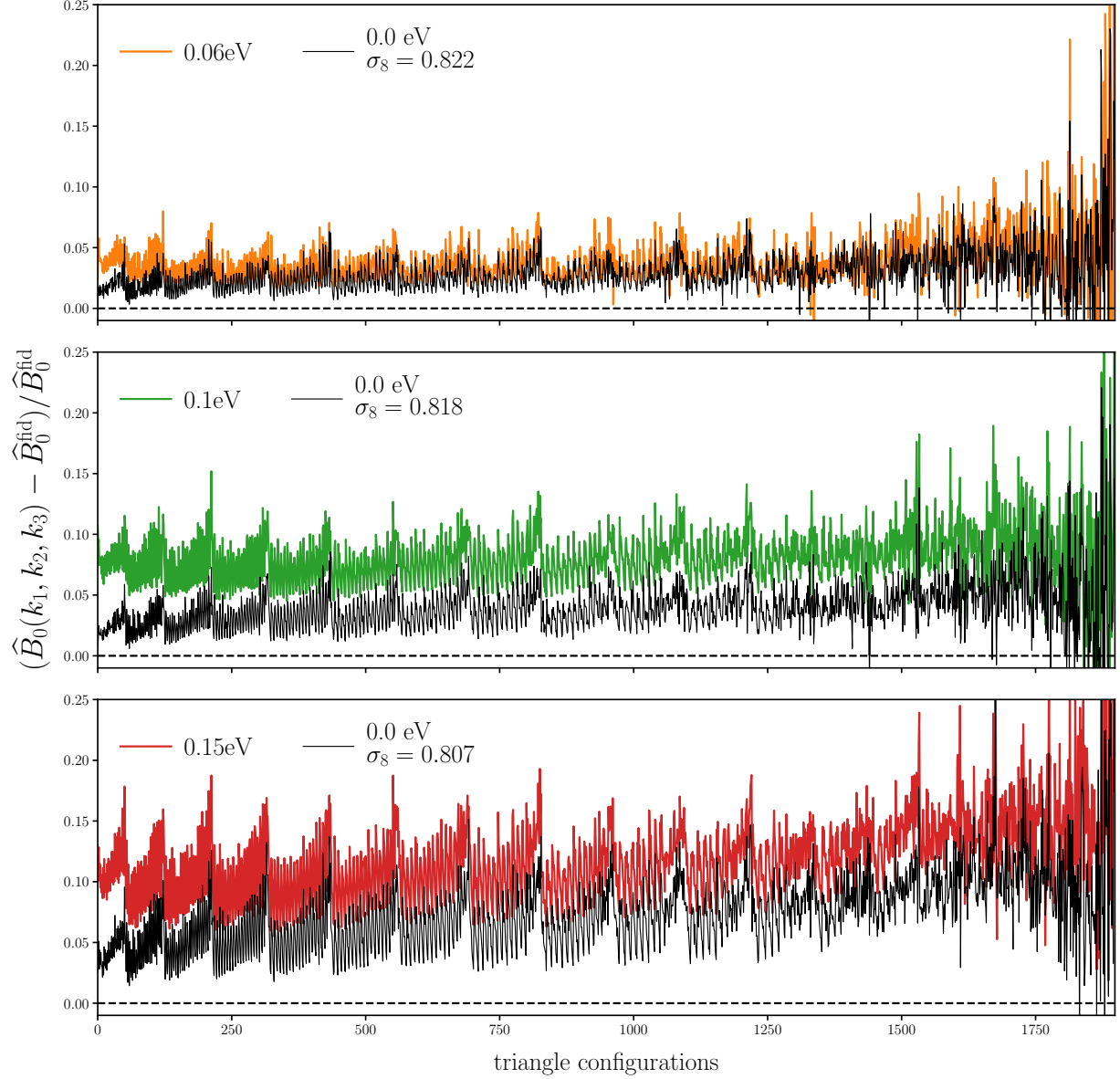


**Figure 4.** The shape dependence of the  $M_\nu$  and  $\sigma_8$  imprint on the redshift-space halo bispectrum,  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$ . We align the  $M_\nu = 0.06, 0.10$ , and  $0.15$  eV HADES models in the upper panels with  $M_\nu = 0.0$  eV  $\sigma_8 = 0.822, 0.818$ , and  $0.807$  models on the bottom such that the top and bottom panels in each column have matching  $\sigma_8^c$ , which produce mostly degenerate imprints on the redshift-space power spectrum. The difference between the top and bottom panels highlight that  $M_\nu$  leaves a distinct imprint on elongated and isosceles triangles (bins along the bottom left and bottom right edges, respectively) from  $\sigma_8$ . *The imprint of  $M_\nu$  has an overall distinct shape dependence on the bispectrum that cannot be replicated by varying  $\sigma_8$ .*

as a function of  $k_2/k_1$  and  $k_3/k_1$  for  $M_\nu = 0.06, 0.10$ , and  $0.15$  eV in the upper panels and  $0.0$  eV  $\sigma_8 = 0.822, 0.818$ , and  $0.807$  in the bottom panels. The simulations in the top and bottom panels of each column have matching  $\sigma_8^c$ . Overall as  $M_\nu$  increases, the bispectrum increases for all triangle shapes (top panels). This increase is due to halo bias (Villaescusa-Navarro et al. 2018, ; see also Figure 1). We impose a fixed  $M_{\text{lim}}$  on our halos so lower values of  $\sigma_8$  translate to a larger halo bias, which boosts the amplitude of the bispectrum. Within the overall increase in amplitude, however, equilateral triangles (upper left) have the largest increase. For  $M_\nu = 0.15$  eV, the bispectrum is  $\sim 15\%$  higher than  $\hat{B}_0^{\text{fid}}$  for equilateral triangles. Meanwhile, the bispectrum increases by  $\sim 8\%$  for folded triangles for  $0.15$  eV (lower center). The noticeable difference in  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  between equilateral and squeezed triangles (upper left) is roughly consistent with the comparison in Figure 7 of Ruggeri et al. (2018). They, however, fix  $A_s$  in their simulations and measure the real-space halo bispectrum so we refrain from any detailed comparisons.

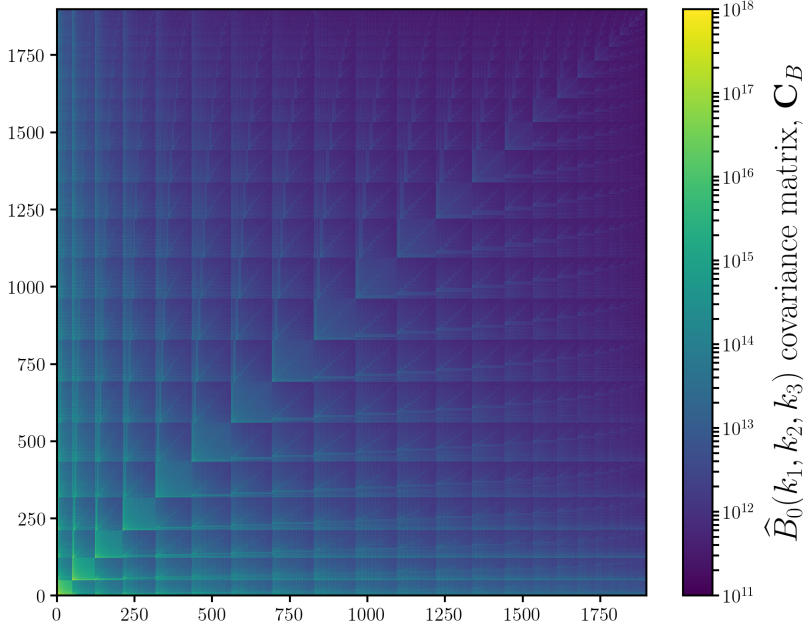
As  $\sigma_8$  increases, with  $M_\nu = 0.0$  eV fixed, the bispectrum increases overall for all triangle shapes (bottom panels). However, the comparison of the top and bottom panels in each column reveals significant differences in  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  for  $M_\nu$  versus  $\sigma_8$  alone. Between  $M_\nu = 0.15$  eV and  $0.0$  eV  $\sigma_8 = 0.807$ , there is an overall  $\gtrsim 5\%$  difference. In addition, the shape dependence of the  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  increase is different for  $M_\nu$  than  $\sigma_8$ . This is particularly clear in the differences between  $0.1$  eV (top center panel) and  $0.0$  eV and  $\sigma_8 = 0.807$  (bottom right panel): near equilateral triangles in the two panels have similar  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  while triangle shapes near the lower left edge from the squeezed to folded triangles have significantly different  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$ . Hence,  $M_\nu$  leaves an imprint on the bispectrum





**Figure 5.** The impact of  $M_\nu$  and  $\sigma_8$  on the redshift-space halo bispectrum,  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$ , for all 1898 triangle configurations with  $k_1, k_2, k_3 \leq 0.5h/\text{Mpc}$ . We compare  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  of the  $M_\nu = 0.06$  (top),  $0.10$  (middle), and  $0.15$  eV (bottom) HADES models to  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  of  $M_\nu = 0.0$  eV  $\sigma_8 = 0.822, 0.818$ , and  $0.807$  models. The impact of  $M_\nu$  on the bispectrum has a significantly different amplitude than the impact of  $\sigma_8$ . For instance,  $M_\nu = 0.15$  eV (red) has a  $\sim 5\%$  stronger impact on the bispectrum than  $M_\nu = 0.0$  eV  $\sigma_8 = 0.798$  (black) even though their powerspectrums only differ by  $< 1\%$  (Figure 1). Combined with the different shape-dependence (Figure 4), the distinct imprint of  $M_\nu$  on the bispectrum illustrate that the bispectrum can break the degeneracy between  $M_\nu$  and  $\sigma_8$  that degrade constraints from two-point analyses.

with a distinct triangle shape dependence than  $\sigma_8$  alone. In other words, unlike the power spectrum, the triangle shape dependent impact of  $M_\nu$  on the bispectrum cannot be replicated by varying  $\sigma_8$ .

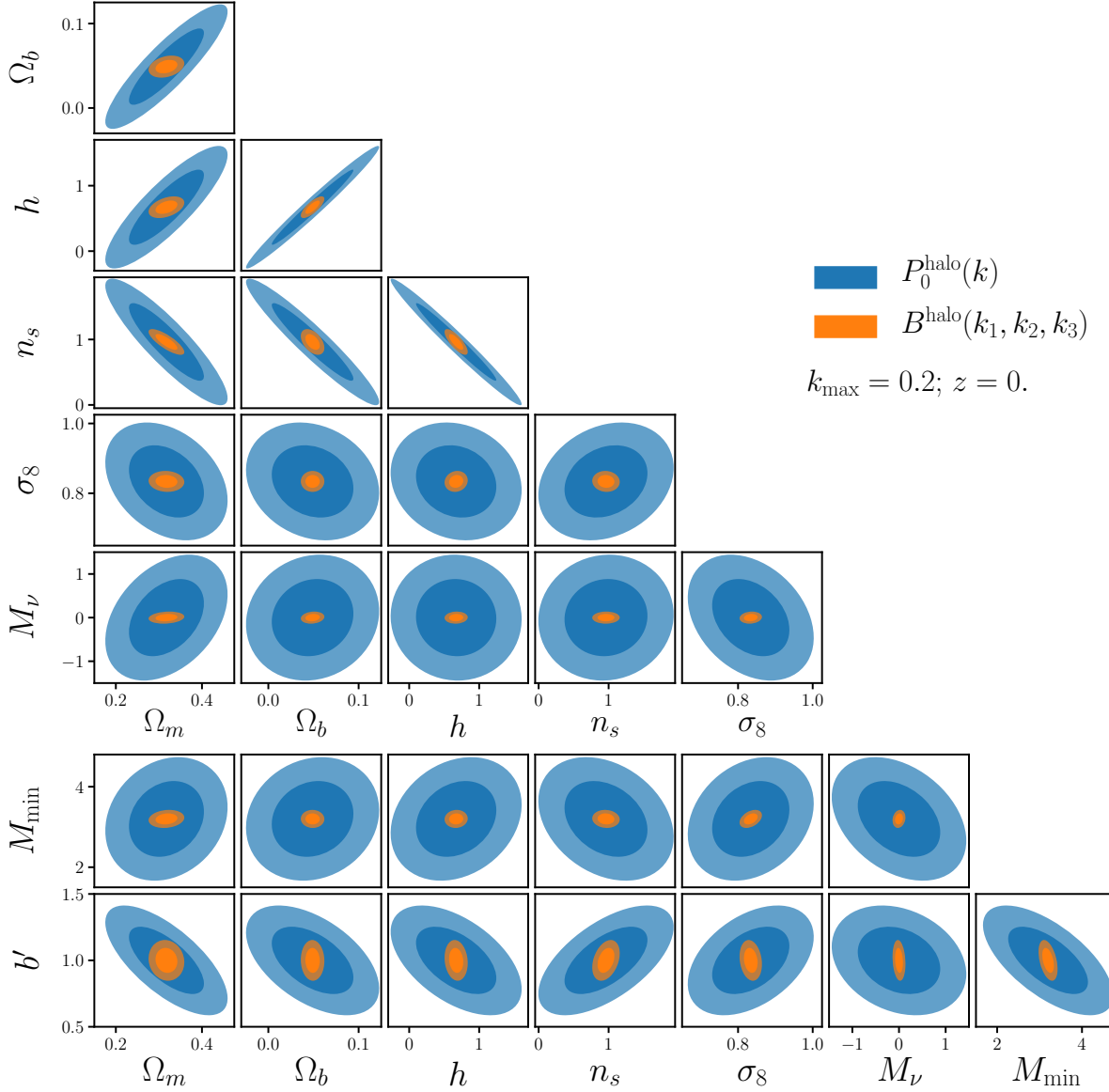


**Figure 6.** Covariance matrix of the redshift-space halo bispectrum estimated using the 15,000 realizations of the Quijote simulation suite at the fiducial cosmology:  $\Omega_m=0.3175$ ,  $\Omega_b=0.049$ ,  $h=0.6711$ ,  $n_s=0.9624$ ,  $\sigma_8=0.834$ , and  $M_\nu=0.0$  eV. We include all possible triangle configurations with  $k_1, k_2, k_3 < k_{\text{max}} = 0.5 h/\text{Mpc}$  and order the configurations (bins) in the same way as Figures 3 and 5. We use the covariance matrix above for the Fisher matrix forecasts presented in Section 4.2.

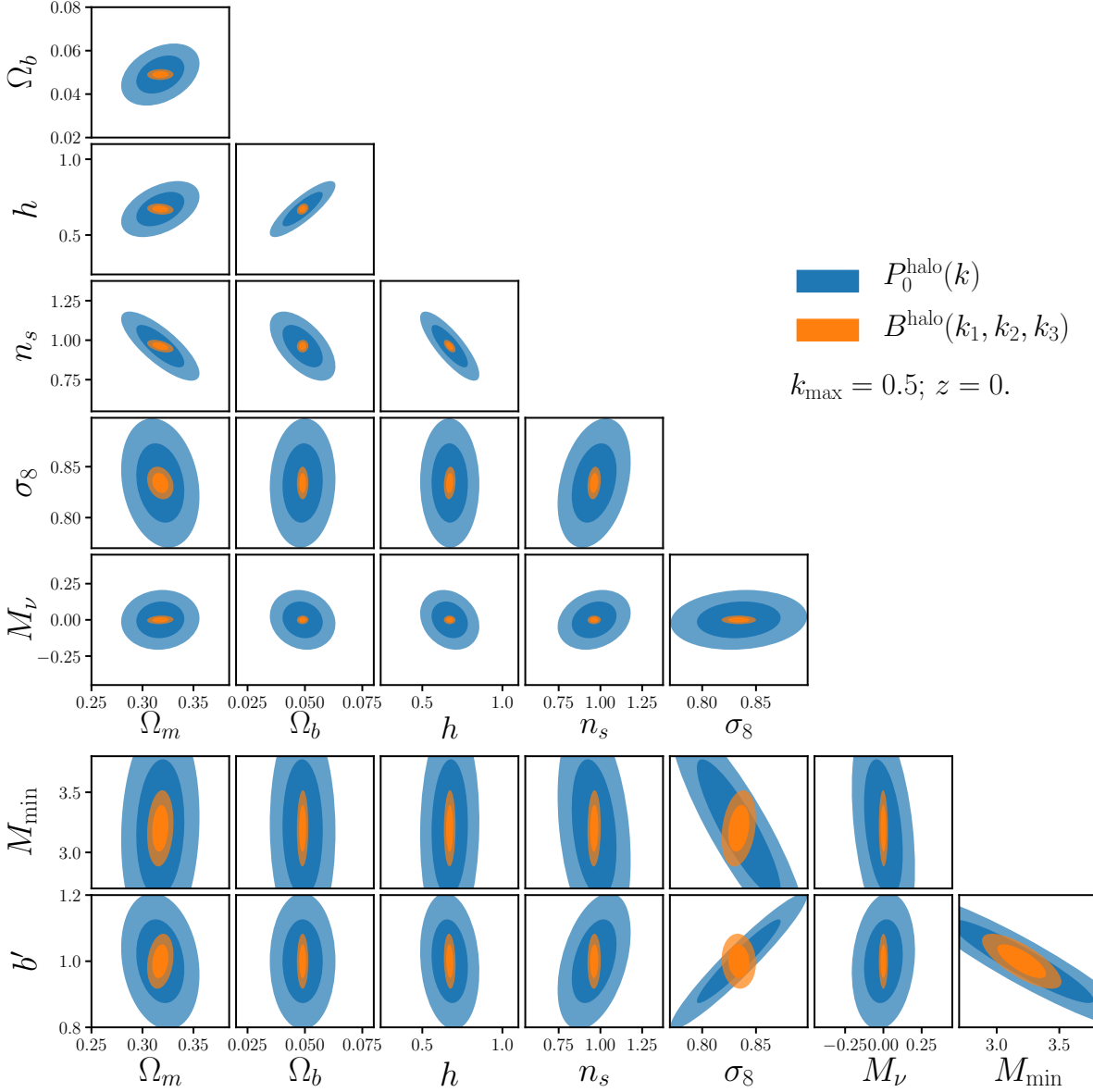
We next examine the amplitude of the  $M_\nu$  imprint on the redshift-space halo bispectrum versus  $\sigma_8$  alone for all triangle configurations. We present  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  for all 1898 possible triangle configurations with  $k_1, k_2, k_3 < k_{\text{max}} = 0.5 h/\text{Mpc}$  in Figure 5. We compare  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  of the  $M_\nu = 0.06, 0.10$ , and  $0.15$  eV HADES models to the  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  of  $M_\nu = 0.0$  eV  $\sigma_8 = 0.822, 0.818$ , and  $0.807$  models in the top, middle, and bottom panels, respectively. The comparison confirms the difference in overall amplitude of varying  $M_\nu$  and  $\sigma_8$  (Figure 4). For instance,  $M_\nu = 0.15$  eV (red) has a  $\sim 5\%$  stronger impact on the bispectrum than  $M_\nu = 0.0$  eV  $\sigma_8 = 0.798$  (black) even though their power spectra differ by  $< 1\%$  (Figure 1).

The comparison in the panels of Figure 5 also reveal a difference in the configuration dependence in  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  between  $M_\nu$  versus  $\sigma_8$ . The triangle configurations are ordered by looping through  $k_3$  in the inner most loop and  $k_1$  in the outer most loop such that  $k_1 \leq k_2 \leq k_3$ . In this ordering,  $k_1$  increases from left to right.  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  of  $M_\nu$  expectedly increases with  $k_1$ : for small  $k_1$  (on large scales), neutrinos behave like CDM and therefore the impact is reduced. However,  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  of  $M_\nu$  has a smaller  $k_1$  dependence than  $\Delta\hat{B}_0/\hat{B}_0^{\text{fid}}$  of  $\sigma_8$ . Combined with the different shape-dependence (Figure 4), the distinct imprint of  $M_\nu$  on the redshift-space halo bispectrum illustrates that the bispectrum can break the degeneracy between  $M_\nu$  and  $\sigma_8$ . Moreover it illustrates that by including the bispectrum, we can more precisely constrain  $M_\nu$  than with the power spectrum alone.

#### 4.2. $M_\nu$ and other Cosmological Parameter Forecasts



**Figure 7.** Fisher matrix constraints for  $M_\nu$  and other cosmological parameters for the redshift-space halo bispectrum monopole (orange). For comparison, we include Fisher parameter constraints for the redshift-space halo powerspectrum monopole in blue. The contours mark the 68% and 95% confidence intervals. We set  $k_{\text{max}} = 0.2 \, h/\text{Mpc}$  for both power spectrum and bispectrum. We include in our forecasts  $b'$  and  $M_{\min}$ , a free amplitude scaling factor and halo mass limit, respectively. They serve as a simplistic bias model and we marginalize over them so that our constraints do not include extra constraining power from the difference in bias/number density in the different Quijote cosmologies. The bispectrum *substantially* improves constraints on all of the cosmological parameters over the power spectrum. *For  $M_\nu$ , the bispectrum improves the constraint from  $\sigma_{M_\nu} = 1.163$  to  $0.112 \, \text{eV}$  — an order of magnitude improvement over the power spectrum.*



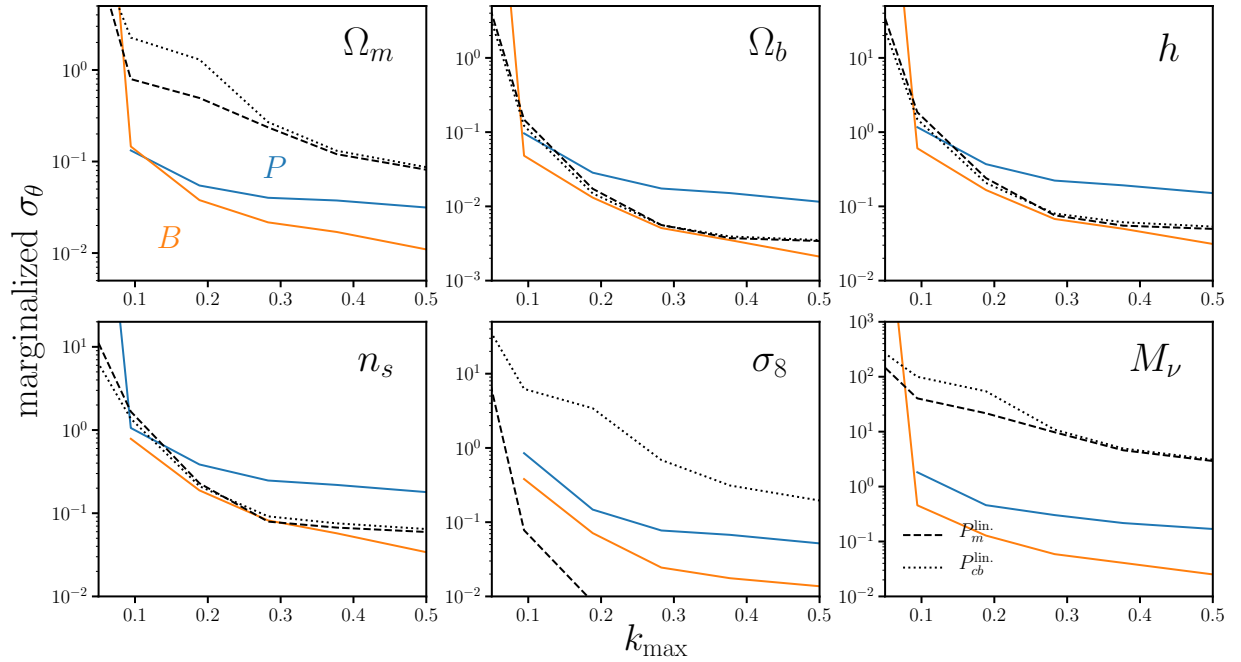
**Figure 8.** Same as Figure ?? but for  $k_{\text{max}} = 0.5 \text{ h/Mpc}$ . Again, the bispectrum substantially improves constraints on all cosmological parameters over the power spectrum. For  $M_\nu$ , the bispectrum improves  $\sigma_{M_\nu}$  from 0.319 to 0.0239 eV — over an order of magnitude improvement over the power spectrum.

We demonstrate in the previous section with the HADES simulations, that the bispectrum helps break the  $M_\nu$ – $\sigma_8$  degeneracy, a major challenge in precisely constraining  $M_\nu$  with the power spectrum. This establishes the bispectrum as a promising probe for  $M_\nu$ . However, we are ultimately interested in determining the constraining power of the bispectrum for an analysis that include cosmological parameters beyond  $M_\nu$  and  $\sigma_8$ — *i.e.*  $\Omega_m$ ,  $\Omega_b$ ,  $h$ , and  $n_s$ . The Quijote suite of simulations is *specifically* designed to answer this question using Fisher matrix forecast.

First, the Quijote suite includes 15,000 realizations run at a fiducial cosmology:  $M_\nu=0.0\text{eV}$ ,  $\Omega_m=0.3175$ ,  $\Omega_b=0.049$ ,  $n_s=0.9624$ ,  $h=0.6711$ , and  $\sigma_8 = 0.834$  (see Table 1. This allows us to robustly

**Table 2.** Marginalized Fisher parameter constraints from the redshift-space halo power spectrum (top) and bispectrum (bottom) for different values of  $k_{\text{max}}$ . We list constraints for cosmological parameters  $M_\nu$ ,  $\Omega_m$ ,  $\Omega_b$ ,  $h$ ,  $n_s$ , and  $\sigma_8$  as well as nuisance parameters  $b'$  and  $M_{\text{min}}$

	$k_{\text{max}}$ ( $h/\text{Mpc}$ )	$M_\nu$ (eV)	$\Omega_m$	$\Omega_b$	$h$	$n_s$	$\sigma_8$	$b'$	$M_{\text{min}}$ ( $10^{13}h^{-1}M_\odot$ )
		0.0	0.3175	0.049	0.6711	0.9624	0.834	1.	3.2
$P_0$	0.2	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$
	0.3	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$
	0.4	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$
	0.5	$\pm 0.319$	$\pm 0.0444$	$\pm 0.0221$	$\pm 0.294$	$\pm 0.315$	$\pm 0.0430$	$\pm 0.35$	$\pm 0.13$
$B_0$	0.2	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$
	0.3	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$
	0.4	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$	$\pm$
	0.5	$\pm 0.0239$	$\pm 0.0103$	$\pm 0.00202$	$\pm 0.0295$	$\pm 0.0321$	$\pm 0.0125$	$\pm 0.046$	$\pm 0.047$



**Figure 9.** Marginalized  $1\sigma$  constraints of  $\Omega_m$ ,  $\Omega_b$ ,  $h$ ,  $n_s$ ,  $\sigma_8$ , and  $M_\nu$  as a function of  $k_{\text{max}}$  for the redshift-space halo bispectrum (orange) and power spectrum (blue). Though not included in the figure, we marginalize over  $b'$  and  $M_{\text{min}}$  in our forecast. **CH: more**

estimate the covariance matrix of the bispectrum,  $\mathbf{C}$ , which has  $\sim 1,800$  triangle configurations (Figure 6). Second, the Quijote suite includes 500 fixed-pair realizations evaluated at 13 different cosmologies, each a small step away from the fiducial cosmology parameter values along one parameter (Section 2 and Table 1). These realizations allow us to precisely estimate the derivatives of the bispectrum with respect to each of the cosmological parameters.

Since their introduction to cosmology over two decades ago, Fisher Information matrices have been ubiquitously used to forecast the constraining power of future experiments (*e.g.* Jungman et al. 1996; Tegmark et al. 1997; Dodelson 2003; Heavens 2009; Verde 2010). Defined as

$$F_{ij} = -\left\langle \frac{\partial^2 \ln \mathcal{L}}{\partial \theta_i \partial \theta_j} \right\rangle, \quad (8)$$

where  $\mathcal{L}$  is the likelihood, the Fisher matrix for the bispectrum can be written as

$$F_{ij} = \frac{1}{2} \text{Tr} \left[ \mathbf{C}^{-1} \frac{\partial \mathbf{C}}{\partial \theta_i} \mathbf{C}^{-1} \frac{\partial \mathbf{C}}{\partial \theta_j} + \mathbf{C}^{-1} \left( \frac{\partial \bar{B}_0}{\partial \theta_i} \frac{\partial \bar{B}_0}{\partial \theta_j}^T + \frac{\partial \bar{B}_0}{\partial \theta_i}^T \frac{\partial \bar{B}_0}{\partial \theta_j} \right) \right]. \quad (9)$$

Since we assume that the  $B_0$  likelihood is Gaussian, including the first term in Eq. 9 runs the risk of incorrectly including information from the covariance already included in the mean (Carron 2013). We, therefore, conservatively neglect the first term and calculate the Fisher matrix,

$$F_{ij} = \frac{1}{2} \text{Tr} \left[ \mathbf{C}^{-1} \left( \frac{\partial \bar{B}_0}{\partial \theta_i} \frac{\partial \bar{B}_0}{\partial \theta_j}^T + \frac{\partial \bar{B}_0}{\partial \theta_i}^T \frac{\partial \bar{B}_0}{\partial \theta_j} \right) \right], \quad (10)$$

directly with  $\mathbf{C}$  and  $\partial B_0 / \partial \theta_i$  along each cosmological parameter from the Quijote simulations. For  $\Omega_m$ ,  $\Omega_b$ ,  $h$ ,  $n_s$ , and  $\sigma_8$ , we estimate

$$\frac{\partial \bar{B}_0}{\partial \theta_i} \approx \frac{\bar{B}_0(\theta_i^+) - \bar{B}_0(\theta_i^-)}{\theta_i^+ - \theta_i^-}. \quad (11)$$

Meanwhile, for  $M_\nu$ , where the fiducial value is 0.0 eV and we cannot have negative  $M_\nu$ , we use the Quijote simulations at  $M_\nu^+$ ,  $M_\nu^{++}$ ,  $M_\nu^{+++} = 0.1, 0.2, 0.4$  eV (Table 1) to estimate:

$$\frac{\partial \bar{B}_0}{\partial M_\nu} \approx \frac{-21\bar{B}_0(M_\nu^{\text{fid}}) + 32\bar{B}_0(M_\nu^+) - 12\bar{B}_0(M_\nu^{++}) + \bar{B}_0(M_\nu^{+++})}{1.2}. \quad (12)$$

By using these  $N$ -body simulations, instead of analytic methods (*e.g.* perturbation theory), we exploit the accuracy of numerical simulations in the nonlinear regime and rely on fewer assumptions and approximations. **CH: something about quantity the information content of the bispectrum in the nonlinear regime unprecedented.** We discuss subtleties of the bispectrum derivatives in Appendix B and discuss tests of convergence in Appendix B.

We present the constraints on  $M_\nu$  and other cosmological parameters  $\{\Omega_m, \Omega_b, h, n_s, \sigma_8\}$  derived from the redshift-space halo bispectrum Fisher matrix (Eq. 10) for  $k_{\text{max}} = 0.2$  and  $0.5 h/\text{Mpc}$  in Figures 7 and 8, respectively. We include Fisher constraints for the redshift-space halo power spectrum monopole, for the same  $k_{\text{max}}$  in blue for comparison. The contours mark the 68% and 95%

confidence intervals. We include in our Fisher constraints the following nuisance parameters:  $b'$ , a scaling factor on the bispectrum amplitude, and  $M_{\min}$ , the halo mass limit.  $b'$  is analogous to linear bias. Meanwhile, we choose  $M_{\min}$  as a nuisance parameter to address the difference in the number densities among the Quijote cosmologies, which impacts the derivatives  $\partial \bar{B}_0 / \partial \theta_i$ . For instance, the  $\sigma_8^+$  and  $\sigma_8^-$  cosmologies have halo  $\bar{n} = 1.586 \times 10^{-4}$  and  $1.528 \times 10^{-4} (h^{-1} \text{Mpc})^3$ . These parameters serve as a simplistic bias model and by marginalizing over them we aim to ensure that our Fisher constraints do not include extra constraining power from the difference in bias or number density.  $b'$  is a multiplicative factor so  $\partial \bar{B}_0 / \partial b' = \bar{B}_0$ . Meanwhile, we numerically estimate  $\partial \bar{B}_0 / \partial M_{\min}$  using  $\bar{B}_0$  evaluated at  $M_{\min}=3.3$  and  $3.1 \times 10^{13} h^{-1} M_{\odot}$ , with all other parameters set to the fiducial value.

The bispectrum substantially improves constraints on all of the parameters over the power spectrum (Figure ??). More precisely, the bispectrum improves the marginalized  $\Omega_m$ ,  $\Omega_b$ ,  $h$ ,  $n_s$ , and  $\sigma_8$  constraints by factors of **CH:  $\sim 4, 11, 10, 10$ , and  $3$**  with respect to their power spectrum constraints. *For  $M_\nu$ , the bispectrum improves the constraint from  $\sigma_{M_\nu} = 0.319$  to  $0.0239 \text{ eV}$  — over an order of magnitude improvement over the power spectrum.*

Even at lower  $k_{\max} < 0.5 \text{ h/Mpc}$ , the bispectrum significantly improves cosmological parameter constraints. We compare the marginalized  $1\sigma$  constraints of  $\Omega_m$ ,  $\Omega_b$ ,  $h$ ,  $n_s$ ,  $\sigma_8$ , and  $M_\nu$  as a function of  $k_{\max}$  for  $B_0$  (orange) and  $P_0$  (blue) in Figure 9. **CH: write this once we have to updated powerspectrum calculations.**

In Table 2, we list the marginalized Fisher parameter constraints of both cosmological and nuisance parameters for  $P_0(k \leq k_{\max} = 0.5 \text{ h/Mpc})$  and  $B_0(k_1, k_2, k_3 \leq k_{\max} = 0.5 \text{ h/Mpc})$ .

discussing the results

- emphasize that this is only for a 1 Gpc box so our constraints will be so much better
- redshift-space helps us out (refer to appendix)

Q: Is this consistent with S/N calculations? For marginalized constraints The unmarginalized constraints are consistent with S/N calculations of scoccimarro and more recently chan.

Q: How does the bispectrum do so much better? A: The bispectrum does better than the power spectrum because it's better at breaking parameter degeneracies. In fact, P has higher Fii terms than B – i.e. the unmarginalized constraints are tighter for P than B. However, once you marginalize, this goes away. One reason for this is that unlike other parameters such as  $\sigma_8$  and  $\Omega_m$ , the response of the B,  $(\partial \log B / \partial M_\nu)_0$  shows very little triangle shape dependence. Show figure or not?

The results above definitively show that the bispectrum has significant constraining power in the weakly nonlinear regime ( $k > 0.1 \text{ h/Mpc}$ ) beyond the power spectrum. Our results also demonstrated the potential of the bispectrum in constraining  $M_\nu$  (an order of magnitude improvement over  $P_0$ ). Below, we underline a few caveats of the results we present above. First, the parameter constraints were derived using the Fisher matrix. This assumes

- convergence concerns (refer to appendix)
- $M_\nu$  dependence of derivatives along  $M_\nu$ . (refer to appendix)
- take Fisher forecasts with a grain of salt



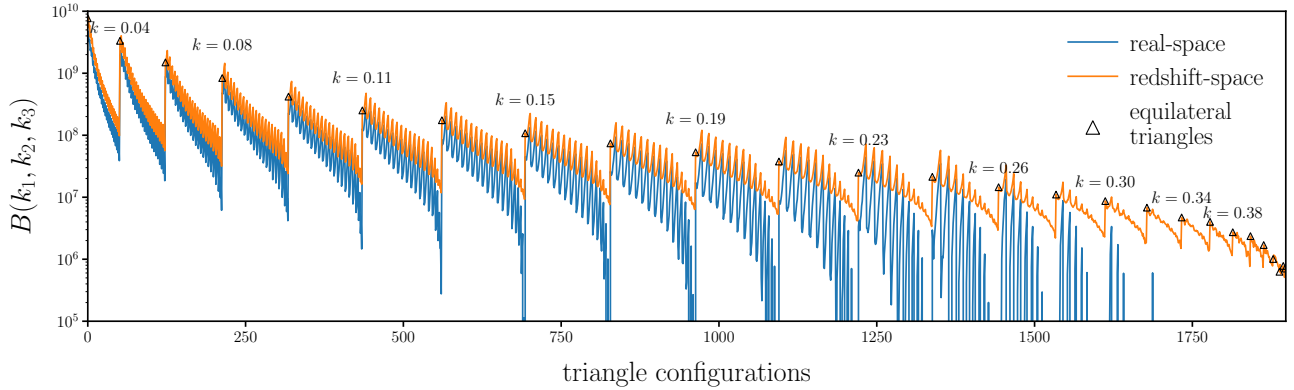
Q: How robust are Fisher forecasts A: There are many caveats to the Fisher forecast. One of the main limitation is that Fisher forecasts assume the posteriors are Gaussian

Concern: A lot of the advantages from breaking parameter degeneracies. How robust is this? Answer: We address this by imposing Planck priors on our forecasts. This expectedly has a large impact on P constraints, but doesn't change B constraints much. B still does better by a lot. This suggests that the breaking of parameter degeneracies are relatively stable to poorly constrained derivatives spuriously breaking degeneracies so this is robust. Also demonstrates that the gains are relatively stable to Fisher forecasts caveats (i.e. projecting a banana).

Concern: extra information leaking in due to non-linear bias and number density differences? Answer: We ran tests that include SN correction parameters in case number density information is leaking in from there. Constraints barely change. We also include and marginalize over  $b_2$  and  $g_2$  but nothing changes. Explain fixed  $\bar{n}$  tests that we ran, which also came out short. Talk about how these tests don't definitely answer whether non-linear bias information is leaking into the forecasts since  $b'$ ,  $M_{\min}$ ,  $b_2$ , and  $g_2$  may not a sufficient model of halo bias. We suspect this is the reason why B does better than P for all  $k_{\max}$  even though on linear scales we expect P to have more constraining power. However since the ultimate goal is to constrain  $M_\nu$  with the galaxy bispectrum. For galaxy bispectrum, HOD is a more robust bias model. The halo bispectrum is an intermediate step and a proof of concept, which our results show is promising. We therefore refrain from an exhaustive investigation of halo bias and go directly to HOD in the next paper of this series.

Another caveat is that our parameter constraints were derived using the power spectrum and bispectrum of halo in a periodic box. We do not consider a realistic survey geometry or radial selection function. A realistic selection function will smooth out the triangle configuration dependence and consequently degrade the constraining power of the bispectrum. In [Sefusatti & Scoccimarro \(2005\)](#), for instance, they find that the signal-to-noise of the bispectrum is significantly reduced once survey geometry is included in their forecast. Survey geometry, however, also degrade the signal-to-noise of their power spectrum forecasts. Hence, with the order of magnitude improve in the  $M_\nu$  constraining power of the bispectrum, even with survey geometry, including the bispectrum will improve  $M_\nu$  constraints.

Although we focus on the halo bispectrum and power spectrum in this paper, constraints on  $M_\nu$  will ultimately be derived from the distribution of galaxies. Besides the cosmological parameters, bias and nuisance parameters that allow us to marginalize over the galaxy—halo connection need to be incorporated to forecast  $M_\nu$  and other cosmological parameter constraints for the galaxy bispectrum. Although we include a *naive* bias model through  $b'$  and  $M_{\min}$ , this is insufficient to describe how galaxies occupy halos. A more realistic bias model such as a halo occupation distribution (HOD) model involve extra parameters that describe the distribution of central and satellite galaxies in halos (*e.g.* [Zheng et al. 2005](#); [Leauthaud et al. 2012](#); [Tinker et al. 2013](#); [Zentner et al. 2016](#); [Vakili & Hahn 2019](#)). **CH:** maybe something about Uros and Nick's model involving a lot of parameters. Marginalizing over these extra parameters, will likely reduce the constraining power at high  $k$ . Even if the constraining power at high  $k$  is reduced, the bispectrum still offers significant improvements over the power spectrum at  $k_{\max} \sim 0.2$ . Jointly analyzing power spectrum and bispectrum will help



**Figure 10.** Comparison of the fiducial HADES simulations real and redshift-space halo bispectrum for triangle configurations with  $k_1, k_2, k_3 \leq k_{\max} = 0.5h/\text{Mpc}$  (blue and orange respectively). We mark equilateral triangle configurations (empty triangle marker) along with their side lengths  $k$ .

constrain these extra bias parameters. Furthermore, we again emphasize that the constraints we present in this paper is for a  $1h^{-1}\text{Gpc}$  box. In Hahn et al. (in preparation), we will include a realistic HOD model and quantify the information content and constraining power of a joint galaxy power spectrum and bispectrum analysis.

## 5. SUMMARY

**CH:** talk about DESI, PFS, WFIRST

## ACKNOWLEDGEMENTS

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## APPENDIX

### A. REDSHIFT-SPACE HALO BISPECTRUM

**CH:** do we want to keep this section?

### B. FISHER FORECASTS USING $N$ -BODY SIMULATIONS

The two key elements in calculating the Fisher matrices we use in our forecasts are the bispectrum covariance matrix ( $\mathbf{C}$ ; Figure 6) and the derivatives of the bispectrum along the cosmological and nuisance parameters,  $\partial B_0/\partial\theta_i$  (Section 4.2). We compute both these elements directly using the  $N$ -body simulations of the Quijote suite (Section 2). This takes advantage of the  $N$ -body simulations and allows us to accurately quantify the constraining power of the bispectrum that come from the nonlinear regime. However, to trust our forecast, we must ensure that both  $\mathbf{C}$  has converged and that the numerical calculated  $\partial B_0/\partial\theta_i$  do not introduce any biases. Below, we tests the convergence of  $\mathbf{C}$  and  $\partial B_0/\partial\theta_i$  and discuss some of the subtleties and caveats of numerically calculating  $\partial B_0/\partial\theta_i$  from the Quijote simulations.

To estimate  $\mathbf{C}$ , we use 15,000 Quijote  $N$ -body simulations at the fiducial cosmology. This is a *significantly* larger number of simulations than previous bispectrum analyses; however, we also consider a larger number of triangle configurations — 1898 triangles out to  $k_{\text{max}} = 0.5 \ h/\text{Mpc}$ . For reference, the recent Gil-Marín et al. (2017) analysis used 2048 simulations to estimate the covariance matrix of the bispectrum with 825 configurations. We, therefore, check the convergence of  $\mathbf{C}$  by varying  $N_{\text{fid}}$ , the number of simulations used to estimate  $\mathbf{C}$ , and examining whether this significantly impacts the Fisher parameter constraints. We present  $\sigma_{\theta}(N_{\text{fid}})/\sigma_{\theta}(N_{\text{fid}} = 15,000)$ , the ratio of the  $1\sigma$  Fisher constraint for  $\theta = \Omega_{\text{m}}, \Omega_{\text{b}}, h, n_s, \sigma_8$ , and  $M_{\nu}$  calculated with  $N_{\text{fid}}$  over the constraint calculated with  $N_{\text{fid}} = 15,000$ , as a function of  $N_{\text{fid}}$  (Figure 11 left panel). The  $1\sigma$  Fisher constraints on the parameters vary by  $< 10\%$  for  $N_{\text{fid}} > 7000$ ; in fact, the constraints vary by  $< 1\%$  for  $N_{\text{fid}} > 14,000$ . Hence, we conclude that we have a sufficient number of simulations to estimate the bispectrum  $\mathbf{C}$  and our forecasts are robust to the convergence of  $\mathbf{C}$ .

We estimate  $\partial B_0/\partial\theta_i$  numerically using 13 sets of  $N_{\text{fp}} = 500$  fixed paired simulations (Table 1). To check the convergence of  $\partial B_0/\partial\theta_i$  and its impact on our forecast we present the ratio of the  $1\sigma$  Fisher constraint for  $\theta$  calculated using  $N_{\text{fp}}$  simulations over the constraint calculated with  $N_{\text{fp}} = 500$ ,  $\sigma_{\theta}(N_{\text{fp}})/\sigma_{\theta}(N_{\text{fp}} = 500)$ , as a function of  $N_{\text{fp}}$  (Figure 11 right panel). Unlike  $\sigma_{\theta}(N_{\text{fid}})$ ,  $\sigma_{\theta}(N_{\text{fp}})$  depend significantly on  $\theta$ . For instance,  $\sigma_{\theta}$  for  $\sigma_8$  and  $\Omega_{\text{m}}$  vary by  $< 10\%$  for  $N_{\text{fp}} > 300$  and  $< 2\%$  for  $N_{\text{fp}} > 450$ .  $\sigma_{\theta}$  for the other parameter vary significantly more. Nonetheless, for  $N_{\text{fp}} > 400$  and  $450$  they vary by  $< 10$  and  $5\%$ , respectively. For  $\Omega_{\text{m}}, \Omega_{\text{b}}, h, n_s, \sigma_8$ , and  $M_{\text{lim}}$  we estimate  $\partial B_0/\partial\theta_i$  using a centered difference approximation (Eq. 11). However, for  $M_{\nu}$ , where we cannot have values below  $0.0 \text{ eV}$ , we cannot estimate the derivative with the same method. If we use the forward difference approximation,

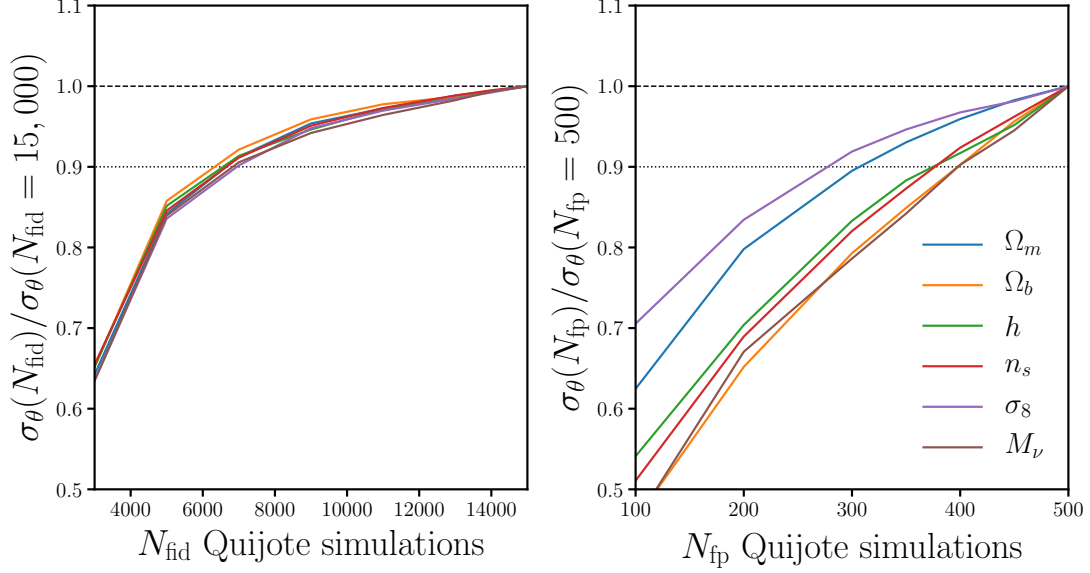
$$\frac{\partial \bar{B}_0}{\partial M_{\nu}} \approx \frac{\bar{B}_0(M_{\nu}^{\text{fid}} + \delta M_{\nu}) - \bar{B}_0(M_{\nu}^{\text{fid}})}{\delta M_{\nu}}, \quad (\text{B1})$$

the error goes as  $\mathcal{O}(\delta M_{\nu})$ . Instead, we use Eq.12 which provides a  $\mathcal{O}(\delta M_{\nu}^2)$  order approximation. In our  $\partial B_0/\partial M_{\nu}$  approximation, we use the Quijote simulations at  $M_{\nu}^+$ ,  $M_{\nu}^{++}$ , and  $M_{\nu}^{+++}$ . However, if we exclude the simulations at  $M_{\nu}^{+++}$ , the  $\sigma_{M_{\nu}} = 0.112$  and  $0.0238 \text{ eV}$  constraints for  $k_{\text{max}} = 0.2$  and  $0.5 \ h/\text{Mpc}$  from Section 4.2 increase to  $\sigma_{M_{\nu}} = 0.125$  and  $0.0288 \text{ eV}$ . Constraints from the power spectrum also increase so the relative improvement from the bispectrum remains roughly the same. We compare  $\partial \log B(k_1, k_2, k_3)/\partial M_{\nu}$  (right) and  $\partial \log P(k)/\partial M_{\nu}$  (left), computed using Eq.12 (blue), excluding  $M_{\nu}^{+++}$  (orange), and the forward difference approximation (green) in Figure 12.

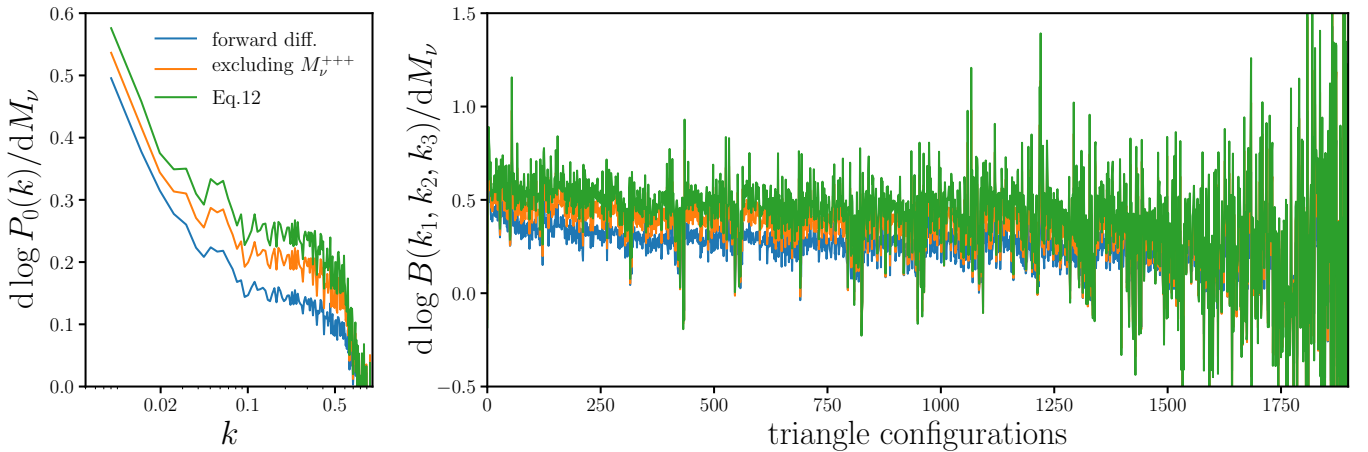
Hence we conclude that the derivatives have sufficiently converged and robust for our forecasts.

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**Figure 11. Left:** The ratio of the  $1\sigma$  Fisher constraint for  $\theta = \Omega_m, \Omega_b, h, n_s, \sigma_8$ , and  $M_\nu$  calculated using  $N_{\text{fid}}$  Quijote simulations over the constraint calculated with all 15,000 simulations,  $\sigma_\theta(N_{\text{fid}})/\sigma_\theta(N_{\text{fid}} = 15,000)$ , as a function of  $N_{\text{fid}}$ . The  $N_{\text{fid}}$  simulations are used to estimate  $\mathbf{C}$  used to calculate the Fisher matrix (Eq. 10). The Fisher parameter constraints vary by  $< 10$  and  $1\%$  for  $N_{\text{fid}} > 7000$  and  $14,000$ , respectively. **Right:** The ratio of the  $1\sigma$  Fisher constraint for  $\theta$  calculated using  $N_{\text{fp}}$  simulations over the constraint calculated with all 500 fixed paired simulations,  $\sigma_\theta(N_{\text{fp}})/\sigma_\theta(N_{\text{fp}} = 500)$ , as a function of  $N_{\text{fp}}$ . The  $N_{\text{fp}}$  fixed paired simulations are used to numerically estimate  $\partial B_0/\partial\theta_i$  in Eq. 10. Although  $\sigma_\theta(N_{\text{fp}})/\sigma_\theta(N_{\text{fp}} = 500)$  vary among the parameters, for  $N_{\text{fp}} > 400$  and  $450$  they vary by  $< 10$  and  $5\%$ , respectively. Hence, *we have a sufficient number of simulations to estimate  $\mathbf{C}$  and the derivatives of the bipsectrum and our forecasts are robust to their convergence.*



**Figure 12.** Comparison of  $\partial \log B(k_1, k_2, k_3)/\partial M_\nu$  (right) and  $\partial \log P(k)/\partial M_\nu$  (left), computed using Eq.12 (blue), excluding  $M_\nu^{+++}$  (orange), and the forward difference approximation (green).

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