The Atacama Cosmology Telescope: Cosmology from cross-correlations of unWISE galaxies and ACT DR6 CMB lensing

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ABSTRACT

We present tomographic measurements of structure growth using cross-correlations of Atacama Cosmology Telescope (ACT) DR6 and *Planck* CMB lensing maps with the unWISE Blue and Green galaxy samples, which span the redshift ranges $0.2 \lesssim z \lesssim 1.1$ and $0.3 \lesssim z \lesssim 1.8$, respectively. We improve on prior unWISE cross-correlations not just by making use of the new, high-precision ACT DR6 lensing maps, but also by including additional spectroscopic data for redshift calibration and by analysing our measurements with a more flexible theoretical model. An extensive suite of systematic and null tests within a blind analysis framework ensures that our results are robust. We determine the amplitude of matter fluctuations at low redshifts ($z \simeq 0.2 - 1.6$), finding $S_8 \equiv \sigma_8 (\Omega_m/0.3)^{0.5} = 0.813 \pm 0.021$ using the ACT cross-correlation alone and $S_8 = 0.810 \pm 0.015$ with a combination of *Planck* and ACT cross-correlations; these measurements are fully consistent with the predictions from primary CMB measurements assuming standard structure growth. The addition of Baryon Acoustic Oscillation data breaks the degeneracy between σ_8 and Ω_m , allowing us to measure $\sigma_8 = 0.813 \pm 0.020$ from the crosscorrelation of unWISE with ACT and $\sigma_8 = 0.813 \pm 0.015$ from the combination of cross-correlations with ACT and *Planck*. These results also agree with the expectations from primary CMB extrapolations in Λ CDM cosmology; the consistency of σ_8 derived from our two redshift samples at $z \sim 0.6$ and 1.1 provides a further check of our cosmological model. Our results suggest that structure formation on linear scales is well described by Λ CDM even down to low redshifts $z \leq 1$.

Contents

1.	Introduction	3
2.	Summary of key results	4
3.	 The data 3.1. unWISE galaxies 3.1.1. Galaxy Selection 3.1.2. Galaxy Redshift Distributions 3.1.3. Removing unWISE correlations with stellar density and WISE depth 3.2. CMB Lensing from the Atacama 	5 5 5 6
	Cosmology Telescope	8
	3.3. CMB Lensing from <i>Planck</i>	9
4. 5	CMB Lensing Tomography Simulations for power spectrum recovery and	9
0.	covariance estimation	10
6.	Covariance Matrices 6.1. Marginalisation over uncertainty in the	11
	lensing normalisation	13
7.	Tests for Systematic Errors 7.1. Testing for Contamination of the Lensing	13
	Reconstruction	13
	 7.2. Testing for Contamination and Homogeneity of the Galaxy Samples 7.3. Simulation Based Test for Extragalactic 	15
	Foregrounds	15
8.	The model 8.1. Hybrid Power Spectrum Model	17 17

	8.1.1. 3D Power Spectra and Bias	
	Expansion	18
	8.1.2. Power Spectra of Projected Fields	19
	8.1.3. Redshift Distribution and Bias	
	Evolution	20
	8.1.4. Redshift Evolution of Higher Order	
	Biases and Marginalisation	21
	8.1.5. Relative Contributions of Model	
	Components	22
8.2.	Marginalisation over Redshift	
	Uncertainties	22
8.3.	Testing the Model on Mocks	22

9.	Cosn	nological Analysis	25
	9.1.	Blinding Policy	25
	9.2.	Post-unblinding Changes	26
	9.3.	Parameter Inference and Priors	26
	9.4.	BAO Likelihoods	27
	9.5.	Constraints on Cosmological Parameters	
		from the cross-correlation of ACT DR6	
		Lensing and unWISE	28
	9.6.	Combination with the cross-correlation of	
		<i>Planck</i> Lensing and unWISE	30
	9.7.	Parameter Based Consistency Test	30
10	D.		0.1
10.	Discu	ission and Conclusions	31
	10.1.	Comparison with predictions from the	
		primary CMB	33
	10.2.	Comparison with other measurements of	
		the large scale structure of the universe	33
	10.3.	Our results in the context of the S_8/σ_8	
		"tension"	36
	10.4.	Conclusions	37

Α.	Off-diagonal correlations in the covariance matrix	43
	maura	10
В.	Detailed summary of null-tests	43
С.	Details of power spectrum modelC.1. Correction for the Fiducial Cosmology Assumed in Determining Cross-correlation RedshiftsC.2. Noise Bias Correction	44 44 50
D.	Posterior distributions for the full parameter space	50
E.	Parameter consistency from Blue and Green	50
F.	Summary of all parameter consistency tests	53
G.	Alternative priors	53
H.	Consistency test of galaxy-galaxy cross-spectrum	55
I.	 Reanalysis of the cross-correlation of <i>Planck</i> lensing and unWISE I.1. Summary of the Changes to the Analysis I.2. Lensing Monte Carlo Norm Correction I.3. Updated cosmological constraints from the cross-correlation of <i>Planck</i> lensing and unWISE 	57 57 58 59
J.	Impact of galaxy systematics weighting	60
К.	Spectroscopic samples used in cross-correlation redshifts	61

L. Detailed discussion of post-unblinding changes 70

1. INTRODUCTION

Measuring the amplitude of low redshift matter fluctuations can probe the growth of cosmic structure over time, reveal the properties of dark matter and dark energy, constrain the masses of neutrinos, and provide important tests of general relativity. A number of lensingrelated techniques have been developed for this purpose, including both galaxy lensing and CMB lensing.

Recent measurements of galaxy weak lensing from the Dark Energy Survey (DES; Flaugher et al. 2015; Abbott et al. 2022), the Kilo-Degree Survey (KiDS; Kuijken et al. 2015; Heymans et al. 2021), Hyper Suprime-Cam (HSC; Aihara et al. 2017; More et al. 2023; Miyatake et al. 2023; Sugiyama et al. 2023), among others, have found a $2 - 3\sigma$ lower amplitude of fluctuations, compared to the prediction from *Planck* primary CMB assuming standard structure growth. In

particular, parameterising the low-redshift amplitude by $S_8 \equiv \sigma_8 (\Omega_m/0.3)^{0.5}$ (which is the quantity best measured by galaxy weak lensing surveys), DES, KiDS, and HSC obtain $S_8 = 0.782 \pm 0.019^{-1}$, $S_8 = 0.765^{+0.017}_{-0.016}^{-2}$, $S_8 = 0.775^{+0.043}_{-0.038}^{-3}$ respectively from a combination of cosmic shear and galaxy clustering (Abbott et al. 2022; Heymans et al. 2021; Sugiyama et al. 2023). HSC also presents an alternative analysis using smaller scales and a halo model based approach using a cosmological emulator (rather than the linear bias approximation used on large scales) finding $S_8 = 0.763^{+0.040}_{-0.036}$ ⁴ (Miyatake et al. 2023). A recent joint reanalysis of the DES and KiDS cosmic shear data hints at a slightly higher value of $S_8 = 0.790^{+0.018}_{-0.014}$ (Abbott et al. 2023a). Comparing to the value $S_8 = 0.834 \pm 0.016$ from the *Planck* primary CMB (or $S_8 = 0.832 \pm 0.013$ when including CMB lensing; Planck Collaboration et al. 2020a), we see that most recent galaxy weak lensing surveys have a preference for slightly lower S_8 , even though no single experiment is in significant tension with Planck. The recent DES + KiDS reanalysis of cosmic shear is consistent at the 1.7σ level.

CMB lensing probes the same physics as galaxy weak lensing, but has different potential systematics; a known redshift of the "source" (the primary CMB) and wellunderstood statistical properties make it a particularly reliable and independent cosmological probe. The CMB lensing auto-correlation is in excellent agreement with the primary CMB predictions: recent measurements from the Atacama Cosmology Telescope (ACT Qu et al. 2023; Madhavacheril et al. 2023) combined with Baryon Acoustic Oscillation (BAO) data yield $S_8 = 0.840 \pm$ 0.028 from ACT alone and $S_8 = 0.831\pm0.023$ when combined with *Planck* lensing (Planck Collaboration et al. 2020b; Carron et al. 2022).

However, the cross-correlation between CMB lensing and galaxies has at times shown a preference for a lower S_8 . For example, cross-correlations of DESI Luminous Red Galaxy (LRG) targets (White et al. 2022; Kitanidis

¹ This value is obtained from the public DES parameter chain with fixed neutrino mass here. We adopt this result throughout for better comparability with our analysis which fixes the neutrino mass to the minimum value allowed in the normal hierarchy.

 $^{^2}$ This differs from the results reported by KiDS, $S_8 = 0.766^{+0.020}_{-0.014}$. KiDS by default reports the maximum a posteriori (MAP) value along with the projected joint highest posterior density region (PJ-HPD). For better comparability, however, we adopt throughout the marginalised mean and credible interval instead.

 $^{^3}$ We note that the HSC results we compare to in Sec. 10.2 differ slightly from these values as they are derived from a reanalysis of the HSC data with prior choices consistent with those adopted in this work.

⁴ See footnote 3.

& White 2020; Hang et al. 2021) or galaxies from the Baryon Oscillation Spectroscopic Survey (BOSS) (Singh et al. 2020; Chen et al. 2022b) with Planck lensing show a preference for a lower S_8 at up to 3σ . An analysis of the cross-correlation between CMB lensing from the South Pole Telescope (SPT) and *Planck* and various data sets from DES finds varying levels of tension between 2.2 and 3σ (see Sec. 10.2 for details; Chang et al. 2023; Abbott et al. 2023b). Cross-correlations with previous CMB lensing reconstructions from ACT have also found this preference at the $\sim 2\sigma$ level, albeit with substantial uncertainties. Using CMB lensing data from ACT DR4 and *Planck* together with galaxy shear from KiDS-1000, Robertson et al. (2021) found $S_8 = 0.64 \pm 0.08$; similarly an analysis of the crosscorrelation between ACT DR4 CMB lensing and galaxy clustering from DES-Y3 yields $S_8 = 0.75^{+0.04}_{-0.05}$ (Marques et al. 2023).

Moreover, Krolewski et al. (2021) used the unWISE full-sky galaxy catalogue split into three tomographic bins spanning the redshift range $0 \leq z \leq 2$ (with the power spectra and redshift distributions measured in Krolewski et al. 2019), together with the *Planck* 2018 CMB lensing maps (Planck Collaboration et al. 2020b), to set a constraint $S_8 = 0.784 \pm 0.015$, about 2.4 σ lower than estimated from the primary CMB. Motivated by the availability of additional data for the unWISE redshift distribution calibration, improved understanding of the calibration of *Planck* CMB lensing maps and the availability of lower-noise CMB lensing maps from Planck PR4 and ACT DR6, we perform a new and improved analysis of the cross-correlation between un-WISE galaxies and CMB lensing. An expanded set of null-tests, enhanced foreground control and improved theoretical modelling add to the robustness of the measurement.

2. SUMMARY OF KEY RESULTS

The key results of this work are constraints on structure formation at $z \simeq 0.2$ –1.6. Within a Λ CDM cosmology the best constrained parameter combination in the analysis presented here is approximately $\sigma_8 \Omega_m^{0.45}$. This differs slightly from the best constrained parameter combination in galaxy weak lensing analyses, $\sigma_8 \Omega_m^{0.5}$. To facilitate comparisons with other probes we mainly present constraints on the commonly used parameter $S_8 \equiv \sigma_8 (\Omega_m/0.3)^{0.5}$ (Abbott et al. 2022; Heymans et al. 2021; Li et al. 2023b; Dalal et al. 2023) in this paper; however, we also define a cross-correlation equivalent $S_8^{\times} \equiv \sigma_8 (\Omega_m/0.3)^{0.45}$. CMB lensing auto-spectrum analyses, in contrast, primarily probe the combination $S_8^{\text{CMBL}} \equiv \sigma_8 (\Omega_m/0.3)^{0.25}$ (Qu et al. 2023; Madhavacheril et al. 2023; Carron et al. 2022; Planck Collaboration et al. 2020b). In this paper, we also combine with BAO measurements that probe Ω_m independently of σ_8 to obtain constraints on σ_8 that are directly comparable to other probes.

When using only the cross-correlation between the ACT DR6 CMB lensing reconstruction and the two unWISE galaxy samples, Blue and Green (previously defined in Krolewski et al. 2019), as well as the autocorrelation of the galaxy samples measured on the overlapping area, we find $S_8 = 0.813 \pm 0.021$ ($S_8^{\times} = 0.817 \pm 0.019$). We combine these data with the crosscorrelation using a lensing reconstruction from *Planck* PR4 (Carron et al. 2022) at the likelihood level and obtain improved constraints of $S_8 = 0.810 \pm 0.015$ ($S_8^{\times} = 0.814 \pm 0.014$). The posterior distributions for S_8 are shown in the left panel of Fig. 1. With the addition of BAO we find $\sigma_8 = 0.813 \pm 0.020$ from unWISE and ACT alone and $\sigma_8 = 0.813 \pm 0.015$ when combining with *Planck* lensing (see the right panel of Fig. 1).

The model best fitting the measured C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ is shown in Fig. 3. We find a minimum $\chi^2 = 20.9$. Our model, described in detail in Sec. 8, includes 20 model parameters. Of those 15 are largely prior dominated yielding an approximate number of degrees of freedom of 21 (there are a total of 26 bandpowers for the Blue and Green samples combined). Therefore, we estimate the probability to exceed (PTE) for the model at 0.5.

These conclusions differ from previous results from the cross-correlation of unWISE with CMB lensing reconstruction from Planck PR3 (Planck Collaboration et al. 2020b) presented in Krolewski et al. (2021). In Appendix I we present a reanalysis of the *Planck* cross-correlation, finding results consistent with those presented here. We discuss the corrections and improvements that lead to the discrepancy. Other crosscorrelation measurements still exhibit a preference for lower S_8 (Singh et al. 2020; Kitanidis & White 2020; White et al. 2022; Chen et al. 2022b; Chang et al. 2023; Abbott et al. 2023b; Robertson et al. 2021; Margues et al. 2023) and while some of the improvements made here (e.g., the inclusion of a Monte Carlo correction for the lensing normalisation) are also relevant to those works, preliminary results show that they are likely insufficient to alleviate the discrepancy completely.

Our results are in good agreement with the predictions for the amplitude of low redshift structure obtained from the primary CMB observed by the *Planck* satellite (within 1.1 σ for S_8 and 0.1 σ for σ_8 ; Planck Collaboration et al. 2020a) as well as measurements from CMB lensing using either ACT (0.3 σ for σ_8 ; Qu et al. 2023; Madhavacheril et al. 2023) or *Planck* (0.04 σ for σ_8 ; Carron et al. 2022; Planck Collaboration et al. 2020b). On the other hand, our results are also not in significant tension with recent results from any one galaxy weak lensing survey such as DES, KiDS, HSC (1.1σ , 2.0σ , 0.8σ respectively for S_8 ; Abbott et al. 2022; Heymans et al. 2021; Sugiyama et al. 2023), although our results favour a higher amplitude of low redshift structure. Overall, our results do not strengthen the case for a real discrepancy in structure growth or for problems with the Λ CDM model, at least in the low-redshift, linear-scale regime to which we are sensitive.

3. THE DATA

We use the unWISE galaxy catalogue (Krolewski et al. 2019; Schlafly et al. 2019) and lensing reconstructions from the anticipated Data Release 6 (DR6) of ACT (Qu et al. 2023; Madhavacheril et al. 2023; MacCrann et al. 2023). We also combine our cross-correlation analysis (results given in Sec. 9.5) with the cross-correlation obtained using the latest CMB lensing reconstruction based on data from the *Planck* satellite (Carron et al. 2022).

In Sec. 3.1 we briefly describe the unWISE samples, their selection, the measurement of their redshift distribution, and the mitigation of observational systematics. Subsequently, we describe the ACT DR6 lensing reconstruction in Sec. 3.2 and the *Planck* PR4 lensing reconstruction in Sec. 3.3.

3.1. unWISE galaxies

The unWISE galaxy catalogue is constructed from the Wide-Field Infrared Survey Explorer (WISE) survey (Wright et al. 2010), including four years of the post-hibernation NEOWISE phase (Mainzer et al. 2011, 2014). The WISE satellite mapped the entire sky at 3.4 (W1), 4.6 (W2), 12 (W3) and 22 (W4) μ m, although NEOWISE only measured in bands W1 and W2 due to a lack of cryogen necessary for the longer-wavelength bands. As a result, unWISE (Lang 2014; Meisner et al. 2017) is constructed only from the much deeper W1 and W2 bands.

3.1.1. Galaxy Selection

We select three galaxy samples from unWISE using W1-W2 colour cuts, called the Blue, Green, and Red samples, at $z \sim 0.6$, 1.1, and 1.5, respectively. These samples are extensively described in Krolewski et al. (2019) and Schlafly et al. (2019). The Red sample of galaxies has a significantly lower number density than the other two samples and was shown to contribute negligible constraining power (Krolewski et al. 2021). Krolewski et al. (2021) also showed that it was difficult

to reproduce the red sample in *N*-body simulations and to recover unbiased cosmology from simulations tuned to reproduce the properties of the red sample. Thus we do not use the Red sample.

In addition to the unWISE colour cuts, we additionally remove any unWISE source within 2.75'' of a Gaia (Gaia Collaboration et al. 2018) point source, and remove potentially spurious sources in unWISE imaging. as described in Krolewski et al. (2019). Residual stellar contamination is < 2%. As pointed out in Krolewski et al. (2019), the power spectrum of stars drops very rapidly with ℓ , such that even at $\ell \sim 50$, the stellar power spectrum multiplied by the contamination fraction is < 0.5% of the galaxy auto power spectrum, compared to the 2% statistical errors of the galaxy autospectrum at that bandpower. In the limit that the stellar power spectrum contributes negligible power, stellar contamination simply modifies the number density in a way that is completely degenerate with the linear galaxy bias.

The unWISE mask is based on the *Planck* 2018 CMB lensing mask (Planck Collaboration et al. 2020b) with a Gaussian apodisation of 1°, and additional cuts around bright stars, nearby galaxies, planetary nebulae, and optical diffraction spikes. Because we remove any source within 2.75" of a Gaia point source and mask out diffraction spikes and other narrow regions around stars, the effective area within each HEALPix pixel is less than the full area of the pixel. We correct the counts in each pixel for the effective area (and mask out any pixel with < 80% coverage), before applying the unWISE mask. The large-scale component of the unWISE mask, derived from the *Planck* lensing mask, is apodised by smoothing with a 1° FWHM Gaussian; the other components of the mask are treated as binary. The effective sky fraction after masking is $f_{\rm sky} = 0.586$.

3.1.2. Galaxy Redshift Distributions

Since the unWISE galaxies are selected from two-band photometry, it is not possible to determine photometric redshifts for individual galaxies. Instead, we use crosscorrelations with Sloan Digitial Sky Survey (SDSS) spectroscopy to measure the redshift distributions of the samples (e.g., Newman 2008; McQuinn & White 2013; Ménard et al. 2013). The unWISE cross-correlation redshifts were originally presented in Krolewski et al. (2019) using spectroscopic samples from LOWZ and CMASS galaxies (Reid et al. 2016), BOSS quasars (Ross et al. 2020), and eBOSS (extended BOSS) DR14 quasars (Ata et al. 2018). Krolewski et al. (2019) only considered objects in the Northern Galactic Cap (NGC) where most



Figure 1. The main result from this work is a constraint on the amplitude of low redshift structure captured by the parameter $S_8 \equiv \sigma_8 (\Omega_m/0.3)^{0.5}$ (left). We obtain $S_8 = 0.810 \pm 0.015$ using the combination of cross-correlation measurements of the unWISE galaxies with ACT DR6 and *Planck* PR4 lensing reconstructions. Combining with baryon acoustic oscillations (BAO), which constrain Ω_m , we obtain constraints on σ_8 of $\sigma_8 = 0.813 \pm 0.015$ (right). Our results show no significant tension with values inferred from the primary CMB from *Planck*.

of the spectroscopy lies. These tracers span the full redshift range of the unWISE samples.

We update the unWISE cross-correlation redshifts first measured in Krolewski et al. (2019) to include additional data. We use the Southern Galactic Cap (SGC) footprints for all tracers, due to the better overlap with the mostly-southern ACT footprint. We replace the eBOSS DR14 quasars with the final eBOSS DR16 quasars (Ross et al. 2020), leading to a significant increase in the area (1178 to 4752 deg^2) and number of quasars used (54708 to 343708). Finally, we also use the eBOSS DR16 LRGs (Ross et al. 2020). We do not use the eBOSS DR16 emission line galaxies (ELGs), due to the significantly smaller area compared to the LRGs or quasars $(1120 \text{ deg}^2 \text{ vs. } 4202 \text{ deg}^2 \text{ for LRG and } 4808 \text{ deg}^2$ for quasars) and potential systematics in the ELG autocorrelation needed to measure the ELGs' spectroscopic bias. These additions significantly improve the crosscorrelation redshifts at $z \sim 0.7-2$, the redshift range where most of the unWISE galaxies lie. In Appendix K we describe the spectroscopic samples used and their relevant properties (linear and magnification biases), then discuss the impact of the additional data used.

Conveniently, cross-correlation redshifts are sensitive to the product of the galaxy bias and the number density, b(z)dN/dz, that appears in the dominant terms in our model (Section 8.1). The conversion from the measured correlation function to b(z)dN/dz leads to a dependence on a fiducial cosmology, although we note that there is no dependence on the amplitude of the power spectrum because we require that b(z)dN/dz is normalised to integrate to unity. The residual cosmology dependence is quite minor, and the correction is described in detail in Appendix C.1 (updating the heuristic correction presented in Krolewski et al. 2021). Although the spectra we measure primarily depend on b(z)dN/dz, the lensing magnification and some higher-order galaxy bias terms also depend on dN/dz, which we measure by cross-matching unWISE galaxies to deep photometric redshifts from the COSMOS2015 catalogue (Laigle et al. 2016). This redshift distribution is consistent with the cross-correlation b(z)dN/dz assuming a simple halo occupation distribution (HOD), consistent with the un-WISE number density, to determine the bias evolution b(z). From the COSMOS cross-match, we find no evidence for unWISE galaxies to lie at z > 1.7 (2.5) for Blue (Green) and therefore set the smoothed redshift distribution to zero above these high redshift thresholds.

The cross-correlation estimates of b(z)dN/dz are shown in the top panel in Fig. 2. We normalise the cross-correlation redshift estimates (for more details see Sec. 8.1). We also show in the lower panel the crossmatch dN/dz from COSMOS2015.

3.1.3. Removing unWISE correlations with stellar density and WISE depth

In addition to cosmological fluctuations, the observed galaxy number densities are determined by foregrounds such as Galactic dust and stars, as well as survey depth and other imaging properties. Both foregrounds and imaging properties affect the galaxy selection. The power spectra of these contaminants are generally red; if uncorrected they therefore add significant power to the galaxy auto-correlation at low ℓ . The standard approach for galaxy surveys is to create a set of weights



Figure 2. The Blue and Green samples of unWISE galaxies span the redshift ranges $z \simeq 0.2 - 1.1$ and $z \simeq 0.3 - 1.8$ with mean redshifts of approximately 0.6 and 1.1 respectively. In the **top** panel we show the normalised estimate of b(z)dN/dz obtained from cross-correlating with spectroscopic tracers from SDSS, BOSS, and eBOSS. The Blue and Green curves are the spline interpolations of the best fitting estimates of b(z)dN/dz for the two galaxy samples. We also show several noise realisations in grey and the clustering redshift measurements in red. The **lower** panel shows estimates of dN/dz obtained by cross-matching with photometric data from COSMOS2015.

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that are designed to remove any relationship between the galaxy overdensity and various imaging property maps (Ross et al. 2012, 2017; Ata et al. 2018; Bautista et al. 2018; Elvin-Poole et al. 2018; Ross et al. 2020; Rodríguez-Monroy et al. 2021). Krolewski et al. (2021) did not follow this approach; instead, the authors applied a high-pass filter to the galaxy data, removing all modes of the galaxy survey at $\ell < 20$, and found that this led to better agreement between the $\ell < 100$ autocorrelation and a theory model. Additionally, Krolewski et al. (2021) did not use the galaxy auto-correlation at $\ell < 100$, where changing the Galactic mask significantly changed the shape of the unWISE auto-correlation⁵.

In this work, we update the method used in Krolewski et al. (2021) to apply weights that explicitly remove correlations between the galaxy density and maps of stellar density and WISE depth. This is similar to the approach taken by other galaxy surveys and ensures that our unWISE galaxy maps are uncorrelated with known foreground survey systematics that may affect the auto correlation at $\ell > 100$. These weights were originally created in Krolewski & Ferraro (2022) to use the low- ℓ unWISE data in cross-correlation with CMB temperature to measure the integrated Sachs-Wolfe effect. We also no longer filter out the low- ℓ ($\ell < 20$) modes in the unWISE map. The large-scale filtering has a similar effect to weighting, also reducing large-scale power by removing correlations between systematics and the true galaxy density. However, removing large scales in harmonic space complicates the use of the MASTER algorithm Hivon et al. (2002) to obtain unbiased bandpowers through mode decoupling. Hence, we no longer adopt this method. This change has only a small impact on the power spectra (< 0.5% on the auto- and crosscorrelation; see Fig. 45), but we consider the updated results more robust because the galaxy density has a significantly reduced dependence on both Galactic stellar density and WISE depth.

In the remainder of this section we provide more detail on the construction of these weights. We measure the correlation between unWISE galaxy density and several templates: 1) Gaia stellar density; 2&3) W1 and W2 limiting magnitude; 4) dust extinction E(B - V) from the Schlegel-Finkbeiner-Davis map corrected for cosmic infrared background (CIB) contamination (cSFD, Schlegel et al. 1998; Chiang 2023)⁶; 5) neutral hydrogen column density NHI from the H14PI survey (HI4PI Collaboration et al. 2016) as an alternative dust map that is noisier than SFD but has much reduced extragalactic contamination (Chiang & Ménard 2019); 6&7) a 3.5 and 4.9 μ m sky brightness from the DIRBE Zodi-Subtracted Mission Average (ZSMA)⁷; 8) a DIRBE measurement of

- ⁶ The relationship between unWISE galaxy density and cSFD is nearly identical if we use the uncorrected SFD instead.
- ⁷ https://lambda.gsfc.nasa.gov/product/cobe/dirbe_zsma_data_get.cfm

⁵ At $\ell \gtrsim 100$, changing the Galactic (or ecliptic latitude) mask produced a scale-independent change in the unWISE autocorrelation (Krolewski et al. 2019). This is due to the fact that the selection properties of the galaxy catalogue vary with Galactic or ecliptic latitude due to variations in the WISE coverage depth. This induces differences in the galaxy bias. We return to this in Sec. 7.2.

the total 4.9 μ m background light at solar elongation of 90°⁸ (WISE always observed at 90° solar elongation; Mainzer et al. 2011); and 9) a separate estimate of the zodiacal background light from the DIRBE Sky and Zodi Atlas (DSZA)⁹ (Kelsall et al. 1998).

We find that the strongest correlations are with Gaia stellar density and W2 limiting magnitude. To conservatively guard against the possibility of overfitting (Weaverdyck & Huterer 2021), we only regress against these two maps to define the systematics weights. Several other correlations are also quite significant (particularly with W1 limiting magnitude), but are significantly mitigated after weighting against stellar density and W2 depth, due to correlations between these contaminant templates. For example, both W1 and W2 depths are set by the WISE scan strategy, so the two maps exhibit the same structure.

For the Blue sample, we fit a linear trend between unWISE galaxy density and Gaia stellar density, and a three parameter piecewise linear trend to the W2 5σ limiting magnitude. We determine uncertainties on the measured correlations using the variance of density values from 100 isotropic Gaussian mocks (no correlation with the maps of imaging systematics). For the Green sample, we use a piecewise linear fit with three free parameters for both stellar density and W2 5 σ magnitude. We show the relationship between unWISE density and imaging systematics in Appendix J (Figs. 43 and 44), both before and after applying weights. Before applying weights, the correlations between the galaxy density and all nine templates listed above are highly significant; after weighting, induced variations in the galaxy number density are reduced to $\sim 2\%$. While the χ^2 for all correlations is significantly reduced, it is still indicative of statistically significant correlations in some cases. However, the covariance estimate is approximate (for instance, the errors are assumed to be diagonal) and potentially is an underestimate. Even the dramatic reduction in χ^2 after applying weights to correct the correlations with stellar density and W2 limiting magnitude leads to only a minor impact on the auto-spectrum (< 3% on scales $\ell > 100$) and does not impact the cross-correlation spectrum significantly. We conclude that any residual correction will be negligible given the size of our statistical

errors. The impact of applying these weights on our parameter inference is $\Delta S_8 = 0.006 \ (\sim 0.3\sigma)^{10}$.

3.2. CMB Lensing from the Atacama Cosmology Telescope

We use the CMB lensing convergence map reconstructed from the CMB temperature and polarisation anisotropy data from the upcoming DR6 of ACT (Qu et al. 2023; Madhavacheril et al. 2023; MacCrann et al. 2023). This release is based on CMB measurements made between 2017 and 2021 (relying only on the nighttime data) at ~90 and ~150 GHz. It uses an early version of the ACT DR6 maps, labelled dr6.01.

The lensing maps cover 9400 deg^2 of the sky and are signal dominated on scales of $\ell < 150$. These maps are reconstructed using a cross correlation based estimator (Madhavacheril et al. 2020). Rather than using a single CMB map, the cross correlation estimator uses several time-interleaved splits ensuring that the instrumental noise of each map is independent. The resulting CMB lensing map is therefore insensitive to modelling of the instrumental noise.

Lensing is reconstructed with CMB scales from $600 < \ell < 3000$. The large scales of the input CMB map, $\ell < 600$, are excluded due to significant atmospheric noise, Galactic foregrounds and a > 10% correction for the large scale transfer functions. The small scales are excluded to minimise contamination from astrophysical foregrounds like the thermal Sunyaev-Zeldovich (tSZ) effect, the cosmic infrared background (CIB), and radio sources MacCrann et al. (2023). To further mitigate extragalactic foregrounds, the lensing reconstruction uses a profile-hardened lensing estimator (Sailer et al. 2020). This involves constructing a quadratic estimator that is immune to the contribution to the CMB mode coupling arising from objects with radial profiles similar to those expected from tSZ clusters.

The baseline ACT DR6 lensing mask is constructed from a Galactic mask which selects the 60% of the sky with the lowest dust contamination (see Qu et al. 2023, for a detailed discussion of the ACT lensing mask). As a consistency test we also use a ACT lensing reconstruction using only the area within the 40% of the sky with the lowest dust contamination (see Sec. 7). We will subsequently refer to these masks as the 60% and 40% Galactic masks respectively.

 $^{^{8}}$ https://lambda.gsfc.nasa.gov/product/cobe/c_90deg_skymap. html

 $^{^{9}}$ https://lambda.gsfc.nasa.gov/product/cobe/dirbe_dsza_data_get.cfm

¹⁰ This includes the small change in the measured clustering redshifts due to the application of the weights (see Sec. K). The mean redshift of the Blue and Green samples changes by $\Delta \bar{z} = 0.007$ and $\Delta \bar{z} = -0.002$ respectively.

3.3. CMB Lensing from Planck

The *Planck* PR4 lensing analysis (Carron et al. 2022) reconstructs lensing with CMB angular scales from $100 \leq \ell \leq 2048$ using the quadratic estimator. This analysis is based on the reprocessed PR4 NPIPE maps that incorporated around 8% more data compared to the 2018 *Planck* PR3 release. It also includes pipeline improvements such as optimal (anisotropic) filtering of the input CMB fields resulting in an increase of the overall signal-to-noise ratio by around 20% compared to *Planck* PR3 (Planck Collaboration et al. 2020b) and a detection of the lensing power spectrum at 42σ .

Qu et al. (2023) demonstrated good consistency between the ACT and *Planck* PR4 lensing bandpowers justifying combining the measurements at the likelihood level to obtain tighter constraints. Similar to the analysis in Qu et al. (2023) and Madhavacheril et al. (2023), here we combine the cross correlation of ACT DR6 CMB lensing and unWISE with the cross correlation measurement between *Planck* PR4 CMB lensing and unWISE at the likelihood level. We do this while accounting for the overlap in area and scales used in the lensing reconstruction using a simulated ACT-*Planck* covariance matrix (see Sec. 6 for details).

4. CMB LENSING TOMOGRAPHY

We measure the auto-spectra of the two galaxy samples described in Sec. 3.1.1 and their cross-correlation spectra with the lensing map from ACT described in 3.2 on the cut sky using a pseudo- C_{ℓ} estimator (Hivon et al. 2002).

Given two fields in the sky $a(\boldsymbol{\theta})$ and $b(\boldsymbol{\theta})$ a simple estimator of the pseudo- C_{ℓ} is

$$\tilde{C}_{\ell}^{ab} = \frac{1}{2\ell + 1} \sum_{m = -\ell}^{\ell} a_{\ell m} (b_{\ell m})^{\dagger}, \qquad (1)$$

where $a_{\ell m}$ and $b_{\ell m}$ are the spherical harmonic transforms (SHT) of the fields a and b. These pseudo- C_{ℓ} differ from the true C_{ℓ} due to mask induced mode-coupling. Their expectation value $\left\langle \tilde{C}_{\ell} \right\rangle$ can be related to the true C_{ℓ} as

$$\left\langle \tilde{C}_{\ell} \right\rangle = \sum_{\ell'} C_{\ell'} M_{\ell\ell'} \tag{2}$$

where $M_{\ell\ell'}$ is a mode coupling matrix which can be computed from the power spectrum of the mask alone. This relation can be approximately inverted if the power spectrum is assumed to be piecewise constant across a number of discrete bins (Hivon et al. 2002; Alonso et al. 2019). To perform the mode-decoupling of the binned

 C_{ℓ} we use the implementation in the NaMaster code¹¹ (Alonso et al. 2019). It should be noted that lensing reconstruction with a quadratic estimator, when performed on the masked sky, effectively convolves the mask with the signal in a non-trivial manner that is not captured exactly by the NaMaster algorithm. We approximate this added complexity by taking the mask used to compute the mode-coupling matrix to be the square of the ACT mask¹². Our tests on simulations (see Sec. 5) suggest that this approximation performs to within better than 1% on the scales of interest. Small residuals are corrected using a transfer function as described in section Sec. 5. We use HEALPix maps with nside = 2048and run NaMaster with $\ell_{\max}^{\text{NaMaster}} = 3000$ even though we use only multipoles $\ell \leq 400$ in our analysis to avoid bias from the pseudo- C_{ℓ} method.

We measure the cross-correlation between the un-WISE galaxies and the ACT DR6 lensing reconstruction (see Fig. 3) with a signal-to-noise ratio of 40 and 51 within our cosmological analysis range of $50 \le \ell \le 400$ for the two galaxy samples, Blue and Green, respectively. This range was set prior to unblinding. If we were able to reliably model smaller scales we would be able to leverage even larger signal-to-noise, increasing to 67 and 74 respectively for $20 \le \ell \le 3000$.

The minimum multipole ℓ_{\min} is chosen based on our systematics tests presented in Sec. 7 to guard against contamination from large scale observational systematics that may be correlated between the galaxy sample and the lensing reconstruction (as well as against potential mis-estimation of the lensing reconstruction mean field, which could lead to an underestimate of errors). The maximum multipole scale cut, ℓ_{max} , is predominantly set by the requirement for unbiased recovery of cosmological parameters, which becomes challenging with our chosen model on smaller scales (see Sec. 8.3). For the galaxy auto-correlation, C_{ℓ}^{gg} , we adopt a more conservative minimum scale cut, $\ell_{\min} = 100$, because the auto-correlation is more susceptible to observational systematics than the cross-correlation. On larger scales we observe significant fluctuations of the auto-correlation signal, for example for different choices of the mask (see Sec. 7.2 and also Sec. 7.3 in Krolewski et al. 2019).

¹¹ https://github.com/LSSTDESC/NaMaster

¹² This is a good approximation (at least when the mask varies on much larger scales than the CMB and lensing scales of interest) since the lensing signal is reconstructed using a quadratic estimator which reconstructs the lensing from the off-diagonal correlations in the temperature and polarisation maps which each carry one power of the mask.



Figure 3. We measure C_{ℓ}^{gg} (top row) and $C_{\ell}^{\kappa g}$ (middle row) for the Blue and Green samples of unWISE galaxies. The total signal to noise in $C_{\ell}^{\kappa g}$ on all measured scales ($20 \leq \ell \leq 3000$) is approximately 67 and 74 respectively. Within the analysis range ($50 \leq \ell \leq 400$) we obtain SNR of 40 and 51 for Blue and Green respectively. The grey line shows the best fit from the joint fit to both samples, with the model residuals shown in the **bottom** row. The total model χ^2 for the joint fit is 21, slightly larger than the sum of the χ^2 for each of the two samples (9.7 and 10.5 for Blue and Green respectively) due to the non-zero off-diagonal covariance between them. We estimate the model PTE at 0.5.

5. SIMULATIONS FOR POWER SPECTRUM RECOVERY AND COVARIANCE ESTIMATION

To test recovery of unbiased power and to compute covariances we use 400 CMB lensing reconstruction simulations and appropriately correlated Gaussian simulations of the galaxy number density fields. The lensing simulations use a Gaussian lensing convergence to lens a randomly drawn CMB realisation. We add realistic survey noise and masking (for details on the noise simulations see Atkins et al. 2023), and reconstruct the lensing convergence with the pipeline used on the data (Qu et al. 2023).

To generate the galaxy simulations we measure \hat{C}_{ℓ}^{gg} and $\hat{C}_{\ell}^{\kappa g}$ on the two unWISE samples in narrow ℓ -bins. We fit a model with fixed cosmology and free nuisance

parameters¹³. We use the range $20 \leq \ell \leq 3000$ for $C_{\ell}^{\kappa g}$ and $100 \leq \ell \leq 3000$ for C_{ℓ}^{gg} . We do not fit \hat{C}_{ℓ}^{gg} and $\hat{C}_{\ell}^{\kappa g}$ jointly but rather individually. When fitting \hat{C}_{ℓ}^{gg} and $\hat{C}_{\ell}^{\kappa g}$ individually the amplitude of the spectrum is completely degenerate with the galaxy bias. We do not assess the consistency of the galaxy bias found for the auto and cross-correlation in order not to compromise our blinding procedure. Using the model fit we compute fiducial $C_{\ell, \text{fid}}^{gg}$ and $C_{\ell, \text{fid}}^{\kappa g}$ and convolve them with the appropriate pixel window-function for our chosen map resolution of nside = 2048. To capture the excess noise present on large scales in \hat{C}_{ℓ}^{gg} , we additionally fit a

¹³ The model and its parameters are described in Sec. 8. For this fit we vary all nuisance parameters, including the galaxy bias, the shot-noise, the higher order bias parameters and the principal component expansion coefficients for the cross-correlation redshifts. In total there are 8 parameters for the Blue sample and 10 parameters for Green

smoothly broken power-law with an exponential cut-off at $\ell_{\rm noise,max} \leq 100$ to $\hat{C}_{\ell}^{gg} - C_{\ell,\rm fid}^{gg}$ on large scales. Using the Gaussian realisations of the lensing convergence field employed in generating the lensing simulations, we compute Gaussian realisations of the galaxy field that match the observed cross-correlation between CMB lensing and galaxy number density as

$$a_{\ell m}^{g} = \frac{C_{\ell, \text{fid}}^{\kappa g}}{C_{\ell, \text{fid}}^{\kappa \kappa}} a_{\ell m}^{\kappa} + a_{\ell m}^{g, \text{uncorrelated}} + a_{\ell m}^{g, \text{noise}}.$$
 (3)

The contributions $a_{\ell m}^{g, {\rm uncorrelated}}$ and $a_{\ell m}^{g, {\rm noise}}$ are drawn such that

$$\langle a_{\ell m}^{g, \text{uncorr.}} (a_{\ell' m'}^{g, \text{uncorr.}})^* \rangle = \delta_{\ell \ell'} \delta_{m m'} \left(C_{\ell, \text{fid}}^{gg} - \frac{(C_{\ell, \text{fid}}^{\kappa g})^2}{C_{\ell, \text{fid}}^{\kappa \kappa}} \right)$$

$$\tag{4}$$

$$\left\langle a_{\ell m}^{g,\text{noise}}(a_{\ell'm'}^{g,\text{noise}})^* \right\rangle = \delta_{\ell\ell'}\delta_{mm'}C_{\ell,\text{noise}}^{gg}.$$
 (5)

This yields the correct auto- and cross-spectra. We do not include correlations between the different galaxy samples (for the purposes of the covariance computation in Sec. 6, we include an analytic approximation for the covariance between different samples). We measure the auto-correlations of these simulations as well as their correlation with lensing reconstruction simulations in exactly the same way as we treat the data. We then compare the measured and decoupled power spectra to the input power spectra. Note that, since the mode decoupling is only approximate, in order to make a fair comparison we have to first convolve the input $C_{\ell,fid}$ with the appropriate mode-coupling matrix (see Eq. 2) before binning them and applying the approximate decoupling applied to the measured \hat{C}_{ℓ} . We find that the recovered \hat{C}_{ℓ}^{gg} and $\hat{C}_{\ell}^{\kappa g}$ are consistent with the inputs to within $\leq 1\%$ in our cosmology range (see Fig. 4). Our simulations mimic the large scale excess of power present in the galaxy data and we use them to compute a multiplicative transfer function correcting for this effect and small inaccuracies in the mode decoupling. This excess (< 0.5% for Blue and < 1% for Green), which is present only in the lowest- ℓ bandpower in C_{ℓ}^{gg} , is much smaller than the uncertainty on our data bandpowers ($\sim 3\%$ in the lowest- ℓ bin), but is detected significantly with 400 simulations.

6. COVARIANCE MATRICES

We use the suite of 400 Gaussian simulations of the galaxy and lensing fields discussed in Sec. 5 to estimate the covariance $\text{Cov}(C_{\ell}^{XY}, C_{\ell'}^{AB})$ $(XY, AB \in \{gg, \kappa g\})$. The diagonal elements of the covariance matrix, as well as the diagonals of the $\text{Cov}(C_{\ell}^{\text{xg}}, C_{\ell'}^{\kappa g})$ part of the co-

variance are shown in Figure 5, and Figure 6 shows the correlation matrix for one of the two samples of galaxies.

We find large correlations (up to 60%) between C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ at the same ℓ . Off-diagonal correlations (between unequal ℓ), on the other hand, are small ($\leq 10\%$). To account for the fact that the inverse of the above covariance matrix is not an unbiased estimate of the inverse covariance matrix, we rescale the inverse covariance matrix by the Hartlap factor (Hartlap et al. 2007):

$$\alpha_{\rm cov} = \frac{N_s - N_{\rm bins} - 2}{N_s - 1}.\tag{6}$$

Given our 400 simulations and the 13 combined data points for C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ the Hartlap factor is approximately $\alpha_{\rm cov} = 0.96$ for the analysis of a single sample and $\alpha_{\rm cov} = 0.93$ for the joint analysis of the two galaxy samples.

Since our Gaussian simulations do not include correlations between the two galaxy samples we are unable to use these simulations to estimate the off-diagonal blocks of the covariance used in the joint analysis of both galaxy samples. Instead we approximate those blocks analytically using the fiducial input spectra used for our simulations, a theory curve for the $C_{\ell}^{g_{\text{Blue}}g_{\text{Green}}}$ -power spectrum obtained from the fixed cosmology fits we employed to obtain the input auto-spectra for our Gaussian simulations (described in Sec. 5), and a measurement of $C_{\ell}^{\kappa\kappa}$ including the reconstruction noise from our simulations. Using the Gaussian covariance module implemented in NaMaster (Alonso et al. 2019; García-García et al. 2019) we compute a theory expression for the offdiagonal covariance. Correlations between the C_{ℓ}^{gg} for different samples are small (< 20% on all scales of interest). The $C_{\ell}^{\kappa g}$ on the other hand are significantly correlated (40-60%) because this part of the covariance has a contribution proportional to the product of the cross-correlations, $C_{\ell}^{\kappa g_{\text{Blue}}} C_{\ell}^{\kappa g_{\text{Green}}}$.

In Sec. 9.6 we also present a combined analysis of the ACT DR6 cross-correlation with the unWISE galaxies and the corresponding cross-correlations with *Planck* PR4 lensing reconstructions. To estimate the covariance for this joint analysis we proceed as follows: We use a set of 480 FFP10 CMB simulations used by Carron et al. (2022). These are Gaussian simulations very similar to those used in the ACT analysis, and for which there exist corresponding lensing reconstructions obtained using the *Planck* PR4 pipeline and the corresponding noise model. In an identical manner to the process described in Sec. 5 we obtain a set of correlated Gaussian galaxy realisations allowing us to estimate the covariance for the *Planck* cross-correlation analysis.

As described in Qu et al. (2023) we also obtained corresponding ACT DR6 reconstructions by analysing the



Figure 4. The recovery of C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ on simulations when correctly accounting for the effect of mode coupling and the approximate inversion of the mode coupling matrix is better than 1% on all scales. The lightly coloured errorbars indicate our measurement errors, while the dark errorbars show the error on the mean of 400 Gaussian simulations.



Figure 5. This figure shows the size of the diagonal elements of the covariance matrix as well as the off-diagonal covariance between C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ which is substantial due to the contribution from auto-correlation of the galaxy sample.

FFP10 simulations using the ACT DR6 pipeline which uses a different, but overlapping, area of the sky and a different range of CMB scales to reconstruct the lensing convergence. These ACT simulations do not include the ACT noise model, but since instrumental and atmospheric noise in the ACT data are not correlated with *Planck* and ACT uses a lensing estimator based on crosscorrelations of independent data splits, the noise does not enter the ACT-*Planck* covariance. The omission of the ACT noise in fact improves the convergence of the cross-covariance. The partially correlated lensing reconstruction noise arising from CMB fluctuations, on the other hand, is captured by these simulations.

The correlation between the ACT and *Planck* lensing reconstructions is relatively small ($\leq 30\%$), as demon-



Figure 6. Correlations between C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ are up to 60% on large scales (here shown for the Green sample of unWISE galaxies). Correlations between different scales are small; less than 10% in magnitude. We have subtracted the diagonal elements of the correlation matrix for better legibility.

strated in Qu et al. (2023). However, the crosscorrelation covariance is substantial even though the survey areas only partially overlap, since the identical galaxy sample is used in both cross-correlations. The correlation between the galaxy auto-correlations measured on the ACT and *Planck* footprints is up to 60%, while the cross-correlations are up to 50% correlated. As with the ACT-only analysis we analytically estimate the covariance between different galaxy samples, which is small in the case of C_{ℓ}^{gg} (< 20%) but again substantial for $C_{\ell}^{\kappa g}$ (up to 50%). We show a full summary of the level of off-diagonal correlations in Appendix A.

To assess the convergence of our covariance we produce an independent set of 400 further simulations. We find that the signal to noise is changed by $\lesssim 5\%$ when computing the covariance using this set of simulations. In addition we perform a consistency test with this set of simulations finding that our results are stable (see Sec. 9.7). We thus conclude that our covariance matrix is sufficiently converged.

6.1. Marginalisation over uncertainty in the lensing normalisation

As discussed in more detail in Qu et al. (2023) (see in particular Appendix B there) the unnormalised lensing reconstruction is sensitive to the product of the lensing deflection field, $\phi_{\ell m}$, and the lensing response function, \mathcal{R}_{ℓ} . The lensing normalisation, \mathcal{R}_{ℓ}^{-1} , is in principle a function of the true underlying CMB two-point spectrum, but is in practice computed assuming a fiducial spectrum, $C_{\ell}^{\text{CMB,fid}_{14}}$. While the CMB power spectrum is well constrained by *Planck*, some residual uncertainty remains that must be propagated to the lensing normalisation. Any deviation from the fiducial spectrum is expected to be small and we thus Taylor expand the normalisation around the fiducial value

$$\mathcal{R}_{L}^{-1}|_{C_{\ell}^{\text{CMB}}} \approx \mathcal{R}_{L}^{-1}|_{C_{\ell}^{\text{CMB,fid}}} \times \left[1 + M_{L}^{\ell}(C_{\ell}^{\text{CMB}} - C_{\ell}^{\text{CMB,fid}})\right],$$
(7)

where $M_L^{\ell} = \partial \ln \mathcal{R}_L^{-1} / \partial C_{\ell}^{\text{CMB}}|_{C_{\ell}^{\text{CMB, fid}}}$ is the linearised normalisation-correction matrix. As in Qu et al. (2023) we sample $1000 \text{ } \Lambda \text{CDM} \text{ } \text{CMB} \text{ power spectra from the}$ ACT DR4 + Planck parameter chains presented in Aiola et al. (2020) to generate an ensemble of smooth power spectrum curves consistent with the ACT DR4 + Planckpower spectrum measurements. Using Eq. 7, we then propagate the scatter in these power spectra to additional uncertainties in the cross-correlation bandpowers. Qu et al. (2023) find this approach to be robust despite the use of smooth ΛCDM power spectra and the use of spectra drawn from the ACT DR4 + Planckchains. We note that at the time of writing the ACT DR6 CMB power spectra are not yet available so that a direct marginalisation over the underlying spectra is not possible. In addition, the effect is significantly smaller for the cross-correlation than for the auto-spectrum so that we do not expect to incur any bias due to our treatment of the normalisation marginalisation.

In Figure 7 we show the increase in the bandpower errorbars; the change is less than 1%. For consistency with



Figure 7. The contribution to the bandpower errors from marginalisation over the lensing normalisation is negligible. Nevertheless, because the contributions to the off-diagonal elements of the covariance are more significant, we include this effect in our analysis.

the analysis in Qu et al. (2023), and because the contributions to the off-diagonal elements of the covariance are slightly more significant, we still include these contributions in the covariance. We also implement the normalisation marginalisation self consistently for the covariance between different galaxy samples and the covariance between the ACT and *Planck* cross-correlations. However, accounting for the uncertainty in the normalisation has no impact on our parameter constraints.

7. TESTS FOR SYSTEMATIC ERRORS

To test that our data is free from significant contamination that would affect our cosmological constraints, we perform a series of consistency checks. The tests are aimed at systematic errors in the lensing reconstruction that may correlate with the galaxy distribution; they are summarised in Sec. 7.1, with more details provided in Appendix B. Sec. 7.2 summarises the consistency tests performed that target, in particular, spatial inhomogeneities in the galaxy samples. In addition, we estimate biases in the lensing cross-spectra due to extragalactic foreground contamination of the lensing reconstruction (see Sec. 7.3).

7.1. Testing for Contamination of the Lensing Reconstruction

We perform a series of null-tests to demonstrate that our lensing reconstruction is free from systematic effects that could correlate with the galaxy samples. Such contamination could arise, for example, from the thermal Sunyaev-Zeldovich (tSZ) effect which originates from inverse Compton scattering of CMB photons off thermal electrons in galaxy clusters; it can, if not appropriately

¹⁴ The ACT lensing reconstruction adopts as the fiducial spectrum the Λ CDM model from *Planck* 2015 TTTEEE cosmology with an updated τ prior as in Calabrese et al. (2017).

mitigated, bias the lensing reconstruction (MacCrann et al. 2023). The tSZ signal also correlates with the large scale matter distribution and hence the galaxy density. Similarly, CIB contamination, which originates from unresolved dusty galaxies, also correlates with the galaxy densities and can bias the lensing reconstruction. The ACT DR6 lensing reconstruction used in this work adopts bias hardening methods to mitigate these biases (for details see MacCrann et al. 2023; Qu et al. 2023). We verify here that contaminants are mitigated sufficiently and that potential residual contamination does not introduce significant biases in our analysis. In addition, there is, in principle, the potential of bias due to Galactic contamination arising from a correlation of galaxy density with Galactic structure that also impacts the lensing reconstruction; this effect is easily tested and is mitigated by both masking of dust-contaminated regions and by the systematics weighting described in Sec. 3.1.3.

All our null-tests are summarised within Appendix B, where we show χ^2 and PTEs in Table 4 and several figures showing the null-test bandpowers (see Figures 28 through 33). We define a null-test to be 'passing' if it returns a PTE > 0.05. Overall we consider our tests to be passing if the number of failures (PTE < 0.05) is consistent with expectations due to random fluctuations.

We observe a total of four failures¹⁵; given that we perform a total of 28 null-tests for $C_{\ell}^{\kappa g}$ across the two samples¹⁶. Blue and Green, this is slightly more than would be expected due to random fluctuations when neglecting correlations between the tests. We note, however, that all failing tests concern the Blue sample of galaxies and at least three of them are highly correlated. To estimate the number of expected failures given the correlations between the null-test we performed these null-tests on our 400 Gaussian simulations discussed in Sec. 5. We find that in 11% of cases we observe more than the 4 failures found on the data. We conclude that the number of observed failures is not inconsistent with what could be expected due to random fluctuations. Nevertheless, we investigate below whether there is additional evidence that would suggest that the observed failures indicate systematic contamination of our data.

The failing tests are the following: (1) bandpower difference between the cross-correlations of the unWISE galaxies with the minimum variance lensing reconstruction and the temperature only reconstruction (henceforth MV-TT, see left panel in Fig. 8), (2) the difference between the cross-correlation bandpowers between the temperature only and polarisation only reconstruction (henceforth TT-MVPOL), and (3) bandpower difference between the cross-correlations of the unWISE galaxies with the minimum variance lensing reconstruction and a reconstruction that explicitly deprojects the CIB (henceforth MV-CIBD). Additionally, we also observe PTE < 0.05 for (4) the null-test between minimum variance lensing reconstruction and polarisation only reconstruction when performed using the 40% Galactic mask (henceforth MV_GAL040-MVPOL_GAL040)¹⁷. We find the tests TT-MVPOL and MV_GAL040-MVPOL_GAL040 to be \sim 77% and $\sim 32\%$ correlated with MV-TT leading us to conclude that the joint failure of these three tests is not unexpected.

Regarding the failure of MV-CIBD we note that the CIB deprojected analysis uses a slightly different lensing mask, removing a small additional area near the Galactic plane due to problems with Galactic dust. As we discuss in more detail in Sec. 7.2 the galaxy selection is not expected to be uniform on large scales so that variations in the mask can lead to a differences in the galaxy bias, leading null-tests to fail without significant impact on the inferred cosmology. When comparing the galaxy auto-correlation on the mask used in the CIB deprojected analysis with the baseline mask we also obtain a low PTE of 0.09. It is thus likely that the observed failure in this case is at least in part due to varying galaxy selection properties which should not affect our cosmological inference.

One plausible systematic contamination that could induce the observed failures, given that all other nulltests are passing, is contamination by polarised emission from Galactic dust. Dust emission correlates with dust extinction which may also affect the galaxy samples. Weighting the unWISE galaxy number density to correct for a correlation with stellar density also significantly reduces the correlation with dust extinction (see Appendix J), but we nevertheless consider the possibility of residual contamination. Such contamination is expected to be significantly stronger near the Galactic plane. We hence perform the same tests of comparing minimum variance, temperature-only and polarisationonly reconstructions on the areas of the sky included in our 60% Galactic mask but not in the more restrictive 40% Galactic mask; most of which are at low Galactic latitude. We find all these tests are passing comfort-

 $^{^{15}}$ But no catastrophic failures with $\mathrm{PTE} < 0.01.$

¹⁶ This excludes all tests for which passing is not necessarily expected due to spatial variations in the galaxy selection which affects the galaxy bias. See Sec. 7.2.

 $^{^{17}}$ The test of MV-MVPOL on the full footprint yields $\mathrm{PTE}=0.06$

ably and hence, given their high sensitivity to dust contamination, conclude that there is no evidence for any contamination from Galactic dust emission.

In Sec. 9.7 we furthermore discuss a parameter consistency test between our baseline analysis and an analysis using the temperature only reconstruction. We find the inferred parameters to be consistent within the expected error on the difference.

To target extragalactic contamination we perform a series of tests that leverage the distinctive frequency dependence of extragalactic foregrounds. We perform two kinds of frequency level null-tests. First, we take the difference of the observed temperature and polarisation maps at 150 and 90 GHz to obtain a map containing only noise (and foregrounds) on which we perform lensing reconstruction. We then cross-correlate this map with our two galaxy samples. Secondly, we investigate the bandpower difference between cross-correlations measured using the reconstructions performed only on the 90 and 150 GHz data respectively (see Fig. 9 for both versions of this test). We find no failures for those tests.

7.2. Testing for Contamination and Homogeneity of the Galaxy Samples

To test the homogeneity of our observations we perform null-tests using different masks. We perform a test on the difference between cross-correlations using our 60% and 40% Galactic masks. Additionally, we construct a null-test that compares our baseline footprint with a footprint restricted to ecliptic latitude larger than 30 degrees. Due to the unWISE survey's scan strategy, the survey depth varies with ecliptic latitude which may affect our inference. Furthermore, zodiacal light, sun light scattering of interplanetary dust, may contaminate the galaxy selection. Finally, we also split our sample into the northern and the southern Galactic cap to perform a null-test with two completely independent samples.

All tests employing the differences between different regions of the sky are complicated by the fact that the galaxy selection is not expected to be uniform on very large scales, due to the varying WISE depth of coverage. Thus, the galaxy bias, the shot noise and the redshift distribution of our samples are expected to vary across the sky (see for example Figs. 10 and 11 in Krolewski et al. 2019). In particular when comparing C_{ℓ}^{gg} , which is measured to extremely high precision, we expect nulltests to fail due to the differences in the galaxy bias and shot noise (see e.g., Fig. 10 for the comparison of different Galactic masks). However, since we marginalise over the galaxy bias, shot noise and uncertainties in b(z)dN/dz in our analysis this does not bias our cos-

mological inference. We thus construct the quantity $(C_{\ell}^{\kappa g})^2/(C_{\ell}^{gg}-\hat{n}_{\rm shot})^{18}$ which at linear order is independent of the bias and directly proportional to S_8 , the main parameter of interest in our analysis. The shot noise is here simply estimated as the inverse of the galaxy number density in the respective footprint. None of the scales used in our analysis are shot noise dominated and hence our tests are insensitive to small misestimations in the shot noise. Fig. 11 shows the null-test for two different Galactic masks using this bias-independent quantity. When considering $(C_{\ell}^{\kappa g})^2/C_{\ell}^{gg}$, we find PTE > 0.05 for all tests, but for C_{ℓ}^{gg} alone we find several tests with $PTE \ll 0.05$ as expected. This implies that, up to differences in the galaxy bias which does not affect our cosmological inference, the samples are homogeneous, yielding consistent amplitudes for different areas on the sky.

7.3. Simulation Based Test for Extragalactic Foregrounds

As a further test for systematic contamination, in particular from extragalactic foregrounds which may contaminate the lensing reconstruction, we perform a series of tests on realistic foreground maps from the WEBSKY simulations (Stein et al. 2019, 2020). Following Mac-Crann et al. (2023) we make the assumption that any extragalactic contamination affects exclusively the observed CMB temperature so that $T = T_{\text{CMB}} + T_{\text{fg}}$, but leaves the polarisation unaffected. We do not expect contaminants such as tSZ and CIB to be polarised at a significant level for our observations, and any bright polarised sources in the DR6 data are masked and inpainted (Qu et al. 2023). Denoting the quadratic estimator used to estimate the lensing convergence field, $\hat{\kappa}$, from the two temperature fields T_A and T_B as $Q(T_A, T_B)$, and the cross-correlation with the galaxy field as $C_{\ell}^{Xg} = \langle Xg \rangle$, we obtain the bias on the crosscorrelation of the temperature only lensing reconstruction with the galaxy sample due to foreground contamination as (compare MacCrann et al. 2023)

$$\Delta C_{\ell}^{\kappa g} = 2 \left\langle Q(T_{\rm CMB}, T_{\rm fg})g \right\rangle + \left\langle Q(T_{\rm fg}, T_{\rm fg})g \right\rangle.$$
(8)

Assuming that the foregrounds are uncorrelated with the CMB temperature, which will be true to high accuracy, we can see that the only relevant bias arises from the correlation of a lensing-like signal due to foreground contamination with the galaxy sample, $\langle Q(T_{\rm fg}, T_{\rm fg})g \rangle$.

We quantify the bias due to foreground contamination in terms of the bias on the cross-correlation amplitude,

 $^{^{18}}$ For brevity we will usually neglect the shot noise subtraction in our notation.



Figure 8. We show here the bandpowers for two of our null-tests comparing different lensing reconstructions (further tests are shown in Appendix B). When comparing bandpowers between the cross-correlations of the unWISE galaxies with the minimum variance lensing reconstruction and the temperature only reconstruction, we observe a null-test failure for the Blue sample of galaxies (left). As discussed in the body of the text, the one plausible systematic contamination which could cause this behaviour is contamination from polarised Galactic dust emission. However, the failure does not show the scale dependence that is expected for such a failure, with the discrepancy increasing on larger scales, nor is the discrepancy observed in regions near the Galactic plane where this effect is expected to be most significant (right). As we discuss in the text the number of observed failures is consistent with random fluctuations.



Figure 9. To test for contamination by extragalactic foregrounds we conduct two kinds of frequency level null-tests. First, we take the difference of the observed temperature and polarisation maps at 150 and 90 GHz to obtain a map containing only noise (and foregrounds) on which we perform lensing reconstruction. We then cross-correlate this map with our two galaxy samples (left). Secondly, we investigate the bandpower difference between cross-correlations measured using the reconstructions performed only on the 90 and 150 GHz data respectively. In both cases our null-tests are passed.

 A_{\times} . Since we mostly aim to constrain, and are sensitive to, the amplitude of structure growth this captures the relevant bias for our analysis well. The bias on A_{\times} in terms of the uncertainty on this amplitude is given by

$$\frac{\Delta A_{\times}}{\sigma(A_{\times})} = \frac{\sum_{\ell,\ell'} \Delta C_{\ell}^{\kappa g} \mathbb{C}_{\ell\ell'} C_{\ell'}^{\kappa g, \text{fid}}}{\sum_{\ell,\ell'} C_{\ell}^{\kappa g, \text{fid}} \mathbb{C}_{\ell\ell'} C_{\ell'}^{\kappa g, \text{fid}}}.$$
(9)

Here $\Delta C_{\ell}^{\kappa g}$ is the foreground induced bias, i.e. the crosscorrelation spectrum between the galaxy sample and the lensing estimated from the foreground only maps. Furthermore, $C_{\ell}^{\kappa g, \text{fid}}$ is the true cross-correlation signal and $\mathbb{C}_{\ell\ell'}$ is the appropriate covariance matrix.

In addition to estimating the lensing signal from the foreground-only maps as discussed in MacCrann et al. (2023) we populate the WEBSKY halo catalogue with galaxies according to an HOD similar to the one described in Krolewski et al. (2019), with the minimum halo mass increased by 3-5% to better account for the



Figure 10. One of the tests targeting the spatial homogeneity of the sample is the comparison of our baseline footprint with a more conservative Galactic mask. While this null-test passes for $C_{\ell}^{\kappa g}$ (left) we observe a failure in C_{ℓ}^{gg} for the Green sample (right). This is not unexpected due to varying galaxy bias arising from large scale inhomogeneity of the galaxy selection. We thus investigate the approximately bias independent combination $(C_{\ell}^{\kappa g})^2/C_{\ell}^{gg}$ which we observe to be passing (see Fig. 11).



Figure 11. The combination $(C_{\ell}^{\kappa g})^2/C_{\ell}^{gg}$ is approximately independent of the linear order bias and thus less sensitive to spatial variations in the galaxy selection. Comparing this combination between our baseline mask and a more conservative Galactic mask shows consistency within the errors.

different mass resolution of the WEBSKY simulations. We then sample the resulting galaxy catalogue to match the redshift distribution of our samples. We find a match between the spectra obtained from WEBSKY and from CROWCANYON described in Krolewski et al. (2019) to within about 5% on the scales used in our cosmology analysis¹⁹.

We find that in our baseline analysis, which employs frequency coadds and geometric profile hardening (Sailer et al. 2020), the residual bias is highly subdominant and can be safely neglected; $|\Delta A_{\times}/\sigma(A_{\times})| \simeq 0.04$ for both our galaxy samples²⁰ (see Fig. 12). Additionally, explicitly deprojecting the CIB does not yield any further gains. This is consistent with findings from Mac-Crann et al. (2023) and indicates that profile hardening, despite adopting a profile appropriate for tSZ clusters, also partially accounts for contamination from CIB, as was also found in Sailer et al. (2023). (For comparison, when performing no bias hardening we find significant biases of $A_{\times}/\sigma(A_{\times}) \simeq -4.2$ and -3.1 for the Blue and Green samples respectively.) We conclude that biases due to extragalactic foregrounds are small in comparison to the statistical errors.

8. THE MODEL

In Sec. 4 we discussed the measurement of the angular power spectra of the unWISE galaxies, C_{ℓ}^{gg} and their cross-spectra with CMB lensing reconstruction from the ACT. In this section we introduce the model to fit these data when measuring cosmological parameters. We present the 3D power spectrum model and its projection in Sec. 8.1 and detail our procedure for marginalising over redshift uncertainties in Sec. 8.2. Finally, in Sec. 8.3 we test the model on N-body simulations to verify its accuracy.

8.1. Hybrid Power Spectrum Model

¹⁹ Since we are interested mainly in the amplitude of the fractional bias, an exact match to the data, in particular to the galaxy bias, is not required.

²⁰ Note that this is a upper limit on the bias since the minimum variance reconstruction we employ in our analysis also receives contributions from polarisation which contributes with a weight of about 30%.



Figure 12. We estimate biases due to extragalactic foregrounds using realistic foreground simulations from WEBSKY (Stein et al. 2019, 2020). We perform lensing reconstruction on foreground only maps using different foreground mitigation strategies and cross-correlate them with a galaxy number density maps that we obtain by populating the WEBSKY halo catalogue using a HOD. We find that our baseline analysis reduces all biases to $< 0.05\sigma$ while the analysis without any mitigation yields significant biases (up to $\sim 4.2\sigma$).

In the following four subsections we detail the different components of our model. First, in Sec. 8.1.1, we describe our model for the three dimensional clustering of galaxies and matter. Because we observe the galaxy number density and CMB lensing only in projection, we describe in Sec. 8.1.2 how we obtain the power spectra for the projected fields.

The projection kernel requires knowledge of the redshift distribution of the galaxy samples. Previously, we discussed how we measure these redshift distributions (see Sec. 3.1.2 with more details provided in Appendix K). Sec. 8.1.3 introduces the way in which we implement these redshift distribution measurements in the projected power spectra. Subsequently, in Sec. 8.1.4, we describe our model for the redshift evolution of all higher order biases and the marginalisation over their contribution to the observed power spectra. Finally, we show the relative contributions of various model components (Sec. 8.1.5).

8.1.1. 3D Power Spectra and Bias Expansion

We adopt an empirical hybrid model for the three dimensional power spectra of the clustering of galaxies, matter, and their cross-clustering (P_{gg} , P_{mm} , and P_{mg} respectively). This model combines fits to numerical simulations with beyond linear order terms from Lagrangian perturbation theory (LPT; Vlah et al. 2015, 2016; Modi et al. 2017; Chen et al. 2020a,b). We note that these additional terms beyond linear bias are necessary to ensure unbiased results, as discussed in Sec. 8.3. In particular, we adopt the HMCode model from Mead et al. (2016) as implemented in CAMB²¹ (Lewis et al. 1999; Howlett et al. 2012) and the formalism for P_{gg} and P_{gm} described in Kitanidis & White (2020) with some substitutions described below (similar to Pandey et al. 2020).

In the LPT model the bias expansion for the clustering of galaxies and their cross-clustering with matter is given by (Modi et al. 2017; Chen et al. 2020a,b; Kitanidis & White 2020)

$$P_{gg} = \left(1 - \frac{\alpha_{\text{auto}}k^2}{2}\right) P_Z + P_{1-\text{loop}}$$
(10a)
+ $b_{1,L}P_{b_1} + b_{2,L}P_{b_2} + b_{s,L}P_{b_s}$
+ $b_{1,L}b_{2,L}P_{b_1b_2} + b_{1,L}b_{s,L}P_{b_1b_s} + b_{2,L}b_{s,L}P_{b_2b_s}$
+ $b_{1,L}^2P_{b_1}^2 + b_{2,L}^2P_{b_2}^2 + b_{s,L}^2P_{b_s}^2 + P_{\text{shot noise}}$
$$P_{mg} = \left(1 - \frac{\alpha_{\text{cross}}k^2}{2}\right) P_Z + P_{1-\text{loop}}$$
(10b)
+ $\frac{b_{1,L}}{2}P_{b_1} + \frac{b_{2,L}}{2}P_{b_2} + \frac{b_{s,L}}{2}P_{b_s}$

where the $b_{X,L}$ are free bias coefficients and P_Z and $P_{1-\text{loop}}$ are the Zeldovich and 1-loop power spectra, the lowest order contributions to the theory. The P_Z , $P_{1-\text{loop}}$, and higher order contributions $(P_{b_2}, P_{b_s} \text{ etc.})$ can be computed analytically from the linear power spectrum $P_{mm,\text{lin}}$.

The term

$$P_{mm} = \left(1 - \frac{\alpha k^2}{2}\right) P_Z + P_{1-\text{loop}} \tag{11}$$

²¹ https://github.com/cmbant/CAMB

captures the dark matter contribution to the clustering (Vlah et al. 2015). The parameter α allows marginalisation over the impact from small scale physics beyond the cut-off of the theory, and is generically not assumed to be identical in the auto- and cross-clustering. This contribution is what the HMCode model emulates numerically, which motivates the following substitution

$$P_{mm} = \left(1 - \frac{\alpha k^2}{2}\right) P_Z + P_{1-\text{loop}} \to P_{mm,\text{HM}}.$$
(12)

This approach incorporates all higher order effects in the dark matter clustering.

The remaining terms in the LPT bias expansion encode the response of the galaxy clustering to the large scale matter distribution. Assuming that at lowest order galaxies are linearly biased tracers of the matter density yields

$$P_{gg} = b_{1,E}^2 P_{mm,HM} + \text{higher order terms} \quad \text{and} \quad (13a)$$
$$P_{mg} = b_{1,E} P_{mm,HM} + \text{higher order terms}, \quad (13b)$$

where $b_{1,E}$ is the lowest order Eulerian bias. It is related to the lowest order bias in the Lagrangian formalism, $b_{1,L}$, as $b_{1,E} = b_{1,L} + 1$ (we use subscripts E and L to distinguish Eulerian and Lagrangian biases). Assuming furthermore that the impact of small scale physics on the cross- and auto-clustering is identical (i.e. applying the substitution from Eq. 12 to both Eqs. 10a and 10b, setting $\alpha = \alpha_{auto} = \alpha_{cross}$) then requires the further substitutions

$$P_{b_1} \to 2P_{mm,HM}$$
 and (14a)

$$P_{b_1^2} \to P_{mm,\text{HM}},$$
 (14b)

to recover the linear bias model at lowest order.

With these substitution the galaxy-galaxy, galaxymatter and matter-matter power spectra are given in detail by

$$P_{gg}(k,z) = b_{1,E}^2(z)P_{mm,HM} + b_{2,L}(z)P_{b_2}(k,z) + b_{s,L}(z)P_{b_s}(k,z)$$

$$+ b_{1,L}(z)b_{2,L}(z)P_{b_{1}b_2}(k,z) + b_{1,L}(z)b_{s,L}(z)P_{b_{1}b_s}(k,z) + b_{2,L}(z)b_{s,L}(z)P_{b_{2}b_s}(k,z)$$
(15a)

$$+ b_{2,L}^2(z) P_{b_2^2}(k,z) + b_{s,L}^2(z) P_{b_s^2}(k,z) + P_{\rm shot noise},$$

$$P_{gm}(k,z) = b_{1,E}(z)P_{mm,HM} + \frac{b_{2,L}(z)}{2}P_{b_2}(k,z) + \frac{b_{s,L}(z)}{2}P_{b_s}(k,z), \quad \text{and} \quad (15b)$$

$$P_{mm}(k,z) = P_{mm,HM}(k,z). \quad (15c)$$

on simulations (e.g. Pandey et al. 2020). To compute the effective field theory contributions, P_{b_2} , P_{b_s} etc., we rely on the velocileptors code²². Following Chen et al. (2022b) we adopt the baryon and dark matter only power spectrum for all contributions to P_{gg} (i.e. we exclude the contribution from neutrinos which, to a good approximation, do not cluster and hence do not contribute to the galaxy clustering). For P_{mg} we use the cross power spectrum between the baryon and dark matter density on one hand and the total matter density (including neutrinos) on the other.

8.1.2. Power Spectra of Projected Fields

Since we observe the galaxy overdensity and CMB lensing only projected along the line of sight we require a model for the projected galaxy-galaxy and galaxy-CMB lensing power spectra. To obtain those spectra we employ the Limber approximation²³ (Limber 1953; LoVerde & Afshordi 2008)

$$C_{\ell}^{gg} = \int d\chi \frac{W_g^2(z)}{\chi^2} P_{gg}(k\chi = l + 1/2, z)$$
(16)
$$C_{\ell}^{\kappa g} = \int d\chi \frac{W_g(z)W_{\kappa}(z)}{\chi^2} P_{gm}(k\chi = l + 1/2, z).$$
(17)

²³ Using the implementation in the Core Cosmology Library (CCL) package (https://github.com/LSSTDESC/CCL; Chisari et al. 2019) we checked that the leading beyond-limber contribution, arising from redshift-space distortions, is less than 0.1% on all scales used in our analysis and thus safely negligible.

²² https://github.com/sfschen/velocileptors

Here χ is the comoving distance to redshift z along the line of sight. The galaxy and lensing projection kernels, W_q and W_{κ} , are given by

$$W_g(z) = H(z) \frac{dN}{dz}$$
 and (18)

$$W_{\kappa}(z) = \frac{3}{2} \Omega_m H_0^2 (1+z) \frac{\chi(\chi_{\star} - \chi)}{\chi_{\star}}.$$
 (19)

The total matter density, Ω_m , includes the density of neutrinos which are non-relativistic at low redshifts. Additionally, χ_{\star} is the comoving distance to the last scattering surface, H(z) is the Hubble rate, and $H_0 = H(z = 0)$.

In addition to the contribution from the correlation of galaxy clustering with the matter field that gives rise to the CMB lensing, the measured angular power spectra also contain relevant contributions from the lensingmagnification bias (Krolewski et al. 2019). This effect arises because individual galaxies may be gravitationally lensed and magnified (or demagnified) by foreground structure, affecting our sample selection by artificially increasing (or decreasing) the magnitude of a galaxy in a way that is correlated with the large scale structure. We use the index μ to denote quantities related to the magnification bias. We model the additional contributions as

$$C_{\ell}^{g\mu} = \int d\chi \frac{W_g(z)W_{\mu}(z)}{\chi^2} P_{gm}(k\chi = l + 1/2, z), \quad (20)$$

$$C_{\ell}^{\kappa\mu} = \int d\chi \frac{\eta \kappa(z) \eta \mu(z)}{\chi^2} P_{mm}(k\chi = l + 1/2, z), (21)$$

and

$$C_{\ell}^{\mu\mu} = \int d\chi \frac{W_{\mu}(z)W_{\mu}(z)}{\chi^2} P_{mm}(k\chi = l + 1/2, z)$$
(22)

with the lensing-magnification kernel given by (Villumsen 1995; Moessner et al. 1997; Bartelmann & Schneider 1999)

$$W_{\mu}(z) = (5s_{\mu} - 2)\frac{3}{2}\Omega_m H_0^2 (1+z) \\ \times \int_{\chi}^{\chi_{\star}} d\chi' \frac{\chi(\chi' - \chi)}{\chi'} H(z') \frac{dN}{dz'}.$$
(23)

The parameter $s_{\mu} \equiv d \log_{10} N/dm$ is the response of the galaxy number density to a change in magnitude. This parameter is measured from the data by perturbing the photometry of the unWISE galaxies and reapplying the selection criteria (see Appendix D of Krolewski et al. 2019). However, in principle the magnification bias parameter is the derivative of the number density of the true underlying galaxy population (rather than the observed one) which is unknown. In addition, Krolewski

et al. (2019) showed some variations of s_{μ} with survey depth. Therefore, we marginalise over this parameter in our analysis with a conservative 10% Gaussian prior. We note that $C_{\ell}^{\kappa\mu}$ and $C_{\ell}^{\mu\mu}$ depend on the matter-matter power spectrum, P_{mm} . The total, observed, galaxygalaxy and galaxy-CMB lensing spectra are then given by $C_{\ell}^{gg} + 2C_{\ell}^{g\mu} + C_{\ell}^{\mu\mu}$ and $C_{\ell}^{\kappa g} + C_{\ell}^{\kappa\mu}$.

8.1.3. Redshift Distribution and Bias Evolution

While the lensing projection kernel (Eq. 19) is determined by the cosmological model alone, the galaxy and magnification bias kernels (Eqs. 18 and 23) depend on measurements of the galaxy redshift distributions, dN/dz. As discussed in Sec. 3.1.2 we have two ways of measuring the redshift distribution of the unWISE samples, one relying on cross-correlations with spectroscopic tracers and one relying on cross-matching with a deep photometric sample from COSMOS.

Conveniently, cross-correlation redshifts are sensitive to the product $b_{1,E}(z)dN/dz$. We normalise those redshift distributions so that

$$W^{\rm xc}(z) = \frac{\left(\widehat{b_{1,E}(z)\frac{dN}{dz}}\right)_{\rm cross-correlation}}{b_{1,E}^{\rm eff}},\qquad(24)$$

where b_{eff} is chosen such that $\int dz W^{\text{xc}}(z) = 1$. In this context \hat{X} shall denote our observational estimate of the quantity X. Wherever the product $b_{1,E}(z)W_g(z)$ appears in our model we thus use $b_g W^{\text{xc}}(z)H(z)^{24}$ where b_g is a free parameter in our analysis over which we marginalise. In this manner we encode the correct bias evolution for all the dominant terms in C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ while allowing the amplitude of the bias to vary. This comes at the cost of introducing a dependence on the fiducial cosmology assumed when computing the crosscorrelation redshifts. In Appendix C.1 we discuss how we correct for this effect. Where $W_g(z)$ appears in a term that is not also proportional to $b_{1,E}$ we rely on the cross-match derived redshifts, defining a normalised kernel

$$W_g(z) \to W^{\rm xm}(z)H(z) = \frac{\left(\frac{\widehat{dN}}{dz}\right)_{\rm cross-match}}{N}H(z)(25)$$

with N chosen again such that $\int dz W^{\text{xm}}(z) = 1$. A special case arises for terms proportional to $b_{1,L}(z)W_g(z)$. Since $b_{1,L} = b_{1,E} - 1$ we model these terms as $b_{1,L}(z)W_g(z) \rightarrow b_g W^{\text{xc}}(z)H(z) - W^{\text{xm}}(z)H(z)$.

 $^{^{24}}$ Note the factor of the Hubble rate H(z) in the definition of $W_g(z)$ (Eq. 18).



Figure 13. We show here the coevolution relations between the higher order Lagrangian bias parameters and the lowest order bias used to obtain the redshift evolution of higher order biases. The higher order Lagrangian biases, $b_{2,L}$ and $b_{s,L}$, are given as functions of the lowest order bias, $b_{1,L}$. These relations are derived from fits to numerical simulations presented in Abidi & Baldauf (2018).

8.1.4. Redshift Evolution of Higher Order Biases and Marginalisation

In the preceeding section we discussed the inclusion of the bias evolution for the lowest order bias. The higher order biases, relevant to the LPT terms in our model, also exhibit significant redshift evolution. Keeping with the model in Krolewski et al. (2021), we capture the redshift dependence of $b_{2,L}$ and $b_{s,L}$ (in part) by defining them through co-evolution relations as functions of the lowest order bias $b_{1,L} = b_{1,E} - 1$. To obtain $b_{2,L}^{\text{co-evol.}}(b_{1,L})$ and $b_{s,L}^{\text{co-evol.}}(b_{1,L})$ we use the relations presented in Abidi & Baldauf (2018) (see Fig. 8 in Abidi & Baldauf (2018) and Fig. 13 for our fit to those measurements) which are measured from clustering of protohalos in N-body simulations. Since we do not know $b_{1,L}$ this requires assuming a fiducial redshift evolution, $b_{1,L}^{\text{fid}}$ which we adopt from Krolewski et al. (2019, 2021),

$$b_{1,L}^{\text{fid}}(z) = 0.8 + 1.2z - 1$$
 Blue (26a)

$$b_{1,L}^{\text{fid}}(z) = \max\left(1.6z^2, 1\right) - 1$$
 Green (26b)

$$b_{1,L}^{\text{fid}}(z) = \max\left(2z^{1.5}, 1\right) - 1$$
 Red (26c)

This evolution is consistent with the observed clustering and with the expected bias evolution from a simple HOD of the unWISE samples (Krolewski et al. 2019). We show these fiducial bias evolutions as well as the resulting evolution of $b_{2,L}$ and $b_{s,L}$ in Fig. 14.

In going beyond the model described in Krolewski et al. (2021), we allow some freedom in the co-evolution by adding a free offset over which we marginalise in our analysis. The higher order biases are thus given effectively as $b_{X,L}(z) = b_{X,L}^{co-evol.}(b_{1,L}^{fid}(z)) + c_{X,L}$. To set pri-



Figure 14. Here we show the fiducial redshift evolution of lowest order linear bias parameters (taken from Krolewski et al. (2019, 2021); **upper** panel) and the resulting Lagrangian bias evolution for all two unWISE samples (lower panel). The redshift evolution of the higher order biases is obtained by applying the coevolution relation shown in Fig. 13 to the fiducial lowest order bias.

ors on each $c_{X,L}$ we run our pipeline on the *N*-body simulations we use for model verification while holding the cosmology fixed. We then take the prior on $c_{X,L}$ to be a Gaussian approximation of the resulting posterior but increase the width (if necessary) such that the standard co-evolution is allowed at one sigma, i.e. $\mathcal{N}(\mu_{c_{X,L}}, \max(|\mu_{c_{X,L}}|, \sigma_{c_{X,L}})).$

Since we observe only projected spectra, redshift dependent deviations from the fiducial bias evolution can largely be absorbed in an effective amplitude. To demonstrate this we tested a rescaling of the fiducial bias evolution (in addition to the free offset), letting $b_{X,L}(z) = a_{X,L} b_{X,L}^{\text{co-evol.}}(b_{1,L}^{\text{fid}}(z)) + c_{X,L}$ where both $a_{X,L}$ and $c_{X,L}$ are free parameters. From runs on simulations we found these parameters to be nearly completely degenerate. Furthermore, we tested including an offset in the form of a third degree polynomial $(b_{X,L}(z) =$ $b_{X,L}^{\text{co-evol.}}(b_{1,L}^{\text{fid}}(z)) + c_{X,L} + \sum_{n=1}^{3} c_{X,L}^{(n)}(1+z)^n)$ after minimising over the constant offset within the prior range used in our model (while holding all other parameters fixed to their best-fit values) we find no significant deviations in the goodness of fit $(\Delta \chi^2 < 1)$ even for a bias evolution which differs significantly $(\mathcal{O}(1))$ from the fiducial one. This indicates that that the parameterisation through a single offset allows for sufficient freedom in the higher order contributions to the spectrum.

In simulations we also find the cosmology dependence of all higher order contributions to the final projected spectra (arising from P_{b_2} , P_{b_s} , $P_{b_1b_2}$, etc) can effectively be absorbed into a change in the amplitude of those model components. We thus choose to compute them at a fixed fiducial cosmology taken to be the mean cosmology found in Planck Collaboration et al. (2020a)²⁵.

8.1.5. Relative Contributions of Model Components

In Fig. 15 we show the contributions of various model components assuming the best fitting cosmology for the joint fit to the cross-correlation of the Blue and Green samples with ACT DR6 including BAO information on the matter density (as described in Sec. 9.5). The contribution of higher order terms to the total C_{ℓ}^{gg} is at most 7% for the Blue sample and 5% for the Green sample. While lensing magnification does not exceed 1% for the Blue sample it contributes up to 5% to the signal on large scales for the Green sample. For $C_{\ell}^{\kappa g}$ the contributions from higher order terms are even more subdominant (at most 2% for the Blue sample) (see Fig. 16). Lensing magnification contributes about 6% to the observed $C_{\ell}^{\kappa g}$ for the Green sample. Since these contributions to our model are subdominant we do not expect our results to be sensitive to the detailed modelling choices (e.g. the exact higher order bias evolution or the estimate of the magnification bias parameter).

In Sec. 9.7 we return to the impact of the different model components on our cosmological inference. We find that neglecting all higher order corrections only yields a small shift in S_8 ($\Delta S_8 \simeq 0.1\sigma$). However, the shift obtained is sensitive to the value of the higher order biases in our data which is a priori unknown and so we conservatively maintain those free parameters in our model. Neglecting the magnification bias terms leads to a shift of $\Delta S_8 \simeq -0.2\sigma$.

8.2. Marginalisation over Redshift Uncertainties

As discussed in Sec. 3.1.2 and Krolewski et al. (2019, 2021) the measured redshift distribution of galaxies is subject to significant statistical errors. Here we describe a method based on principle component analysis for propagating these uncertainties to our cosmological results.

In principle both our measurements of the redshift distribution, those based on cross matching with spectroscopic samples, $W^{\rm xm}(z)$, as well as those based on cross-correlations with spectroscopic surveys, $W^{\rm xc}(z)$, are affected by statistical uncertainty. However, the contribution to the uncertainty in our final spectra from any uncertainty in the cross-match redshift distributions is expected to be subdominant since those measurements only affect terms that do not depend on the galaxy bias, and these terms contribute at most at the few percent level to the total signal (see Figures 15 and 16). Hence we choose to neglect this source of uncertainty and focus instead on the uncertainty in the cross-correlation based redshift distribution measurements.

We generate 1000 realisations of $W^{\rm xc} \propto b(z)dN/dz$ as described in Krolewski et al. (2021). We draw 1000 Gaussian realisations of the cross-correlation redshift estimate using the noise covariance obtained for those measurements and interpolate the resulting samples using a B-Spline with positivity constraint before normalising them. We then perform a principal component analysis on the difference of these realisations and the normalised spline interpolation of the best fit, $\hat{W}^{\rm xc}(z)$. We show the resulting principal components in Fig. 17. In our cosmology analysis we then use

$$W^{\rm xc}(z) = \hat{W}^{\rm xc}(z) + \Delta W_0^{\rm xc}(z) + \sum_{i=1}^n c_i \Delta W_i^{\rm xc}(z) \,(27)$$

where $\Delta W_0^{\rm xc}(z)$ is the mean difference between the 1000 random realisations and the best fit and $\Delta W_i^{\rm xc}(z)$ are the *i*-th principal components. The coefficients c_i are marginalised over. We keep principal components up to n = 3 and 5 for the Blue and Green samples respectively. This *n* is chosen to account for at least 90% of the variance observed in the realisations.

The uncertainty on our measurement of the clustering redshifts is then expressed in terms of priors on the c_i . We transform the 1000 realisations above into the basis of principal components and fit the distribution of coefficients observed using skew-normal distributions (shown in Fig. 18) which we use as priors on each c_i in our analysis.

We note, that the marginalisation over the uncertainties in the redshift distribution measurements introduces a noise bias in our model prediction. This is because we enforce a positivity constraint on the spline interpolation of the cross-correlation redshift estimates. Additionally, C_{ℓ}^{gg} depends on the square of $W^{xc}(z)$ and so the final spectra are a non-linear function of this noisy quantity. The resulting noise bias can be as large as 5% for some of the components contributing to C_{ℓ}^{gg} . Our method for mitigating this bias is discussed in Appendix C.2.

8.3. Testing the Model on Mocks

²⁵ Adopting the anlysis including *TT*, *TE*, and *EE* spectra for multipoles $\ell \geq 30$, as well as the *EE* spectrum for multipoles $2 \leq \ell \leq 29$, *Planck* lensing, and BAO.



Figure 15. Contribution of various model components to C_{ℓ}^{gg} ($C_{\ell,\text{HM}}^{gg}$: HMCode matter power spectrum; $C_{\ell,\text{CLEFT}}^{gg}$: higher order terms; $C_{\ell,\text{shot}}^{gg}$: shot noise; $C_{\ell}^{g\mu}$ and $C_{\ell}^{\mu\mu}$: magnification bias). The most important contributions to the galaxy power spectrum arise from the HMCode contribution at lowest order and the shot noise. The higher order contributions contribute at most 7% and 5% of the signal for the Blue and Green samples respectively. While lensing magnification plays a negligible role for the Blue sample it contributes up to 5% of the signal on large scales for the Green sample. This figure assumes the best fitting cosmology for the joint fit to the cross-correlation of the Blue and Green samples with ACT DR6.



Figure 16. Contribution of various model components to $C_{\ell}^{\kappa g}$ ($C_{\ell,\text{HM}}^{\kappa g}$: HMCode matter power spectrum; $C_{\ell,\text{CLEFT}}^{\kappa g}$: higher order terms; $C_{\ell}^{\kappa \mu}$: magnification bias). As in the case of C_{ℓ}^{gg} the signal is dominated by the contribution from the HMCode power spectrum and has small but non-negligible contributions from lensing magnification (< 1% for Blue and about 6% for Green) and higher order contributions (up to about 2% for Blue and < 1% for Green).



Figure 17. To propagate uncertainties in the measurement of the cross-correlation redshifts to our cosmological analysis we draw 1000 Gaussian realisations of W^{xc} in a manner consistent with the noise covariance (the best fitting \hat{W}^{xc} and a subset of the noisy realisations are are reproduced here from Fig. 2 to facilitate comparison with the PCA noise components). We then perform a principal component analysis on the difference between the realisations and the best fit. The lower panel shows the principal components ($i = 1, \ldots, 4$) along with the mean difference between the best fitting \hat{W}^{xc} and the noise realisations (i = 0). In our cosmological analysis we marginalise over the principal component coefficients.



Figure 18. We obtain priors on the principal component coefficients, c_i , by transforming the 1000 realisations of W^{xc} into the basis of principal components and fit the distribution of coefficients observed using skewnormal distributions.

For model verification we adopt the CrowCanyon simulations discussed in Krolewski et al. (2021). These simulations are generated using the FastPM code (Feng et al. 2016) using 8192^3 equal mass particles in a $4096h^{-1}$ Mpc a side cubic volume with periodic boundary conditions. The resulting mass resolution is $1.1 \times 10^{10}h^{-1}M_{\odot}$. The simualtion uses a fiducial cosmology close to the mean cosmology found in Planck Collaboration et al. (2020a) with $\Omega_m = 0.3092$, $\Omega_b = 0.0496$, h = 0.677, $n_s = 0.968$, and $\sigma_8 = 0.822$.

The lensing convergence and magnification fields are computed by integrating the matter density on the lightcone and weighting by the appropriate kernels (for details see Krolewski et al. 2021). We note that since no lensing reconstruction is performed the lensing observable does not include any reconstruction noise. To obtain realistic galaxy samples a HOD tuned to reproduce the unWISE samples is applied to the halos identified in the CrowCanyon simulation. This HOD is available as part of the SIMPLEHOD package²⁶. Finally, C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ are measured from these simulations on the full sky; they hence differ from data in that no mask has been applied and consequently no mode coupling is introduced.

The total simulated volume is ~ $69h^{-3}$ Gpc³ and thus not overwhelmingly larger than the volume probed by the data (approximately by a factor of 4). Since we restrict ourselves to relatively large scales in our analysis we expect some fluctuations due to cosmic variance and sample variance. Therefore, we consider our model unbiased if we are able to recover the input cosmology to within 0.5σ when using the covariance appropriate for our observations with real data. We also perform a test in which we adopt a theory covariance appropriate for the simulation volume which additionally neglects the lensing reconstruction noise as this is not present in the mocks. For this test we still recover the input cosmology to within 0.7σ across all cosmological parameters (see Table 1).

Krolewski et al. (2021) measured b(z)dN/dz for these mock samples and found a good match to the observed clustering redshift estimates. Random realisations of $W^{xc}(z)$ were generated in a similar way as for the data assuming the same noise covariance found for the data.

We find that modelling biases on all cosmological parameters considered in our analysis are well under control, not exceeding 0.3σ in all cases (see Table 1). This

is a slight improvement over the $\sim 0.5\sigma$ modelling biases in Krolewski et al. (2021).

To establish that an alternative, much simpler choice of model, would not suffice to model the data we analyse the N-body simulations described above with alternative models. We consider two such simpler models: using only the matter power spectrum from HMCode or adopting a fixed fiducial evolution of higher order bias parameters without a free offset (as in Krolewski et al. 2021). For this test we adopt the covariance appropriate for the simulation volume to perform a maximally stringent test of the input parameter recovery. We find that both models are significantly biased, exhibiting shifts much larger than the expected $\sim 1\sigma$. A HMCode-only model underpredicts S_8 by $\sim 2.2\sigma$, while the model from (Krolewski et al. 2021) yields a value for S_8 which is about 4σ larger than the true value in the simulations. The shifts are less significant when considering a realistic data covariance ($\sim 0.3\sigma$ for HMCode-only model and $\sim 1.2\sigma$ for the model with fixed higher order biases), but because the value of the true higher order biases is unknown and could induce larger biases in our data analysis we adopt the more conservative model.

9. COSMOLOGICAL ANALYSIS

9.1. Blinding Policy

In the process of preparing the analysis presented in this paper, constraints on cosmological parameters were blinded until we could demonstrate passing a sequence of tests detailed below. We were, however, not blind to the measured spectra which, at least in the case of C_{ℓ}^{gg} had already been presented in previous work (Krolewski et al. 2019). Our process towards unblinding cosmological constraints involved the following steps:

- We tested parameter recovery with our model on N-body simulations as discussed in Sec. 8.3. Since the volume of these simulations is comparable to that of the observations, deviations from the the input cosmology are expected. Hence, we consider our model to be unbiased if we are able to recover the input cosmology to within 0.5σ when using the covariance appropriate for our observations.
- To verify that our data are not contaminated by foregrounds, including Galactic and extragalactic foregrounds, we run a series of tests described in Sec. 7. We do not perform such extensive testing for the galaxy-galaxy auto spectrum since these have been presented in previous work (Krolewski et al. 2019, 2021). However, since we added sys-

²⁶ https://github.com/bccp/simplehod/blob/master/scripts/wlen. py

	Ω_m	$\Delta\Omega_m/\sigma_{\Omega_m}$	σ_8	$\Delta\sigma_8/\sigma_{\sigma_8}$	S_8	$\Delta S_8/\sigma_{S_8}$	$S_8^{ imes}$	$\Delta S_8^{\times} / \sigma_{S_8^{\times}}$
Simulation input value	0.3092	_	0.8222	_	0.8347	_	0.8335	_
			Simulation errors					
Blue	$0.319^{+0.034}_{-0.043}$	0.19	0.825 ± 0.043	0.05	$0.8466^{+0.0091}_{-0.021}$	0.60	$0.8443^{+0.0076}_{-0.018}$	0.67
Green	$0.304_{-0.018}^{+0.016}$	-0.31	0.830 ± 0.018	0.40	$0.8350\substack{+0.0095\\-0.013}$	-0.05	$0.8344^{+0.0080}_{-0.012}$	0.01
Joint	$0.306\substack{+0.013\\-0.015}$	-0.26	0.829 ± 0.016	0.39	$0.8360\substack{+0.0068\\-0.0089}$	0.10	$0.8353\substack{+0.0059\\-0.0078}$	0.20
			ACT	$\Gamma DR6 \times un$	nWISE-like erro	rs		
Blue	$0.335\substack{+0.053\\-0.100}$	0.18	$0.819\substack{+0.083\\-0.098}$	-0.07	$0.848\substack{+0.026\\-0.041}$	0.28	$0.844_{-0.036}^{+0.024}$	0.26
Green	$0.318\substack{+0.035\\-0.043}$	0.18	$0.817\substack{+0.043\\-0.048}$	-0.14	0.837 ± 0.025	0.07	0.835 ± 0.023	0.05
Joint	$0.318\substack{+0.033\\-0.041}$	0.19	$0.816\substack{+0.042\\-0.047}$	-0.16	0.836 ± 0.020	0.08	0.834 ± 0.019	0.03

Table 1. When analysing synthetic galaxy clustering and galaxy-lensing cross-correlation observations from N-body simulations we recover the input cosmology to within better than 0.3σ when adopting the appropriate data covariance for our ACT DR6 cross-correlation analysis. With the smaller theory covariance appropriate for the (somewhat larger) volume of the simulations we still recover the input cosmology to within about 0.6σ in terms of the now substantially tighter constraints.

tematics weighting for this work, we perform a smaller number of tests to re-verify the stability of the galaxy-galaxy autospectrum to different choices of the mask.

We qualify a test to be passing if it yields a PTE greater than 0.05. We are satisfied that our mitigation strategies are sufficient if the number of failures (PTE < 0.05) is approximately consistent with what would be expected from random fluctuations when accounting for the correlations between different tests²⁷.

- As a final test we investigate the consistency of parameters recovered with different analysis choices, including different sky masks, different lensing reconstruction and bias mitigation methods, varying choices for the measurement of the cross-correlation redshifts, and for the higher order bias priors. For this purpose we run parameter inference on the data, but add random offsets to all cosmological parameters that we explicitly sample in the analysis or vary as a function of other parameters $(A_s, \Omega_m, \omega_{cdm} H_0, \sigma_8 \text{ and } S_8)$.
- Before unblinding we freeze all baseline analysis choices. This includes the range of angular scales used in C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ and all priors on cosmological as well as nuisance parameters.

9.2. Post-unblinding Changes

A few minor aspects of the analysis were corrected post-unblinding, but the impact of these change is small and they partially offset. Consequently, the inferred value of S_8 for our baseline analysis including only the galaxy auto-correlation of both redshift samples and the cross-correlations with CMB lensing from ACT is practically unaffected, changing by -0.1σ from our initial unblinded results ($\Delta S_8 = -0.002, -0.25\%$). The postunblinding changes included a modification to the treatment of mask induced mode coupling of the shot noise component and a small correction to the transfer function obtained from simulations $(+0.2\sigma)$. Additionally, the cross-covariance between the two redshift samples had erroneously been neglected (-0.3σ) . Finally, we noticed a bug in the model code which meant that the fiducial bias evolution for $b_{2,L}$ was erroneously adopted for $b_{s,L}$ as well. This correction required redetermining the priors on the higher order biases from the N-body simulations and re-verifying the model. Fortunately, the model remains unbiased and we infer the same value of S_8 for the joint analysis of the cross-correlations of the Blue and Green sample with ACT with only a 5% increase in uncertainty. More detail on these postunblinding changes is provided in Appendix L.

9.3. Parameter Inference and Priors

We obtain cosmological constraints by constructing a Gaussian likelihood

$$-2\ln\mathcal{L} \propto \sum_{bb'} \begin{bmatrix} \Delta \hat{C}_b^{gg}(\boldsymbol{\theta}) \\ \Delta \hat{C}_b^{\kappa g}(\boldsymbol{\theta}) \end{bmatrix} \mathbb{C}^{-1} \begin{bmatrix} \Delta \hat{C}_{b'}^{gg}(\boldsymbol{\theta}) \\ \Delta \hat{C}_{b'}^{\kappa g}(\boldsymbol{\theta}) \end{bmatrix}$$
(28)

where the $\Delta \hat{C}_{b}^{gg}$ and $\Delta \hat{C}_{b}^{\kappa g}$ are the residuals between our observed galaxy-galaxy and galaxy-lensing spectra,

 $^{^{27}}$ We chose to not consider tests yielding PTE >0.95 as failures, to minimise the potential for spurious failures and to focus our null-tests on systematic contamination of the data in contrast to a miss-modelling of the covariance matrix. Furthermore, the lensing noise modelling has already been tested in detail in Qu et al. (2023). We also note, after the fact, that we do not observe any PTE >0.95.

 \hat{C}_b^{gg} and $\hat{C}_b^{\kappa g},$ and the respective band window convolved theory spectra, C_b^{gg} and $C_b^{\kappa g}.$ The covariance $\mathbb C$ has the form

$$\mathbb{C} = \begin{bmatrix} \mathbb{C}^{gg-gg}_{bb'} & \mathbb{C}^{gg-\kappa g}_{bb'} \\ \left(\mathbb{C}^{gg-\kappa g}_{bb'} \right)^T & \mathbb{C}^{\kappa g-\kappa g}_{bb'} \end{bmatrix}$$
(29)

where $\mathbb{C}_{bb'}^{gg-gg}$, $\mathbb{C}_{bb'}^{\kappa g-\kappa g}$, and $\mathbb{C}_{bb'}^{gg-\kappa g}$ are the galaxy autospectrum covariance, the galaxy-lensing cross-spectrum covariance, and the cross-covariance between them. These are estimated from simulations as described above in Sec. 6.

We use the Markov chain Monte Carlo code $cobaya^{28}$ (Torrado & Lewis 2021) to infer parameters from our galaxy-galaxy and galaxy-CMB lensing data using the model described in Sec. 8. We consider chains to be converged if the Gelman-Rubin statistic (Gelman & Rubin 1992; Brooks & Gelman 1998) satisfies $R-1 \leq 0.01$.

Our data is insensitive to the optical depth to reionisation and we thus fix it to the best fit value from $Planck^{29}$, $\tau = 0.0561$ (Planck Collaboration et al. 2020a). Since we are using low redshift projected tracers alone, the information from the BAO feature in the power spectrum is largely erased and thus our data is mostly sensitive to the total matter density, Ω_m . We therefore again choose to fix $\Omega_b h^2$ to the central value from *Planck*, $\Omega_b h^2 = 0.2242$ (Planck Collaboration et al. 2020a), and vary only Ω_m . Additionally, our measurement of the lensing amplitude is largely degenerate with the distance to the unWISE redshifts which is primarily set by the Hubble parameter, h. In keeping with Krolewski et al. (2021) we break this degeneracy by fixing the projected angular size of the sound horizon to the value measured from the CMB by the Planck Collaboration, $\theta_{\rm MC}$. It can be shown that $\theta_{\rm MC}$ is predominantly sensitive to the product $\Omega_m h^3$ (Percival et al. 2002). Hence, we fix this combination to $\Omega_m h^3 = 0.09635$, the mean value obtained by the Planck Collaboration (Planck Collaboration et al. 2020a). It should be noted that this combination is largely independent of the complex physics and possible observational systematics that affect the broadband shape of the CMB. It is determined to approximately 0.3% by *Planck* so that any uncertainty on this combination would be subdominant to our measurement uncertainties. Finally, we also fix

the tilt of the primordial power spectrum to $n_s = 0.9665$ (Planck Collaboration et al. 2020a) and assume the minimum neutrino mass allowed in the normal hierarchy ($\sum m_{\nu} = 0.06 \text{ eV}$). We explore an alternative set of priors on cosmological parameters adopted from the ACT DR6 lensing power spectrum analysis (Qu et al. 2023) in Appendix G. We show that these alternative priors do not significantly affect any of our conclusions.

Our procedure for setting priors on the higher order bias parameters, $c_{X,L}$, and the redshift marginalisation parameters, $c_{dN/dz PCA,i}$, are described in detail in Sections 8.1 and 8.2, respectively, and the resulting priors are summarised in Table 2. We note that the priors on the redshift marginalisation parameters shown here are the ones adopted for our baseline analysis which includes both the full spectroscopic data set from BOSS and eBOSS and the appropriate weighting of the unWISE data to determine the cross-correlation redshifts (see Sec. 3.1.2). For some of our consistency tests we adopt different subsets of the spectroscopic data or neglect the systematics weighting when measuring the cross-correlation redshifts; in each case we adopt the priors appropriate for the specific cross-correlation measurement used which can differ somewhat from the ones shown here. As in Krolewski et al. (2021) we adopt 10%Gaussian priors on the measured lensing magnification bias and Gaussian priors with standard deviation 0.2 on the logarithm of the shot noise, $\log_{10} P_{\text{shot}}$ (translating to a prior of approximately 60% on the amplitude of the shot noise).

9.4. BAO Likelihoods

Weak lensing measurements depend primarily on the amplitude of matter fluctuations σ_8 and the matter density Ω_m . In order to break the degeneracies of our σ_8 constraint with the latter parameter and allow for more powerful comparisons with other probes (including the primary CMB, optical weak lensing measurements, and constraints from CMB lensing auto-spectra), we include information on Baryon Acoustic Oscillations from the 6dF and SDSS surveys. We adopt the same likelihoods used in Madhavacheril et al. (2023). The data we include measure the BAO signature in the clustering of galaxies with samples spanning redshifts up to $z \simeq 1$, including 6dFGS (Beutler et al. 2011), SDSS DR7 Main Galaxy Sample (MGS; Ross et al. 2015), BOSS DR12 LRGs (Alam et al. 2017), and eBOSS DR16 LRGs (Alam et al. 2021). We do not use the higher-redshift ELGs (Comparat et al. 2016), Lyman- α (du Mas des Bourboux et al. 2020), and quasar samples (Alam et al. 2021), though we hope to include these in future analyses. We only include the BAO information from these surveys (which for our

²⁸ https://github.com/CobayaSampler/cobaya

²⁹ We adopt the results obtained in Planck Collaboration et al. (2020a) by including the TT,TE, and EE spectra for multipoles $\ell \geq 30$, the EE spectrum for lower multiples $2 \leq \ell \leq 29$ ("lowE" likelihood), and additionally *Planck* lensing and BAO (Planck Collaboration et al. 2020a, see Table 2 therein).

Parameter	Prior						
Sampled cosmological parameters							
Ω_m	[0.01, 0.95]						
$\ln(10^{10}A_s)$	[1.0, 4.0]						
Fixed and deriv	ed cosmological parameters						
n_s	fixed (0.9665)						
au	fixed (0.0561)						
$\Omega_m h^3$	fixed (0.09635)						
$\Omega_b h^2$	fixed (0.02242)						
$\sum m_{ u}$	fixed (0.06 eV)						
h	derived $\left(\left[\Omega_m h^3 / \Omega_m\right]^{1/3}\right)$						
$\Omega_c h^2$ deriv	ved $(\Omega_m h^2 - \Omega_b h^2 - \sum m_\nu / 93.14 \text{ eV})$						
Galaxy model p	arameters						
Blue sample (\bar{z} :	= 0.6)						
$b_{1,E}^{\mathrm{Blue}}$	[0.5, 3.9]						
$\log_{10} P_{\rm shot}^{\rm Blue}$	$\mathcal{N}(-7.05, 0.2)$						
$s_{\mu}^{ m Blue}$	$\mathcal{N}(0.455, 0.0455)$						
$c_{2,L}^{\text{Blue}}$	$\mathcal{N}(0.55, 0.55)$						
$c_{s,L}^{\text{Blue}}$	$\mathcal{N}(0.17, 0.29)$						
$c_{dN/dz \text{ PCA},1}^{\text{Blue}}$	$\mathcal{N}_s(-0.58, 0.86, 1.61)$						
$c_{dN/dz \ PCA,2}^{ m Blue}$	$\mathcal{N}_s(-0.40, 0.54, 2.94)$						
$c_{dN/dz \ PCA,3}^{ m Blue}$	$\mathcal{N}_s(-0.14, 0.23, 1.27)$						
Green sample (2	$\bar{z} = 1.1$)						
$b_{1,E}^{\text{Green}}$	[0.7, 4.2]						
$\log_{10} P_{\rm shot}^{\rm Green}$	$\mathcal{N}(-6.79, 0.2)$						
$s_{\mu}^{ m Green}$	$\mathcal{N}(0.653, 0.0653)$						
$c_{2,L}^{\text{Green}}$	$\mathcal{N}(0.42, 0.42)$						
$c_{s,L}^{\text{Green}}$	$\mathcal{N}(0.22, 0.50)$						
$c_{dN/dz \ PCA,1}^{ m Green}$	$\mathcal{N}_s(-0.30, 0.57, 0.88)$						
$c_{dN/dz \ PCA,2}^{ m Green}$	$\mathcal{N}_s(-0.30, 0.54, 0.99)$						
$c_{dN/dz \ { m PCA},3}^{ m Green}$	$\mathcal{N}_s(-0.31, 0.44, 1.90)$						
$c_{dN/dz \ { m PCA},4}^{ m Green}$	$\mathcal{N}_s(-0.09, 0.26, 0.51)$						
$c_{dN/dz \ { m PCA},5}^{ m Green}$	$\mathcal{N}_s(-0.24, 0.33, 2.65)$						

Table 2. Parameters and priors used in this work. $\mathcal{N}(\mu, \sigma)$ indicates a Gaussian prior with mean μ and variance σ^2 , while $\mathcal{N}_s(\xi, \omega, \alpha)$ indicates a skew-normal prior with location ξ , scale ω and shape α . Uniform priors are indicated by square brackets. The priors on the magnification bias parameters are adopted from Krolewski et al. (2019), the procedure for obtaining priors on the free parameters in the higher order bias evolution is described in Sec. 8.1.4, and the priors on the redshift marginalisation parameters are described in Sec. 8.2.

purposes primarily constrains Ω_m) and do not include the structure growth information in the redshift-space distortion (RSD) component of galaxy clustering. We make this choice so as to isolate information on structure formation purely from the cross-correlation alone.

9.5. Constraints on Cosmological Parameters from the cross-correlation of ACT DR6 Lensing and unWISE

Analysing jointly the auto-correlation of the two samples of unWISE galaxies and the cross-correlation of each with the ACT DR6 lensing reconstruction we obtain a 2.6% constraint on $S_8 \equiv \sigma_8 (\Omega_m/0.3)^{0.5}$ of

$$S_8 = 0.813 \pm 0.021. \tag{30}$$

The best-constrained parameter in our analysis actually differs slightly from S_8 and we empirically determined it to be closer to $S_8^{\times} \equiv \sigma_8 (\Omega_m/0.3)^{0.45}$ which we constrain to 2.3% ($S_8^{\times} = 0.817 \pm 0.019$). All parameter constraints are summarised in Table 3, with the posteriors on cosmological parameters shown in Fig. 19.

We note that when analysing only a single tomographic sample without additional information to break the degeneracy between σ_8 and Ω_m these parameters are individually only poorly constrained. While the marginalised one dimensional posteriors on Ω_m and σ_8 obtained from the Blue and Green sample may visually appear to be in some tension, the two samples are consistent within 1.7σ in the full σ_8 - Ω_m parameter space³⁰. We discuss the consistency of the two redshift samples in more detail in Appendix E.

We can break the degeneracy between σ_8 and Ω_m through the addition of external information. In particular, we combine our analysis with publicly available BAO likelihoods described in Sec. 9.4 which primarily constrain the matter density, Ω_m . In the combination with BAO we are able to place competitive constraints on σ_8 alone. With the degeneracy broken the amplitude of the spectra becomes the dominant source of information on σ_8 . We find that the combination of the crossand auto-correlations of the Blue and Green samples with BAO yields

$$\sigma_8 = 0.813 \pm 0.020. \tag{31}$$

The change in S_8 inferred from the joint analysis of the Blue and Green cross-correlations using the alternative priors from Appendix G is small $(0.3\% \text{ or } 0.14\sigma)$; we find $S_8 = 0.810 \pm 0.021$ indicating that the choice of alternative priors does not significantly affect our constraining power on S_8 . When including BAO we find $\sigma_8 = 0.813 \pm 0.022$ for this alternative set of priors, showing no shift from the mean value obtain with our baseline analysis, but 10% wider errors.

³⁰ We used the TENSIOMETER package to estimate the significance (https://github.com/mraveri/tensiometer)

UNWISE X ACT DR6 COSMOLOGY

	Ω_m	σ_8		S_8		$S_8^{ imes}$			
	ACT DR6 \times unWISE only								
Blue	$0.523^{+0.093}_{-0.14}$	(0.476)	$0.654_{-0.074}^{+0.058}$	(0.659)	$0.849^{+0.024}_{-0.040}$	(0.831)	$0.827^{+0.021}_{-0.035}$	(0.812)	
Green	$0.259\substack{+0.029\\-0.037}$	(0.245)	$0.868\substack{+0.041\\-0.046}$	(0.878)	$0.803\substack{+0.025\\-0.028}$	(0.793)	$0.809\substack{+0.022\\-0.025}$	(0.801)	
Joint	$0.279^{+0.028}_{-0.035}$	(0.276)	$0.843\substack{+0.038\\-0.044}$	(0.848)	0.813 ± 0.021	(0.813)	0.817 ± 0.019	(0.816)	
			ACT I	$DR6 \times un$	WISE + BAO				
Blue	0.3102 ± 0.0079	(0.310)	0.806 ± 0.028	(0.808)	0.819 ± 0.027	(0.821)	0.818 ± 0.027	(0.820)	
Green	0.3063 ± 0.0076	(0.306)	0.818 ± 0.022	(0.816)	0.826 ± 0.022	(0.824)	0.825 ± 0.021	(0.823)	
Joint	0.3068 ± 0.0077	(0.306)	0.813 ± 0.020	(0.811)	0.822 ± 0.018	(0.819)	0.822 ± 0.018	(0.818)	
		L	ACT DR6 \times un	WISE +	Planck PR4 \times 1	inWISE			
Blue	$0.440^{+0.056}_{-0.082}$	(0.424)	0.692 ± 0.051	(0.690)	0.830 ± 0.021	(0.820)	0.815 ± 0.019	(0.806)	
Green	$0.263^{+0.020}_{-0.023}$	(0.275)	0.861 ± 0.029	(0.857)	0.804 ± 0.019	(0.793)	0.809 ± 0.017	(0.799)	
Joint	$0.274_{-0.022}^{+0.019}$	(0.264)	0.849 ± 0.029	(0.860)	0.810 ± 0.015	(0.808)	0.814 ± 0.014	(0.813)	
$ACT DR6 \times unWISE + Planck PR4 \times unWISE + BAO$									
Blue	0.3109 ± 0.0079	(0.309)	0.801 ± 0.021	(0.805)	0.816 ± 0.019	(0.817)	0.814 ± 0.019	(0.816)	
Green	0.3039 ± 0.0074	(0.303)	0.818 ± 0.017	(0.818)	0.823 ± 0.016	(0.821)	0.823 ± 0.016	(0.821)	
Joint	0.3046 ± 0.0074	(0.305)	0.813 ± 0.015	(0.809)	0.819 ± 0.014	(0.815)	0.819 ± 0.014	(0.815)	

Table 3. Summary of the constraints on cosmological parameters obtained from the cross-correlation of unWISE galaxies with ACT DR6 lensing reconstruction. Best-fit values (maximum a posteriori) are shown in parentheses following the one dimensional marginalised constraints. The combination of the galaxy-lensing cross-correlation and the galaxy clustering auto-spectrum is primarily sensitive to the parameter S_8 . To break the degeneracy between the matter density, Ω_m and the amplitude of fluctuations σ_8 we analyse our data jointly with BAO. We also present constraints from the joint analysis of the *Planck* and ACT cross-correlations.



Figure 19. Parameter constraints from the cross-correlation of ACT DR6 lensing and unWISE galaxies. From the combination of C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ only for both samples of unWISE galaxies (left) we find $S_8 = 0.813 \pm 0.021$. As described in the test the Blue and Green sample are consistent to within 1.7σ . With additional information on Ω_m from BAO (right) the degeneracy between Ω_m and σ_8 is broken and we place competitive constraints on $\sigma_8 = 0.813 \pm 0.020$.

9.6. Combination with the cross-correlation of Planck Lensing and unWISE

In Appendix I we present a reanalysis of the crosscorrelation between *Planck* CMB lensing and the two unWISE galaxy samples studied in this work using our improved model dN/dz and systematics weights, implementing the previously neglected Monte Carlo lensing norm correction, and updating to *Planck* PR4. We find

$$S_8 = 0.805 \pm 0.018 \tag{32}$$

from the cross-correlation alone and

$$\sigma_8 = 0.810 \pm 0.018 \tag{33}$$

when combining with BAO. Despite the significant improvement in lensing noise for ACT these results from the cross-correlation with *Planck* constrain S_8 and σ_8 more tightly than the results presented in Sec. 9.5. This is due to the significantly smaller sky area in ACT (~ 60% vs. ~ 20%). Since the results from ACT and *Planck* are consistent we also present a joint analysis of the cross-correlation of unWISE galaxies with ACT DR6 and *Planck* PR4 lensing.

In Sec. 6 we described how we estimate the cross-covariance between the ACT and *Planck* cross-correlations. We established in Sec. 7.2 that the galaxy nuisance parameters, principally the linear galaxy bias, are expected to vary across the sky. Hence we do not assume these to be identical in the ACT and *Planck* foot-prints. We therefore include C_{ℓ}^{gg} measured with both the ACT and *Planck* masks in our analysis and independently marginalise over the nuisance parameters for both sets of observations. We include the significant cross-covariance between $C_{\ell}^{gg,ACT}$ and $C_{\ell}^{gg,Planck}$ which is also estimated from our Gaussian simulations and correctly captures the overlap in area of both observations.

The combination of ACT DR6 and *Planck* PR4 crosscorrelations with unWISE yields a joint constraint

$$S_8 = 0.810 \pm 0.015, \tag{34}$$

a 29% and 17% improvement over the constraints from ACT DR6 and *Planck* PR4 alone respectively. In combination with BAO we find

$$\sigma_8 = 0.813 \pm 0.015, \tag{35}$$

again improving the individual constraints from ACT(+BAO) and Planck(+BAO) by 25% and 17% respectively. The joint constraints are shown alongside the constraints from ACT and *Planck* alone in Fig. 20.

9.7. Parameter Based Consistency Test

As discussed in Sec. 9.1 we performed a series of parameter level consistency tests on the ACT data by running our full analysis pipeline using different lensing reconstruction methods and data subsets. These tests were originally performed blind by consistently adding random offsets to the cosmological parameters. We present here the unblinded versions of these tests. The alternative analyses using simplified models, discussed towards the end of this section, were not performed prior to unblinding. The results of these tests are summarised visually in Fig. 21.

We compare the following data subsets to our baseline analysis: 1) cross-correlations measured using the temperature only lensing reconstruction, 2) an analysis restricted to ecliptic latitudes greater than 30°, 3) an analysis using a more restrictive Galactic mask which includes only the 40% of the sky with the lowest Galactic contamination, 4) two analyses further restricting the range of scales used in our analysis, once by increasing $\ell_{\rm min}$ to 150 and 100 for C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ respectively, and once by reducing $\ell_{\rm max}$ to 300 for both C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$, 5) an analysis using twice wider priors on the higher order bias parameters, and 6) an analysis using a more restrictive scale cut for the cross-correlation redshift estimation (minimum scale of 4 h^{-1} Mpc rather than 2.5 h^{-1} Mpc).

We also produce a comparison between measurements conducted exclusively on the northern and southern Galactic regions (NGC and SGC). These are fully independent as they also rely on measuring the crosscorrelation redshifts on the respective regions.

We estimate the statistical likelihood of observing the shifts obtained by considering the shift in the best constrained parameter S_8^{\times} , effectively assuming that we are only measuring a single parameter related to the amplitude of the spectrum. For the comparison of nested datasets, i.e. such datasets where one is a proper subset of the other, we estimate the statistical uncertainty on the expected shift in parameters as the difference between the marginalised posterior width of each of the samples approximating them as Gaussian (Gratton & Challinor 2020). In the case where both datasets are independent we add the variance of both measurements. Detailed results of all consistency tests are provided in Table 6 in Appendix F.

We also conduct three variations of our baseline analysis that do not simply consider subsets of the full data set (also shown in Fig. 21). First, we adopt a lensing reconstruction in which the contamination from CIB has explicitly been deprojected using high frequency data from *Planck*. Secondly, we centre the higher order bias parameters on the fiducial coevolution relations rather



Figure 20. We combine the cross-correlations of unWISE galaxies with lensing reconstructions from ACT DR6 and *Planck* PR4 at the likelihood level to obtain further tightened constraints on cosmological parameters. Here we show a comparison of the joint analysis of both redshift samples from ACT only, *Planck* only and the combination of ACT and *Planck*. As above, the **left** panel shows constraints using C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ only, for which we find $S_8 = 0.810 \pm 0.015$. Including BAO information (**right**) yields $\sigma_8 = 0.813 \pm 0.015$.

than the best fit from simulations. Finally, we use a slightly different fiducial cosmology to measure the cross-correlation redshifts, with h = 0.702, $\Omega_m = 0.278$ and $\sigma_8 = 0.805$. For these tests an estimate of the significance of the shifts observed cannot easily be obtained due to the non-trivial covariance between these measurements and our baseline analysis. However, the shifts in the parameters S_8 and S_8^{\times} are small compared to our measurement uncertainties ($< 0.35\sigma$; see Table 7 within Appendix F). As discussed in Sec. 6 we also conduct a stability test using an independent set of simulations to estimate our covariance. This test is also shown in Fig. 21. We find that our results are unchanged with this independent covariance estimate.

All consistency tests pass and the shifts in the parameters of interest are small compared to our uncertainties. This indicates that our baseline cosmology results are unlikely to be significantly affected by remaining systematic contamination.

We note that the null-test failures discussed in Sec. 7.1 indicate a slightly larger lensing amplitude preferred by the temperature-only data. We showed that the difference is consistent with random fluctuations, nevertheless we assessed the impact on cosmological parameters. Using the temperature-only lensing reconstruction leads to an increase in the inferred S_8 compared to our baseline

analysis of $\Delta S_8 = 0.014$ (+0.7 σ ; see Table 6). The shift is consistent within the expected error on the difference between the baseline and temperature-only analysis, and increases the discrepancy with S_8 inferred from galaxy surveys (see comparisons in Sec. 10.2).

In Fig. 22 we also show the impact of neglecting some of the model components in our analysis and find this to have a small impact on the inferred values of S_8^{\times} (< 1 σ in terms of our statistical uncertainties; these analyses are also summarised in Table 7). While analyses with these models do not appear strongly biased, they also do not dramatically improve our constraining power.

Finally, we compare model predictions for the crosscorrelation between the two galaxy samples with the observed $C_{\ell}^{g_{\text{Blue}}g_{\text{Green}}}$. We find that the model prediction, fit only to the two galaxy auto-spectra and the crosscorrelations with the ACT DR6 lensing reconstruction, is consistent with the observed cross-spectrum within unc (see Appendix H for details).

10. DISCUSSION AND CONCLUSIONS

Measuring the amplitude of low redshift fluctuations can reveal the nature of dark matter and dark energy, measure the masses of neutrinos and provide crucial tests of gravity. The cross-correlation of CMB lensing maps with matter tracers (such as galaxies in this work) with a known redshift distribution – or CMB lensing to-



Figure 21. We tested various analyses of data subsets and alternative choices for higher order bias parameters and hyperparameters relating to the measurement of the cross-correlation redshifts. Here we show the impact on the best constrained parameter, S_8^{\times} , of these alternative analysis choices. We note that these results are not independent and where possible we estimate the likelihood of the shifts observed in Table 6. More consistency tests are also summarised in Table 7. We find consistency across all our tests, indicating that our results are unlikely to be significantly affected by residual systematic contamination.



Figure 22. The impact of neglecting various components of our model is very small with all resulting shifts well below 1σ in terms of our statistical uncertainty. This indicate that our results will only have a small dependence on the detailed modelling choices for those model components.

mography – can isolate the redshifts of interest, study the time-dependence of the signal, break the degeneracy with galaxy bias and reduce the sensitivity to several sources of systematics. In this paper, we leveraged the low shot noise and large redshift lever arm of the un-WISE Blue and Green galaxy samples, which fully cover the dark energy dominated era as well as the transition to matter domination. Our results improve and build upon a previous analysis in Krolewski et al. (2021), and the main differences are highlighted in Appendix I.

10.1. Comparison with predictions from the primary CMB

By combining galaxy clustering and cross-correlation measurements, we find $S_8 = 0.813 \pm 0.021$ using ACT lensing, and $S_8 = 0.810 \pm 0.015$ with a combination of *Planck* and ACT lensing, consistent with the prediction from the primary CMB from *Planck* at the 0.8σ and 1.1σ level respectively (*Planck* obtains $S_8 = 0.834 \pm 0.016$ or $S_8 = 0.832 \pm 0.013$ with CMB lensing; Planck Collaboration et al. 2020a) and with similar statistical uncertainty. BAO data breaks the degeneracy between σ_8 and Ω_m and we find $\sigma_8 = 0.813 \pm 0.020$ from ACT alone and $\sigma_8 = 0.813 \pm 0.015$ from the combination of *Planck* and ACT. Again, this is in very good agreement with the *Planck* primary CMB results of $\sigma_8 = 0.8120 \pm 0.0073$ $(0.05\sigma \text{ and } 0.07\sigma \text{ respectively})$ using primary CMB only (or $\sigma_8 = 0.8111 \pm 0.0060$ when including CMB lensing; Planck Collaboration et al. 2020a). The comparisons with the *Planck* results is shown in Figs. 23 along with other large scale structure measurements of S_8 and σ_8 . Our results are also consistent with predictions from independent primary CMB observations, including from the combination of WMAP and ACT (Aiola et al. 2020). We show those constraints in Figs. 24 and 25 together with a more comprehensive set of other measurements.

10.2. Comparison with other measurements of the large scale structure of the universe

In addition to the comparison to predictions for the amplitude of structure formation from the primary CMB (using *Planck* and ACT) we also compare our results to measurements from other large scale structure observables. We include results from CMB lensing autospectrum analyses (Bianchini et al. 2020; Carron et al. 2022; Qu et al. 2023; Madhavacheril et al. 2023), galaxy weak lensing surveys (DES, KiDS and HSC; Abbott et al. 2022; Heymans et al. 2021; More et al. 2023; Sugiyama et al. 2023; Miyatake et al. 2023), from other cross-correlations with CMB lensing from ACT, SPT, and *Planck* (White et al. 2022; Chen et al. 2022; Chang et al. 2023; Abbott et al. 2023; Marques et al. 2023),

and from the three dimensional clustering of galaxies (Alam et al. 2021; Ivanov et al. 2023). We subsequently briefly introduce the datasets we employ:

- CMBL: These are constraints from the analysis of the auto-spectrum of CMB lensing reconstructions. These results are mostly sensitive to linear scales at z = 1 - 2 and primarily constrain the parameter combination $\sigma_8 \Omega_m^{0.25}$. We include results from *Planck* PR4 (Carron et al. 2022), SPTpol (Bianchini et al. 2020), ACT DR6, and a joint analysis of ACT DR6 and *Planck* PR4 (Madhavacheril et al. 2023). To make fair comparisons we combine the CMB lensing measurements with BAO information which breaks the degeneracy between σ_8 and Ω_m .
- WL: From galaxy weak lensing surveys we include constraints from '3x2pt'-analyses, combining measurements of galaxy clustering, galaxy shear, and their cross-correlation. For DES we adopt the results obtained in Abbott et al. (2022) when fixing the neutrino mass. For KiDS we show results presented in Heymans et al. (2021). We note, that in contrast to the DES analysis these results are obtained from the combination of projected galaxy shear and galaxy-galaxy lensing measurements with a measurement of the three dimensional clustering of galaxies in the spectroscopic BOSS and 2dfLenS surveys. Therefore, the KiDS analysis already contains the BAO information. For HSC we adopt a set of results from a reanalysis of the HSC galaxy shear, galaxy-galaxy lensing and galaxy clustering measurements (More et al. 2023) using priors on cosmological parameters consistent with those used in this work (see Table 2). We show results for an analysis using a linear bias model (Sugiyama et al. 2023) and from an analysis using a halo model based emulator which includes non-linear scales (Miyatake et al. 2023).
- CMBLX: We compare our results with other cross-correlations between CMB lensing and galaxy surveys. White et al. (2022) analysed the cross-correlation between DESI LRG targets and a lensing reconstruction from *Planck* PR3 (Planck Collaboration et al. 2020b). We also include a series of works cross-correlating DES-Y3 galaxy shear and galaxy clustering ($\delta_g + \gamma$) with a joint SPT and *Planck* lensing reconstruction (Chang et al. 2023). Building on this work Abbott et al. (2023b) present a '5x2pt' and '6x2pt' analysis combining with all DES internal cross- and



Figure 23. Here we show constraints in the $\Omega_m - \sigma_8$ plane for our baseline results and selected other datasets. The left panel shows results without external information on Ω_m ; all large scale structure probes show a significant degeneracy between Ω_m and σ_8 . For our analysis the best constraint parameter combination is proportional to $\sigma_8 \Omega_m^{0.45}$ shown here as a dotted line. This degeneracy is broken by the addition of BAO information. To illustrate the effect of BAO information we show the joint cross-correlation analysis of ACT and *Planck* with BAO as a dashed contour (in the left panel). The right panel shows a comparison with other datasets when adding BAO information. We can see that our analysis is in no significant tension with any one of the other datasets shown, when considering the full $\Omega_m - \sigma_8$ parameter space, despite favouring a larger value of S_8 and σ_8 compared to galaxy weak lensing surveys. Our results are consistent with predictions from the primary CMB and CMB lensing.

auto-correlations and additionally with the lensing auto-spectrum respectively. The joint lensing reconstruction from SPT-SZ and *Planck* is presented in Omori et al. (2023). A recent crosscorrelation between DES-Y3 clustering (δ_g) and ACT DR4 lensing (Marques et al. 2023) based on the lensing reconstruction from Darwish et al. (2021) is also included. While all aforementioned analyses use photometric galaxy samples in their cross-correlations, the final cross-correlation study included in our comparisons, Chen et al. (2022b), jointly models the three dimensional clustering of the spectroscopic BOSS galaxies and their crosscorrelation with *Planck*.

• GC: Finally, we compare our results with constraints from the three dimensional clustering of galaxies as measured by BOSS and eBOSS. We include the analysis of BAO and RSD from Alam et al. (2021). Furthermore, we include a independent analysis based on the effective theory of large scale structure (EFTofLSS) that fits the 'full shape' of the power spectrum and bispectrum measured in redshift-space (Ivanov et al. 2023). Similar, previous analyses found compatible results (see e.g., D'Amico et al. 2020; Ivanov et al. 2020; Chen et al. 2022a). In the left panel of Fig. 23 we show a comparison of our results with various other probes in the $\Omega_m - \sigma_8$ parameter space. We can see that the large scale structure probes shown exhibit a significant degeneracy between σ_8 and Ω_m . That degeneracy is broken when adding BAO information as we show in the right panel of Fig. 23. A more extensive set of comparisons for S_8 and σ_8 is shown in Figs. 24 and 25 respectively. Generically, we include BAO information in all comparisons of σ_8 to break the $\Omega_m - \sigma_8$ degeneracy. An exception are those measurements employing three dimensional galaxy clustering either on its own, as part of a '3x2pt' analysis, or in cross-correlation. Those measurements already implicitly contain the BAO information.

We note that we do not reanalyse these datasets with our prior choices. An exception are the results from HSC for which the Markov Chain Monte Carlo runs are not publicly available at the time of writing and which were kindly provided by the HSC team with prior choices that match ours. Madhavacheril et al. (2023) discussed the impact of reanalysing the galaxy survey datasets with priors matching those in the ACT DR6 lensing autospectrum analysis and found only minor changes (see Appendix C in Madhavacheril et al. 2023). As previously discussed we also explore an alternative set of priors in Appendix G. We find that while our constraints on S_8 are not very sensitive to the choice of priors, the constraints on σ_8 in combination with BAO are slightly degraded (~10%) when the *Planck* prior on $\Omega_m h^3$ is removed. However, we do not observe any significant shifts in the mean σ_8 or S_8 and therefore do not expect our conclusions to differ substantially.

Our results are in good agreement with various CMB lensing auto-spectrum analyses. Given that CMB lensing is primarily sensitive to the parameter combination $\sigma_8 \Omega_m^{0.25}$ we combine our measurements as well as the CMB lensing measurements from *Planck*, SPTpol and ACT DR6 with BAO to be able to directly compare the resulting σ_8 constraints. The value of σ_8 inferred from the cross-correlation of unWISE with ACT DR6 and *Planck* together with BAO is consistent within 0.3 σ with the constraints from ACT DR6, within 0.04 σ with the result from *Planck* PR4, and within 0.06 σ with the results from the joint analysis of ACT DR6 and *Planck* PR4. In the parameter σ_8 the agreement with SPTpol is less good, though not at any statistical significance (1.2 σ).

These results can also be compared to galaxy weak lensing measurements of S_8 . Analyses of DES and KiDS obtain $S_8 = 0.782 \pm 0.019$ and $S_8 = 0.765^{+0.017}_{-0.016}$ respectively from a combination of cosmic shear and galaxy clustering. The reanalysis of HSC cosmic shear and galaxy clustering with our prior choices yields $S_8 =$ 0.782 ± 0.043 when using only linear scales and $S_8 =$ $0.730^{+0.025}_{-0.029}$ from an analysis using a halo model based emulator³¹. They are all lower than our results, but not in significant statistical tension (approximately 1.1, 2.0, 0.6 and 2.5 σ for DES, KiDS, and the two results from HSC respectively). We note that the DES and KiDS '3x2pt' analyses draw on cosmic shear measurements for which recent reanalyses have found shifts towards higher S_8 . A recent and improved reanalysis of KiDS-1000 cosmic shear data obtains $S_8 = 0.776^{+0.031}_{-0.030}$ (Li et al. 2023a) (increased from the previous $S_8 = 0.759^{+0.024}_{-0.021}$; Asgari et al. 2021), while a recent joint reanalysis of the DES and KiDS cosmic shear data also hints at a higher value of $S_8 = 0.790^{+0.018}_{-0.014}$ (also higher than previous DES results from cosmic shear alone which yielded $S_8 = 0.759^{+0.025}_{-0.023}$ and the KiDS results above; Abbott

et al. 2023a; Amon et al. 2022). For these two renewed cosmic shear analyses the discrepancy with our results is reduced to ~ 1.5 and ~ 0.9σ respectively. It is possible that the shear recalibration from Li et al. (2023a) and changes to the pipeline and prior choices in Abbott et al. (2023a) will also increase S_8 inferred from the '3x2pt' analyses to which we compare here.

In addition to evaluating the consistency of our cosmological parameters with various weak lensing surveys, we also test the goodness of fit when fixing the cosmology to one similar to the one preferred by DES. This is to test the possibility that our preference for a higher value of S_8 is driven in part by projection effects which have been seen (albeit in the opposite direction) in some other cross-correlation analyses (White et al. 2022; Chen et al. 2022b). We fixed S_8 to 0.78, close to the mean value obtained by DES (Abbott et al. 2022). We then vary Ω_m and all nuisance parameters in our analysis to obtain the best fit to the auto-correlations of the Blue and Green samples and their cross-correlations with the ACT lensing reconstruction. We find a minimum $\chi^2 = 22.8$. When instead fixing the cosmology to $S_8 = 0.83$, similar to the *Planck* cosmology, we similarly find $\chi^2 = 23.2$. Our best-fitting baseline model, in which S_8 and Ω_m are free to vary has a minimum χ^2 of 20.9 but includes one more degree of freedom suggesting that both the DES and *Planck* values of S_8 are compatible with our data at ~ 1σ . However, consistency with the DES cosmology comes at the cost of some tension with the matter density inferred from BAO. When we add BAO information to the minimisation at fixed S_8 we find $\Delta \chi^2 = 3.6$ for the DES-like cosmology relative to our baseline fit and $\Delta \chi^2 = 0.3$ for a *Planck*-like cosmology (including the χ^2 contributions from the BAO likelihoods). This corresponds to a cosmology with the DES-like value of S_8 being slightly disfavoured by $\sim 1.6\sigma$.

Our results appear to show a discrepancy with previous CMB lensing cross-correlations. This is true for the previous analysis of the cross-correlation of unWISE and *Planck* lensing (Krolewski et al. 2021); we discuss this comparison and the origin of the discrepancy in detail in Appendix I. We also find some discrepancy with the results from the cross-correlation of the DESI LRG targets with *Planck*. White et al. (2022) measured $S_8 = 0.73 \pm 0.03$ corresponding to about a 2.5 σ discrepancy with our analysis using both ACT and Planck. A similarly significant difference is also found with the cross-correlation of BOSS and *Planck*, which yields $S_8 =$ $0.707 \pm 0.037 \ (\sim 2.6\sigma)$ and with the cross-correlation of DES-Y3 with SPT and *Planck* $(S_8 = 0.736^{+0.032}_{-0.028};$ $\sim 2.2\sigma$). This discrepancy is reduced when adding the DES-internal cross- and auto-correlations. The '5x2pt'

³¹ These results differ by approximately 0.2 and 1.2σ respectively from the fiducial results reported by HSC (Sugiyama et al. 2023; Miyatake et al. 2023). The difference is largely driven by the fact that we fix $\Omega_m h^3$ to the mean value for this parameter combination found by *Planck*. In terms of the expected uncertainty on the difference between the two analyses the shifts correspond to about 0.3 - 0.6 and $1.2 - 1.5\sigma$ respectively (assuming that the fiducial analysis constitutes a subset of the analysis with our priors; Gratton & Challinor 2020). These shifts are thus within what is expected given the difference in prior choices.

analysis yields $S_8 = 0.773 \pm 0.016 \ (\sim 1.7\sigma)$. Adding the CMB lensing auto-correlation from *Planck* further increases S_8 to 0.792 ± 0.012 and reduces the discrepancy to $\sim 1.0\sigma$ ('6x2pt'). Given the consistency of our results with *Planck* lensing this is unsurprising. The analysis of DES-Y3 and ACT DR4 lensing also obtains a lower value of S_8 , albeit with much larger errors $(S_8 = 0.75^{+0.04}_{-0.05}; \sim 1.3\sigma)$.

Some of the improvements made here to the previous analysis of the cross-correlation between unWISE and *Planck* are also relevant for some of the other crosscorrelation measurements discussed above. In particular, the inclusion of the Monte Carlo correction to the normalisation of the *Planck* lensing maps (for details see Appendix I) will also impact the analyses in White et al. (2022); Chen et al. (2022b); Chang et al. (2023) and Abbott et al. (2023b). However, at least in some cases preliminary results show that they are unlikely to completely alleviate the discrepancy. White et al. (2022)and Chen et al. (2022b) also found indications for significant projection effects in their analyses. We also note the importance of the better redshift calibration of the unWISE samples which contributed a shift of $\sim 0.8\sigma$ to our reanalysis of the cross-correlation with *Planck* (see Appendix I) which may also affect other analyses and should be significantly improved in the near future with the availability of spectroscopic redshifts from DESI.

Our results are consistent with the eBOSS analysis of BAO and RSD which finds $\sigma_8 = 0.850 \pm 0.035$ (consistent with our baseline results within 1.0σ). The 'full shape' measurements on the other hand prefer a lower value of σ_8 by about 2σ .

10.3. Our results in the context of the S_8/σ_8 "tension"

We briefly discuss the implications of our results for the claimed S_8 tension. It is worth noting that most of the galaxy weak lensing constraints come at least partially from smaller scales, where gravitational nonlinearities, as well as the effects of galaxy formation (assembly bias, baryonic feedback, etc.) are more pronounced. Our cosmological constraints derive almost entirely from perturbative scales at the redshifts of interest $(k \lesssim 0.3 \ h/{\rm Mpc})$, and therefore we expect uncertainties due to galaxy formation to be negligible and fully absorbed in our bias expansion. In Fig. 26 we show the scales on which our signal originates. Extensive tests on mocks presented in this paper also show no evidence for "projection effects" in the Bayesian parameter estimation. Together with the large number of consistency and null tests, this provides confidence in the results.

The fact that CMB lensing power spectrum measurements (Qu et al. 2023; Madhavacheril et al. 2023; Carron



Figure 24. Here we compare with an extensive set of measurements of S_8 . We include measurements from CMB lensing analyses in combination with BAO (in green), from galaxy weak lensing surveys (in blue), from cross-correlations with CMB lensing (in red), from the three dimensional clustering for spectroscopic galaxy surveys (in gold), and predictions from the primary CMB (in magenta). Our results are in good agreement with CMB lensing analyses, as well as with the predictions from the primary CMB. They are also in no significant tension with any one galaxy weak lensing survey, or the S_8 inferred from the three dimensional clustering of spectroscopic galaxies. We do, however, find some discrepancies at up to 3σ with previous cross-correlation work.

et al. 2022; Planck Collaboration et al. 2020b) are consistent with the CMB power spectrum prediction deriving from high redshifts excludes the possibility of substantial effects of new physics on structure growth at high redshifts and large scales; however, the possibility of new physics at low redshifts ($z \leq 1$) or smaller scales remains when considering only CMB lensing power spectrum data. Although the analyses presented in this paper are sensitive to the growth of structure at $z \simeq 0.2 - 1.6$ (on large scales), our measurements show excellent consistency with CMB power spectrum predictions. In short, using the cross-correlations of CMB lensing and galaxy density we find no evidence for a S_8 (or σ_8) tension on scales $k \leq 0.3 h/Mpc$ and for $z \gtrsim 0.2$. Our results together with CMB lensing auto-spectrum analyses, im-


Figure 25. Here we compare with an extensive set of measurements of σ_8 . As above we include measurements from CMB lensing analyses in combination with BAO (in green), from galaxy weak lensing surveys (+BAO) (in blue), from cross-correlations with CMB lensing (+BAO) (in red), from the three dimensional clustering for spectroscopic galaxy surveys (in gold), and predictions from the primary CMB (in magenta). As in the case of S_8 , our results are in good agreement with most CMB lensing analyses, as well as with the predictions from the primary CMB. Galaxy weak lensing surveys favour smaller σ_8 at varying levels of significance $(0.8 - 3.2\sigma)$. We also find some discrepancies at up to 3σ with previous cross-correlation work.

ply that, if new physics is indeed responsible for the S_8 tension observed in galaxy weak lensing data sets, any effects of new physics must be mostly limited to small scales ($k \geq 0.3 \ h/Mpc$) to which our unWISE measurements are insensitive. This is consistent with resolutions to the S_8 "tension" based on modifications to the matter power spectrum on non-linear scales (Amon & Efstathiou 2022; Preston et al. 2023). Our work therefore motivates further efforts to test the Λ CDM predictions for structure formation on smaller, non-linear scales.

10.4. Conclusions

We have presented cosmological results from the crosscorrelation of the Blue and Green unWISE galaxy samples and the latest *Planck* and ACT CMB lensing maps. Thanks to its large footprint on the sky, extended redshift lever arm and high number density, the unWISE extragalactic samples are ideal for a number of crosscorrelations, including CMB lensing tomographic analyses.



Figure 26. We show the change in the total signal-to-noise as function of the maximum k considered. With our scale cuts ($50 \le \ell \le 400$ for $C_{\ell}^{\kappa g}$ and $100 \le \ell \le 400$ for C_{ℓ}^{gg}) the signal for the galaxy cross- and auto-correlations originates largely on scales $k \le 0.3 h/\text{Mpc}$. For the Green sample of galaxies and its cross-correlation with CMB lensing almost all the signal comes from k < 0.2 h/Mpc. While the Blue sample has minor contributions from $k \ge 0.3 h/\text{Mpc}$, most of the cosmological constraining power comes from larger scales since we marginalise over higher order contributions to the signal which become important on smaller scales.

Using conservative choices and a blinded analysis we were able to obtain a 1.7% and 1.8% measurement of S_8 and σ_8 , respectively, when combining the crosscorrelations of the Blue and Green unWISE samples with ACT and *Planck* lensing reconstructions. Both samples and their combination are in full agreement with CMB lensing and the primary CMB results, providing an important test of the standard cosmological model at $z \simeq 0.2 - 1.6$. Our results imply that, if new physics is responsible for the S_8 tension, any effects of new physics must be mostly limited to small scales ($k \gtrsim 0.3 h/Mpc$) to which our unWISE measurements are insensitive.

The statistical consistency with CMB lensing autocorrelation results, together with the comprehensive suite of systematic and null tests, motivates deriving joint constraints between the cross-correlation presented here and the CMB lensing auto-power spectrum. This analysis will be presented in upcoming work.

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Figure 27. Maximum off-diagonal correlations between different samples and between cross-correlations using the ACT and *Planck* lensing reconstructions. The most significant off-diagonal correlations are present between the $C_{\ell}^{\kappa g}$ from different samples which is due to the use of identical lensing maps and between C_{ℓ}^{gg} measured on the ACT and *Planck* footprints due to the use of the identical galaxy samples.

APPENDIX

A. OFF-DIAGONAL CORRELATIONS IN THE COVARIANCE MATRIX

Further to the discussion of the off-diagonal correlations in Sec. 6 we show here a summary of the level of correlations between different samples and between the cross-correlations using ACT and *Planck* lensing reconstructions. Fig. 27 shows the maximum off-diagonal correlations within our analysis range that uses bandpowers in the range $50 \le \ell \le 400$ in $C_{\ell}^{\kappa g}$ and $100 \le \ell \le 400$ in C_{ℓ}^{gg} .

B. DETAILED SUMMARY OF NULL-TESTS

A detailed summary of all our lensing reconstruction null-test results can be found in Table 4. We also show the corresponding bandpowers not already shown in the body of this paper in Figures 28, 29, and 31. The tests targeting the spatial homogeneity of the data are summarised in Table 5 with bandpowers shown in Fig. 32 and additionally Fig. 33.

We perform twelve null-tests using different lensing reconstructions for each of the two galaxy samples, Blue and Green, and additionally a test on the cross-correlation between our galaxy samples and the curl mode of the baseline lensing reconstruction. The latter is primarily a test of our covariance estimation since we do not expect any signal in the curl mode of the lensing reconstruction at current levels of precision.

In Sec. 7.1 we discuss the observed failures and summarise our null-tests targeting contamination of the lensing reconstruction. The number of observed failures is consistent with what would be expected due to random fluctuations given the correlations between the various tests. We nevertheless investigated whether the observed failures were indicative of contamination, for example by considering alternative masks to maximise or minimise the contribution from potential contaminants. We performed the comparison of the minimum variance and polarisation reconstructions on a more restrictive Galactic mask (see Fig. 30); if the failures observed on the full footprint was due to contamination by polarised Galactic dust, we should have seen an improvement further away from the Galactic plane and we do not.



Figure 28. At current levels of precision we do not expect any signal from the curl mode of the lensing reconstruction. The cross-correlation of the unWISE samples with the curl mode of the lensing reconstruction thus serves as a test of our covariance estimation.

Alternatively, the effect should be more significant near the Galactic plane (i.e. in the region which is included in our baseline mask but not the more restrictive Galactic mask), but this is not the case (see Fig. 29 panels (d) and (f)).

Our tests for the homogeneity of the data sets were discussed in Sec. 7.2. As pointed out there we do not expect null-tests to pass due to fluctuations in the selection properties of the galaxy sample on large scales. Those fluctuations affect, among other things, the galaxy bias. We thus construct an approximately bias independent quantity, $(C_{\ell}^{\kappa g})^2/C_{\ell}^{gg}$, on which we perform null-tests to confirm the homogeneity of the data. In Fig. 32 (panel (e)) we show this null-test for the difference between our baseline analysis and an analysis restricted to ecliptic latitude larger than 30°. This test primarily targets trends with unWISE depth which varies with ecliptic latitude due to the satellites scan strategy. We also compared the bandpowers obtained on the SGC and NGC (panel (f)) and find those to be consistent as well. Finally, in Fig. 33, we show tests for $C_{\ell}^{\kappa g}$ and C_{ℓ}^{gg} comparing different stellar masks. For this test we mask 0.1° around an additional 1.85 million stars with W2, W2 or 2MASS K_m brighter than 8th magnitude, beyond the 12890 very bright stars (W2 < 2.5) that we mask by default. This decreases the unmasked sky fraction in the unWISE mask from 0.58 to 0.45. The tests are passing and because we are changing the mask on small scales rather than selecting a different region we expect this test to be unaffected by large scale fluctuations in the galaxy selection.

C. DETAILS OF POWER SPECTRUM MODEL

In this appendix we discuss the details of our correction for the dependence of the cross-correlation redshifts on the fiducial cosmology, as well as our correction for the noise bias introduced by marginalisation over redshift uncertainties.

In Sec. 8.2 we raised the need to assume a fiducial cosmology to interpret the cross-correlation between the photometric and spectroscopic samples as estimates of b(z)dN/dz. In this appendix we lay out a formalism for correcting for this assumption to avoid biasing our results (Sec. C.1).

In Sec. 8.2 we discussed the effect of imposing smoothness and positivity on the normalised cross-correlation redshifts, $W^{\rm xc}(z)$. The result is a noise bias introduced by our marginalisation procedure, since the final C_{ℓ} effectively become a non-linear function of a noisy quantity. This bias can be as large as ~ 5% for some of the terms contributing to C_{ℓ}^{gg} in some range of ℓ s. The effect is less significant for $C_{\ell}^{\kappa g}$, but still remains comparable to the data covariance. In Sec. C.2 we detail how we subtract this noise bias.

C.1. Correction for the Fiducial Cosmology Assumed in Determining Cross-correlation Redshifts

As described in Sec. 3.1.2 (with further details provided in Krolewski et al. (2019) and Appendix K) we measure the redshift distribution of unWISE galaxies using cross-correlations with spectroscopic surveys. This introduces a dependence on the fiducial cosmology assumed in the process of converting from the measured cross-correlation functions to an estimate of b(z)dN/dz (Krolewski et al. 2021). To correct for this dependence we introduce a new method



(e) difference of cross-correlation bandpowers using temperature only and polarisation only reconstructions

(f) same as (e) but on the difference region of the 60% and 40% masks

Figure 29. Summary of additional null-tests related to potential foreground systematics in the lensing reconstruction. We compare cross-correlations with lensing reconstructions using only one of the two ACT frequencies (panels (a) and (b)), cross-correlations using the minimum variance and polarisation only reconstructions (panels (c) and (d)), and cross-correlations using the temperature only and polarisation only reconstructions (panels (e) and (f)). The number of failures is consistent with expectations when taking into account the correlations between different tests.

	Blue		Gre	en
Null-test	χ^2	PTE	χ^2	PTE
Curl	11.06	0.136	13.64	0.058
Minimum Variance - Temperature Only - Polarisation Only				
$C_{\ell}^{\kappa_{ m MV}g} - C_{\ell}^{\kappa_{ m TT}g}$	15.20	0.033	8.52	0.289
$C_{\ell}^{\kappa_{\mathrm{MV}}g} - C_{\ell}^{\kappa_{\mathrm{MVPOL}}g}$	13.62	0.058	10.44	0.165
$C_{\ell}^{\kappa_{\mathrm{TT}}g} - C_{\ell}^{\kappa_{\mathrm{MVPOL}}g}$	14.97	0.036	9.67	0.208
$C_{\ell}^{\kappa_{\rm MV}g} - C_{\ell}^{\kappa_{\rm MVPOL}g}$ (40% gal. mask)	15.18	0.034	5.63	0.584
Minimum Variance - Temperature Only - Polarisation Only (near Galactic plane)				
$C_{\ell}^{\kappa_{\rm MV}g} - C_{\ell}^{\kappa_{\rm TT}g}$ (near Galactic plane)	7.40	0.388	6.76	0.454
$C_{\ell}^{\kappa_{\rm MV}g} - C_{\ell}^{\kappa_{\rm MVPOL}g}$ (near Galactic plane)	4.11	0.767	8.17	0.318
$C_{\ell}^{\kappa_{\mathrm{TT}}g} - C_{\ell}^{\kappa_{\mathrm{MVPOL}}g}$ (near Galactic plane)	4.98	0.663	6.37	0.497
Frequency difference tests				
$C_{\ell}^{\kappa_{\rm MV,noise only}g}$ (150GHz - 90GHz map level null-test, minimum variance)	6.70	0.460	10.89	0.143
$C_{\ell}^{\kappa_{\rm TT,noise only}g}$ (150GHz - 90GHz map level null-test, temperature only)	9.45	0.222	7.33	0.396
$C_{\ell}^{\kappa_{\rm MV,150GHz}g} - C_{\ell}^{\kappa_{\rm MV,90GHz}g}$ (150GHz - 90GHz bandpower null-test, minimum variance)	3.36	0.850	6.78	0.452
$C_{\ell}^{\kappa_{\text{TT},150\text{GHz}}g} - C_{\ell}^{\kappa_{\text{TT},90\text{GHz}}g}$ (150GHz - 90GHz bandpower null-test, temperature only)	2.35	0.938	10.79	0.148
Other				
baseline vs. CIB deprojection	16.09	0.024	10.21	0.177

Table 4. Summary of null-tests for $C_{\ell}^{\kappa g}$. We compute the χ^2 within our cosmological analysis range of $50 \leq \ell \leq 400$ and estimate the PTE using a χ^2 -distribution with 7 degrees of freedom (given the seven bandpowers in the analysis range). We find 4 failures (PTE < 0.05) which are highly correlated; all relate to comparing the cross-correlation of the Blue sample of galaxies with different lensing reconstruction options (minimum variance, temperature only, and polarisation only). We investigate in Sec. 7.1 potential systematic contamination, but stress that the number of failures is not inconsistent with random fluctuations and no further evidence of contamination is found.

	Bhie						Green					
	С	ıkg l	C_{i}	9 <i>9</i>	$(C_{\ell}^{\kappa g})$	C_{ℓ}^{gg}	C'_{ℓ}	κg	C_{i}	99 l	$(C_{\ell}^{\kappa g})$	C_{ℓ}^{gg}
Null-test	χ^2	PTE	χ^2	PTE	χ^2	PTE	χ^2	PTE	χ^2	PTE	χ^2	PTE
40% Galactic mask	4.99	0.662	6.78	0.341	2.57	0.861	4.65	0.702	16.51	0.011	3.39	0.758
$\beta_{\rm ecliptic} > 30^{\circ}$	7.43	0.386	30.19	0.000	5.63	0.466	10.39	0.168	45.79	0.000	2.65	0.851
SGC - NGC	1.94	0.963	2.66	0.850	2.18	0.903	2.03	0.958	4.59	0.597	4.04	0.672
stricter stellar mask	4.83	0.681	7.65	0.265	_	_	4.72	0.694	4.50	0.609	_	_

Table 5. Summary of null-tests targeting the spatial homogeneity of the galaxy samples. As expected we find some inhomogeneity in the galaxy sample, in particular with respect to cuts in ecliptic latitude. The χ^2 is computed within our cosmological analysis ranges of $50 \le \ell \le 400$ for $C_{\ell}^{\kappa g}$ and $100 \le \ell \le 400$ for $C_{\ell}^{\kappa g}$. We estimate the PTE using a χ^2 -distribution with 7(6) degrees of freedom (given the 7(6) bandpowers in the analysis range for $C_{\ell}^{\kappa g}$ (C_{ℓ}^{gg})). As argued in Sec. 7.2, the failures are likely due to large scale inhomogeneities in the galaxy selection and thus varying galaxy bias. The bias independent combination $(C_{\ell}^{\kappa g})^2/C_{\ell}^{gg}$, however, passes our null-tests.



Figure 30. Null-test showing the difference of cross-correlation bandpowers using minimum variance and polarisation only reconstructions on the more restrictive (40%) Galactic mask. We observe a failure for the Blue sample which is highly correlated with other observed failures shown above. Overall the number of failures is consistent with expectations taking into account the correlations between our tests.



Figure 31. Null-test showing the difference of cross-correlation bandpowers using our baseline reconstruction and one which explicitly deprojects CIB contamination using higher frequency data from *Planck*. Again we see a failure for the Blue sample. We note however that we cannot rule out the possibility that this is related to a slightly different mask used for the CIB deprojected reconstruction which removes some additional areas near the Galactic plane. Looking at the null-test for the galaxy auto-correlation on the modified masks also yields a borderline PTE of 0.09.

which is exact in the limit of narrow redshift bins used to measure the cross-correlations between the photometric and spectroscopic samples.

We measure the cross-correlations between the photometric unWISE sample (subsequently indicated by a subscript p) and multiple spectroscopic tracers (subsequently denoted with a subscript s) as well as the spectroscopic sample's auto-correlation in N narrow redshift bins between $z_{\min,i}$ and $z_{\max,i}$ (i = 1, ..., N). Assuming scale independent bias³³

³³ This assumption is tested to some extent in our mocks used for model verification, where we have also measured the crosscorrelation redshifts on the same scales as in our analysis. The ability to recover an unbiased cosmology discussed in Sec. 8.3 indicates that this does not bias our constraints. Moreover, we find similar cosmological parameters (shift of -0.1σ in S_8) when increasing the minimum scale used in the cross-correlation redshifts, which decreases the sensitivity to scale-dependent bias.



(a) Difference of $C_{\ell}^{\kappa g}$ measured in the baseline footprint and when restricted to ecliptic latitudes > 30°.



(c) Difference of $C_{\ell}^{\kappa g}$ measured in the southern and northern galactic cap (SGC and NGC)















Figure 32. Summary of additional null-tests related to the spatial homogeneity of the observed spectra. We compare crosscorrelations and galaxy auto-correlation spectra for different ecpliptic latitudes (panels (a) and (b)) and between the SGC and NGC (panels (c) and (d)). In addition, we show the approximately bias independent quantity $(C_{\ell}^{\kappa g})^2/C_{\ell}^{gg}$ for these two comparisons (panels (e) and (f)). While we find several highly significant failures for C_{ℓ}^{gg} caused by variations in the galaxy bias arising from large scale fluctuations in the galaxy selection properties, the bias independent quantity $(C_{\ell}^{\kappa g})^2/C_{\ell}^{gg}$ passes our tests.



Figure 33. These null-tests compare our baseline mask, which removes any source within 2.75'' from a Gaia point source (with W2 < 2.5), with one which additionally masks any source within 0.1° from a further 1.85 million stars brighter than 8th magnitude. We find no evidence for problems with stellar contamination. This is consistent with results from alternative estimates of the stellar contamination in Krolewski et al. (2019).

the relevant correlation functions are given in theory by

$$w_{sp}(\theta, z_i) = \int \frac{kdk}{2\pi} \int_{z_{\min,i}}^{z_{\max,i}} dz J_0 \left[k\theta\chi(z) \right] b_s \frac{dN_s}{dz} b_p \frac{dN_p}{dz} H(z) P_{gg}(k, z) \tag{C1}$$

$$w_{ss}(\theta, z_i) = \int \frac{kdk}{2\pi} \int_{z_{\min,i}}^{z_{\max,i}} dz J_0 \left[k\theta\chi(z) \right] \left(b_s \frac{dN_s}{dz} \right)^2 H(z) P_{gg}(k, z).$$
(C2)

Here b_s and b_p are the bias of the spectroscopic and photometric samples respectively and dN_s/dz and dN_p/dz are the corresponding redshift distributions. The comoving radial distance, $\chi(z)$, the Hubble rate H(z) and the galaxy-galaxy power spectrum $P_{qq}(k, z)$ depend on the assumed cosmology.

Furthermore, the correlation functions are binned into n = 3 log-spaced angular bins between $\theta_{\min} = s_{\min}/\chi_{\text{fid}}(z_{\text{central}})$ and $\theta_{\max} = s_{\max}/\chi_{\text{fid}}(z_{\text{central}})$ where $s_{\min} = 2.5 \text{ h}^{-1}\text{Mpc}$ and $s_{\max} = 10 \text{ h}^{-1}\text{Mpc}$. Here χ_{fid} denotes the comoving radial distance within the chosen fiducial cosmology. Finally, the resulting quantity is integrated over the full range of scales weighting by θ^{-1} to increase the signal to noise (Ménard et al. 2013), yielding

$$\bar{w}_{XY}(z_i) = \sum_{j=1}^n \frac{2\Delta\theta_j}{\theta_j(\theta_{\max,j}^2 - \theta_{\min,j}^2)} \int_{\theta_{\min,j}}^{\theta_{\max,j}} \theta d\theta \ w_{XY}(\theta, z_i).$$
(C3)

In the limit of narrow redshift bins we can estimate the spectroscopic bias, b_s , from the measured binned and integrated auto-correlation function as

$$\hat{b}_s(z_i) = \sqrt{\hat{w}_{ss}(z_i)} \left[\Delta z_i \left(\frac{dN_s}{dz} \right)^2 H(z_i) \int k dk P_{gg}(k, z_i) W(k, z_i) \right]^{-1/2}, \tag{C4}$$

where W(k, z) is a weighting function which can be computed analytically as

$$W(k,z) = \sum_{j=1}^{n} \frac{\Delta \theta_j \left[\theta_{\max,j} J_1\left(k\chi(z)\theta_{\max,j}\right) - \theta_{\min,j} J_1\left(k\chi(z)\theta_{\min,j}\right)\right]}{k\chi(z)\theta_j \pi \left(\theta_{\max,j}^2 - \theta_{\min,j}^2\right)}.$$
 (C5)

Finally, we can obtain an estimate of $b_p dN_p/dz$ as

$$\widehat{b_p \frac{dN_p}{dz}} = \frac{\widehat{w}_{sp}(z_i)}{\sqrt{\widehat{w}_{ss}(z_i)}} \left[\Delta z_i H(z_i) \int k dk P_{gg}(k, z_i) W(k, z_i) \right]^{-1/2}.$$
(C6)

Given $\left(b_p dN_p/dz\right)_{\text{fid}}$ measured in a chosen fiducial cosmology the dependence on the choice of fiducial cosmology can therefore be corrected as

$$\widehat{b_p \frac{dN_p}{dz}} = \left(\widehat{b_p \frac{dN_p}{dz}}\right)_{\text{fid}} \frac{\mathcal{C}}{\mathcal{C}_{\text{fid}}}.$$
(C7)

with

$$\mathcal{C} = \left[\Delta z_i H(z_i) \int k dk P_{gg}(k, z_i) W(k, z_i)\right]^{-1/2}.$$
(C8)

In our analysis we compute C for every new cosmology to avoid biasing our results through the dependence on the assumed fiducial cosmology. For consistency with the method used to measure the cross-correlation redshifts we use a HMCode model for $P_{gg}(k, z)$. We show in Table 6 that our S_8 constraints are unchanged even when using a different fiducial cosmology, validating the correction of Eq. C.1.

C.2. Noise Bias Correction

Similarly to Krolewski et al. (2021) we compute a correction for the noise bias introduced by marginalisation over uncertainties in the redshift distributions within a fiducial cosmology (again taken to be the mean cosmology from Planck Collaboration et al. 2020a). First we draw 9000 sets of PCA coefficients for each of the galaxy samples according to the skewnormal priors obtained as described in Sec. 8.2 and summarised in Table 2. This number is chosen such that the resulting uncertainty in the correction is highly subdominant compared to the data covariance. We then evaluate the power spectrum model term by term for each set of coefficients and obtain the noise bias as

$$(\Delta C_{\ell}^{XY})_{\text{noise bias}} = \langle C_{\ell}^{XY,\text{theory}} \rangle_{c_1,\dots,c_n} - C_{\ell}^{XY,\text{theory}} \Big|_{c_1 = \dots = c_n = 0}$$
(C9)

where $C_{\ell}^{XY,\text{theory}}$ are the various terms contributing to our theory expressions for C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$. During cosmology analysis we then correct the theory spectra as

$$(C_{\ell}^{XY,\text{theory}})_{\text{corrected}} = C_{\ell}^{XY,\text{theory}} - (\Delta C_{\ell}^{XY})_{\text{noise bias}}.$$
(C10)

In this process we scale our noise bias correction appropriately with the value of the nuisance parameters including galaxy bias parameters and lensing magnification parameters.

D. POSTERIOR DISTRIBUTIONS FOR THE FULL PARAMETER SPACE

In Figs. 34 and 35 we show the posterior distributions for the full parameter space including the galaxy nuisance parameters for the Blue and Green sample of unWISE galaxies respectively. The posteriors shown are for our baseline analysis without BAO. Alongside the posterior distributions we also show the maximum a posterior (MAP) and the priors on all sampled parameters. We observe that the data does not constrain the redshift distribution marginalisation parameters beyond the information provided by the priors. These parameters are also only very weakly degenerate with the cosmological parameters of interest. In the case of the free parameters modifying the higher order bias evolution we see that the only parameter which is constrained by the data is the offset in the second order Lagrangian bias for the Blue sample. This parameter has some degeneracy with cosmology (see Appendix E for some discussion of this). For the Green sample higher order contributions are less significant and the constraints on the higher order bias parameters are prior dominated.

E. PARAMETER CONSISTENCY FROM BLUE AND GREEN

In Sec. 9.5 we presented posteriors on cosmological parameters obtained from the analysis of the auto-correlation of the two unWISE redshift samples, Blue and Green independently, and their cross-correlation with CMB lensing reconstructions from ACT. Our analysis is primarily sensitive to the amplitude of the signal which depends on the parameter combination $S_8^{\times} \equiv \sigma_8 (\Omega_m/0.3)^{0.45}$. When using only a single tomographic bin (and neglecting structure growth within the bin) σ_8 and Ω_m are exactly degenerate at linear order. Constraints on either parameter individually therefore arise only from beyond linear order contributions to the spectrum. Therefore, the inference of σ_8 and Ω_m



Figure 34. Posterior distributions in the full parameter space including the galaxy nuisance parameters for the Blue sample of unWISE galaxies. The dashed grey lines indicate the maximum a posterior (MAP) and the orange dashed lines show the priors on all samples parameters. We observe that the data does not constrain the redshift distribution beyond the priors provided which are described in Sec. 8.2. On the other hand the data is informative for the second order Lagrangian bias.

is very sensitively dependent on the priors on higher order parameters. In Fig. 36 we can see that σ_8 and Ω_m show a strong degeneracy with $c_{2,L}^{\text{Blue}}$, the free parameter offsetting the second order Lagrangian bias for the Blue sample. S_8 on the other hand shows a much weaker degeneracy with the higher order bias parameters. The degeneracy is broken when both tomographic samples are included or when we additionally include BAO information.

Given the sensitivity to the higher order bias priors we thus do not necessarily expect the two samples to yield the same values for σ_8 or Ω_m and the fact that they are consistent at 1.7σ suggest that our choice of priors is reasonable.



Figure 35. Same as Fig. 34 for the Green sample of unWISE galaxies. As with the Blue sample the redshift distribution parameters are not constrained by the data. In contrast with Blue the higher order biases are also prior dominated since these contributions are significantly smaller for the Green sample.

Furthermore, our tests described in Sec. 7.2 specifically ensured that the bias-independent amplitude of our spectra remained consistent across different subsets of the data. We can thus not rule out that the exact scale dependence of the spectrum which gives rise to the weak constraints on σ_8 and Ω_m in this type of analysis is affected to some degree by residual systematic contamination.

In the parameter which is sensitive to the amplitude of the spectrum, S_8 (or S_8^{\times}), on the other hand, we find good agreement between the samples and thus consider it justified to combined them.



Figure 36. When plotting the posterior for the second order Lagrangian bias parameter, $c_{2,L}^{\text{Blue}}$, along with the cosmological parameters we can see that the σ_8 and Ω_m are highly degenerate with this parameter when using only the combination of C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ (left). This is in contrast to S_8 which is only weakly dependent on the higher order bias. With BAO data (right) this degeneracy is broken and the inferred parameters are less sensitive to the higher order bias priors.

F. SUMMARY OF ALL PARAMETER CONSISTENCY TESTS

In Table 6 we show a summary of several parameter consistency tests performed in this work; these employ various subsets of our baseline dataset. These are described in Sec. 9.7. We also summarise more consistency tests for alternative analysis choices in Table 7.

G. ALTERNATIVE PRIORS

In this section we explore the impact of an alternative set of priors on cosmological parameters. In particular we adopt a set of priors compatible with Qu et al. (2023). The main difference between the priors adopted in this work (see Table 2) and those from Qu et al. (2023) is the fact that H_0 is left to vary with a uniform prior between $40 \text{ km s}^{-1}\text{Mpc}^{-1}$ and $100 \text{ km s}^{-1}\text{Mpc}^{-1}$. Additionally, Qu et al. (2023) adopt priors on n_s and $\Omega_b h^2$ rather than fixing those parameters to the mean values from *Planck* (Planck Collaboration et al. 2020a). We summarise this alternative set of priors in Table 8.

We find $S_8 = 0.810 \pm 0.021$ from the combination of the galaxy auto-correlation and the cross-correlation measurements using both samples of unWISE galaxies, only a small degradation of constraining power compared to our baseline analysis (see Sec. 9.5). Adding BAO information we obtain $\sigma_8 = 0.812 \pm 0.023$ which is a more significant degradation of constraining power compared to our baseline analysis with BAO. This is explained by the fact that BAO data provide constraints primarily in the $\Omega_m - H_0$ plane. In our baseline analysis these two parameter do not vary independently so that the addition of BAO information directly represents a constraint on Ω_m which breaks the degeneracy between σ_8 and Ω_m more effectively than in the case where H_0 is allowed to vary independently.

As we argued in Sec. 9.3 the angular size of the sound horizon, $\theta_{\rm MC}$, which within a ACDM cosmology determines the combination $\Omega_m h^3$, is measured with great precision by *Planck* (Planck Collaboration et al. 2020a). In the context of aiming to compare the amplitude of low redshift structure to expectations from fits to the primary CMB it is reasonable to adopt this constraint that helps break the degeneracy between distances and structure growth. Since

		ΔS_8	ΔS_8^{\times}	PTE
	Blue	$-0.035(^{+0.027}_{-0.043})^a$	$-0.031(^{+0.025}_{-0.039})^a$	_
TT-only	Green	0.012 ± 0.010	0.010 ± 0.009	0.24
	Blue+Green	-0.014 ± 0.010	-0.011 ± 0.009	0.22
	Blue	$0.008(^{+0.033}_{-0.040})^{a}$	$0.019(^{+0.031}_{-0.038})^{a}$	_
$\beta_{\rm ecliptic} > 30^{\circ}$	Green	-0.0003 ± 0.025	-0.0004 ± 0.021	0.99
	Blue+Green	-0.005 ± 0.020	-0.003 ± 0.017	0.86
	Blue	$-0.0001(^{+0.027}_{-0.034})^a$	$0.006(^{+0.024}_{-0.031})^{a}$	-
40% Galactic mask	Green	-0.001 ± 0.012	0.002 ± 0.011	0.85
	Blue+Green	-0.002 ± 0.008	0.001 ± 0.008	0.87
	Blue	$0.019(^{+0.027}_{-0.034})^{a}$	$0.019(^{+0.027}_{-0.033})^a$	-
$\ell_{\min}^{gg} = 150, \ell_{\min}^{\kappa g} = 100$	Green	0.006 ± 0.015	0.005 ± 0.011	0.64
	Blue+Green	0.006 ± 0.011	0.005 ± 0.009	0.55
	Blue	$-0.002(^{+0.028}_{-0.041})^a$	$-0.003(^{+0.025}_{-0.036})^{a}$	_
$\ell_{\rm max} = 300$	Green	-0.003 ± 0.014	-0.003 ± 0.012	0.80
	Blue+Green	-0.001 ± 0.010	-0.002 ± 0.010	0.88
	Blue	$-0.033(^{+0.027}_{-0.077})^a$	$-0.031(^{+0.023}_{-0.070})^a$	_
Higher order bias priors $\times 2$	Green	-0.009 ± 0.020	-0.009 ± 0.019	0.63
	Blue+Green	-0.010 ± 0.016	-0.010 ± 0.015	0.47
	Blue	$0.002(^{+0.025}_{-0.039})^a$	$0.002(^{+0.022}_{-0.034})^a$	-
More restrictive $b(z)dN/dz$ scale cut	Green	-0.004 ± 0.005	-0.004 ± 0.005	0.46
	Blue+Green	-0.002 ± 0.002	-0.002 ± 0.003	0.52
	Blue	-0.029 ± 0.063	-0.039 ± 0.059	0.48
NGC - SGC	Green	0.101 ± 0.059	0.088 ± 0.052	0.09
	Blue+Green	0.032 ± 0.040	0.032 ± 0.044	0.46

^aThe posteriors on S_8 and S_8^{\times} for analyses using only the Blue sample are not well approximated by Gaussians. Therefore, we were unable to reliably estimate the significance of the shift for these consistency tests. We nevertheless show here the shift in the parameter mean, which is in most cases small compared to the measurement errors shown in parentheses.

Table 6. We perform a series of parameter level consistency tests detailed here. These tests were originally performed blind using random offsets on the cosmological parameters. We show the difference in the inferred S_8 and S_8^{\times} for these analyses compared to our baseline analysis (and the difference between analyses using only the NGC and SGC respectively). Note that the uncertainty quoted is the uncertainty on the difference between the baseline analysis and the subset analysis estimated under the assumption that the posteriors are approximately Gaussian following the prescription from Gratton & Challinor (2020). In the case of the comparison of analyses performed on the NGC and SGC, the two analyses are fully independent and we estimate the consistency between the two data sets taking into account the full non-Gaussian posterior using the TENSIOMETER package. We estimate the statistical likelihood of the shifts obtained assuming that we are effectively probing a single parameter, S_8^{\times} . We find all consistency tests to yield shifts consistent with expectations based on random fluctuations within $< 2\sigma$.

we are here primarily interested in investigating the claimed ' S_8 tension' we adopt this approach as our baseline, but will investigate other options that allow one to probe alternative cosmologies in future work. We also note that the comparison with other works that constrain S_8 without the addition of BAO is still meaningful as we showed above that our S_8 constraints are not significantly affected by this choice of priors.

		S_8	ΔS_8	$S_8^{ imes}$	ΔS_8^{\times}
	Blue	$0.833^{+0.026}_{-0.034}$	0.016	$0.807^{+0.024}_{-0.031}$	0.020
CIB deprojection	Green	0.789 ± 0.027	0.014	0.794 ± 0.024	-0.015
	Blue+Green	0.800 ± 0.021	0.013	0.803 ± 0.019	0.013
	Blue	$0.886\substack{+0.030\\-0.070}$	-0.037	$0.860\substack{+0.028\\-0.064}$	-0.033
Higher order biases centred on co-evolution	Green	$0.812\substack{+0.026\\-0.032}$	-0.009	$0.818\substack{+0.024\\-0.029}$	-0.008
	Blue+Green	$0.826\substack{+0.022\\-0.026}$	-0.013	$0.829^{+0.020}_{-0.024}$	-0.012
	Blue	$0.842^{+0.024}_{-0.038}$	0.007	$0.819\substack{+0.021\\-0.034}$	0.008
Different fid. cosmo. for $b(z)dN/dz$	Green	0.801 ± 0.026	0.002	0.807 ± 0.024	0.002
	Blue+Green	0.809 ± 0.020	0.004	0.812 ± 0.019	0.004
	Blue	$0.847^{+0.025}_{-0.037}$	0.002	$0.825^{+0.022}_{-0.033}$	0.002
Independent set of simulations	Green	$0.803\substack{+0.025\\-0.028}$	0.0004	$0.809\substack{+0.022\\-0.025}$	0.0003
	Blue+Green	0.813 ± 0.021	0.0001	0.816 ± 0.019	0.0001
	Blue	0.821 ± 0.024	0.028	0.806 ± 0.023	0.021
HMCode-only model	Green	0.802 ± 0.025	0.001	0.808 ± 0.023	0.002
	Blue+Green	0.812 ± 0.019	0.002	0.814 ± 0.018	0.002
	Blue	0.869 ± 0.025	-0.02	0.842 ± 0.023	-0.015
Fixed fid. higher order biases	Green	0.807 ± 0.025	-0.004	0.812 ± 0.022	-0.003
	Blue+Green	0.826 ± 0.021	-0.013	0.829 ± 0.018	-0.012
	Blue	$0.856^{+0.019}_{-0.047}$	-0.007	$0.833^{+0.017}_{-0.042}$	-0.006
No mag. bias	Green	$0.815_{-0.029}^{+0.026}$	-0.012	$0.823^{+0.024}_{-0.027}$	-0.013
	Blue+Green	0.817 ± 0.021	-0.004	0.823 ± 0.019	-0.007

Table 7. Here we show further the parameter consistency tests not involving a subset of our baseline analysis. For our consistency test using a lensing reconstruction with explicit CIB deprojection, alternative analysis choices in regards to the higher order bias priors, and the fiducial cosmology assumed for the cross-correlation redshifts it is less straightforward to estimate the statistical significance of the observed parameter shifts. However, as can be seen here the impact of these alternative analysis choices is small compared to our uncertainties. We also show an analysis run with a covariance obtained from a independent set of 400 simulations. The stability of the results indicates that our covariance is sufficiently converged. Furthermore, we summarised analyses using simplified models, an analysis using only the HMCode matter power spectrum, an analysis fixing all higher order biases to the fiducial coevolution relations, and an analysis neglecting all magnification bias contributions. While these analyses are not necessarily expected to be consistent with out baseline results, they provide an indication of the importance of different model components. We find that the shifts induced by these simplified models are small compared to our uncertainties, but they also do not yield dramatic improvements in constraining power.

H. CONSISTENCY TEST OF GALAXY-GALAXY CROSS-SPECTRUM

As a further consistency test we compare the prediction for the cross-correlation between the Blue and Green sample of unWISE galaxies obtained from the model inference we presented in Sec. 9.5 with the observed cross-spectrum. This cross-spectrum contains potentially valuable information to constrain the magnification bias (particularly for the Green sample where the magnification kernel entirely overlaps with the Blue sample) and the high redshift tail of the Blue sample. Nevertheless, we chose to neglect this information in the model fit since the ratio of potential systematic contamination to signal is less favourable as compared to the auto-spectrum. While the signal is approximately a factor 1.5-2.5 smaller than the auto-correlation of the galaxy samples, observational systematics are expected to be of the same order as in the auto-correlation since they are likely common between both samples. We never performed any null or consistency test on the cross-spectrum and hence do not use it for cosmology analysis.

Here we present a comparison of the model we constrain in Sec. 9.5 to the observed cross-spectrum as a consistency test (see Fig. 38). We find that the observed cross-spectrum is consistent with the model predictions within the joint data and model errors ($\chi^2 = 8.9$, PTE = 0.18). We sample 500 random points from posterior of the joint fit to the Green and Blue auto- and cross-correlations (including BAO information) to obtain the scatter on the model prediction

Parameter	Prior
$\ln(10^{10}A_s)$	[2, 4]
H_0	[40, 100]
n_s	$\mathcal{N}(0.96, 0.02)$
$\Omega_b h^2$	$\mathcal{N}(0.0223, 0.0005)$
$\Omega_c h^2$	[0.005, 0.99]

Table 8. The priors adopted from Qu et al. (2023). These are identical to those used in the lensing power spectrum analysis performed by the *Planck* team (Planck Collaboration et al. 2020b). For the galaxy nuisance parameters we adopt the same priors as in Table 2.



Figure 37. Parameter constraints from the cross-correlation of ACT DR6 lensing and unWISE galaxies using the priors from Qu et al. (2023) (see Table 8). The posteriors on Ω_m and σ_8 are significantly more non-Gaussian with this choice of priors compared to our baseline priors (Table 2). The best constrained parameters S_8 and S_8^{\times} are not strongly affected by this choice of priors, however. We find $S_8 = 0.810 \pm 0.021$ from the combination of C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ for both samples of unWISE galaxies (left). Adding BAO data (right) yields $\sigma_8 = 0.812 \pm 0.023$, a slightly more significant degradation of constraining power compared to our baseline.

(shown as grey lines in Fig. 38). To compute the χ^2 we add the covariance of these 500 model predictions (which has significant off-diagonal components) to an analytical approximation of the data covariance and compute the residual as the difference between the observed cross-spectrum and the best fit model prediction.

UNWISE X ACT DR6 COSMOLOGY

	Ω_m	σ_8	S_8	S_8^{\times}				
	ACT DR6 \times unWISE only							
Blue	$0.47\substack{+0.12\\-0.24}$	$0.72\substack{+0.10 \\ -0.19}$	$0.843^{+0.026}_{-0.033}$	$0.828^{+0.028}_{-0.037}$				
Green	$0.249_{-0.082}^{+0.044}$	$0.896^{+0.074}_{-0.12}$	$0.799^{+0.027}_{-0.031}$	0.807 ± 0.024				
Joint	$0.261\substack{+0.049\\-0.066}$	$0.882^{+0.069}_{-0.11}$	0.810 ± 0.021	0.817 ± 0.019				
		ACT DR6 \times u	nWISE + BAO					
Blue	0.363 ± 0.025	$0.768^{+0.029}_{-0.034}$	$0.843^{+0.026}_{-0.034}$	$0.835^{+0.025}_{-0.033}$				
Green	0.299 ± 0.015	0.819 ± 0.024	0.817 ± 0.023	0.817 ± 0.023				
Joint	0.305 ± 0.014	0.813 ± 0.022	0.819 ± 0.019	0.818 ± 0.019				

Table 9. The table summarises the constraints on cosmological parameters obtained from the cross-correlation of unWISE galaxies with ACT DR6 lensing reconstruction using the priors from Qu et al. (2023) (see Table 2). It should be compared to Table 3. The constraining power on the best constrained parameters S_8 and S_8^{\times} is similar to our baseline analysis, but the addition of BAO proves less powerful in constraining σ_8 with these alternative priors.



Figure 38. The cross-correlation between the Blue and Green sample of galaxies is consistent with our model prediction from a joint fit to the Blue and Green sample from Sec. 9.5. The grey lines show the spread in model predictions consistent with the posterior of the model fit. The best fit prediction is shown in orange, with the measured cross-spectrum shown in Blue. The grey bands indicate the range of scales not used for cosmological analysis.

I. REANALYSIS OF THE CROSS-CORRELATION OF PLANCK LENSING AND UNWISE

In this paper we present an analysis pipeline for the cross-correlation between CMB lensing reconstruction from ACT and unWISE galaxies. In the subsequent sections we discuss the differences to the previous analysis of the cross-correlation between *Planck* lensing and unWISE presented by Krolewski et al. (2021) (with spectra and redshift distributions measured in Krolewski et al. 2019). In Sec. I.3 we also show a reanalysis of the cross-correlation with *Planck* based on the newest available lensing reconstruction (Carron et al. 2022).

I.1. Summary of the Changes to the Analysis

In Sec. 8.1 we outlined the improvements we made to the model that was used for analysis of the cross-correlation of *Planck* PR3 lensing (Planck Collaboration et al. 2020b) with the unWISE galaxies presented in Krolewski et al. (2019). This includes marginalisation over higher order biases, improved marginalisation over redshift uncertainties, and a different treatment of the fiducial cosmology dependence of the cross-correlation redshift estimates. Additionally, we have also introduced systematics weighting for the unWISE samples (see Sec. 3.1.3) and we outline in Sec. 3.1.2 (with details provided in Appendix K) the additional data that are available to improve the cross-correlation redshift

	Impact on S_8
Monte Carlo lensing norm correction	$+0.6\sigma$
modelling improvements	-0.5σ
Systematics weighting	$+0.4\sigma$
Additional spectroscopic data	$+0.8\sigma$
use of <i>Planck</i> PR4 lensing reconstruction	$+0.2\sigma$
PCA based dN/dz marginalisation	$-0.2\sigma + \sim 15\%$ wider posteriors
fid. cosmo. correction	change in degeneracy directions
Total	$+1.3\sigma+{\sim}15\%$ wider posteriors

Table 10. We make various improvements to the previous analysis of the crosscorrelation between *Planck* lensing and unWISE galaxies. This includes the exploitation of updated and additional data such as the PR4 lensing reconstruction from *Planck* and additional spectroscopic data from eBOSS to constrain the sample redshifts. Additionally, we include systematics weighting for the unWISE galaxies, improvements to the modelling, and the Monte Carlo lensing norm correction for the *Planck* lensing map. Here we summarise the impact of these improvements.

estimates for the unWISE samples. New and improved data are also available on the *Planck* lensing side with the improved PR4 release (Carron et al. 2022) briefly described in Sec. 3.3. Finally, it was noticed that previous cross-correlation work had neglected to include a Monte-Carlo correction that is required to achieve the correct normalisation of the lensing reconstruction; we briefly discuss this aspect in more detail in the subsequent section.

The impact of the various changes and improvements is summarised in Table 10. The Monte Carlo lensing norm correction is described in Appendix I.2 below, and the modelling improvements are described in Section 8. The change in spectra from applying systematics weighting is shown in Fig. 45. The most significant driver of the change in S_8 is the shift towards higher $C_{\ell}^{\kappa g}$ for the Blue sample when applying the weights, while C_{ℓ}^{gg} is largely unchanged. In contrast, for the Green sample both $C_{\ell}^{\kappa g}$ and C_{ℓ}^{gg} increase, leading to a smaller change in S_8 . The change in redshift distribution is shown in Fig. 50; the impact on parameters is dominated by the Blue sample, where the redshift distribution changes much more (0.74σ) than for the Green sample (0.14σ) . When adding the new data to the cross-correlation redshifts, the Blue sample shifts to lower redshift. At fixed parameters, this increases C_{ℓ}^{gg} (due to the larger amplitude of structure at low redshift) while decreasing $C_{\ell}^{\kappa g}$ (due to the smaller overlap with the lensing kernel). To fit to the fixed data vector, the parameters must change in the opposite direction, leading to a higher S_8 . The Green sample shifts to slightly higher redshift, although the dominant change is likely from the narrower width of the new dN/dz, which increases C_{ℓ}^{gg} more than $C_{\ell}^{\kappa g}$ at fixed parameters, which likewise leads to a smaller upwards shift in S_8 . The new redshift marginalisation is described in Section 8.2. Finally, in Fig. 39 we show the impact of switching the approximate fiducial cosmology correction from Krolewski et al. (2021) to the formalism described in Sec. C.1. The impact is mostly a change in the $S_8 \cdot \Omega_m$ degeneracy direction.

I.2. Lensing Monte Carlo Norm Correction

A well known effect in CMB lensing reconstruction is a misnormalisation of the lensing reconstruction resulting from performing lensing reconstruction in the presence of a mask (see Sailer et al. (in prep.) for a pedagogical discussion of this effect and Benoit-Levy et al. (2013) or Carron (2023) for a more technical discussion). Typically CMB lensing auto-spectrum analyses have computed Monte Carlo corrections for this effect using Gaussian simulations (Qu et al. 2023; Carron et al. 2022; Planck Collaboration et al. 2020b). The normalisation correction is obtained as the ratio of the cross-correlation between appropriately masked input lensing convergence with the lensing reconstruction to the auto-correlation of the known input convergence. For the case of the cross-correlation this implies

$$A_{\ell}^{\rm MC} = \frac{C_{\ell}^{\kappa_{\rm in,\kappa-mask}\kappa_{\rm in,g-mask}}}{C_{\ell}^{\hat{\kappa}\kappa_{\rm in,g-mask}}}.$$
(I11)



Figure 39. Comparison of three methods for the fiducial cosmology correction on the Blue sample in *Planck* x unWISE. Replacing the approximate fiducial cosmology correction in Krolewski et al. (2021) with the expressions in Sec. C.1 rotates the S_8 - Ω_m contours.

Here $\hat{\kappa}$ is the masked CMB lensing reconstruction, $\kappa_{in,\kappa-mask}$ is the input lensing convergence masked with the lensing mask, and $\kappa_{in,g-mask}$ is the input lensing convergence masked using the galaxy mask. As one can see here the correction in principle depends on the region of overlap and thus cannot be universally computed, though we find in practice that the footprint dependence is small (< 1%). An unbiased estimate of the CMB lensing cross-correlation can then be obtained as

$$C_{\ell}^{\hat{\kappa}_{\rm MC}g} = C_{\ell}^{\hat{\kappa}g} A_{\ell}^{\rm MC}.$$
(I12)

We compute A_{ℓ}^{MC} using 480 Gaussian simulations and corresponding lensing reconstructions provided by Carron et al. (2022). We find that on the scales of interest for our analysis the norm correction results in a nearly scale invariant increase in the amplitude of $C_{\ell}^{\kappa g}$ by around 2% (see Fig. 40). We note that Krolewski et al. (2021) used the *Planck* PR3 lensing reconstruction (Planck Collaboration et al. 2020b) which also exhibits very similar behaviour (see Fig. 40). Given the errorbars on our cross-correlation measurement accounting for this effect is unavoidable and leads to a ~ 0.6 σ increase in the inferred value of S_8 (see Table 10).

We perform the same test of unbiased power spectrum recovery shown for ACT in Sec. 5 for *Planck* and find recovery to within 1% after application of the norm correction (see Fig. 41).

I.3. Updated cosmological constraints from the cross-correlation of Planck lensing and unWISE

With the improvements laid out above we perform a reanalysis of the cross-correlation of *Planck* lensing and unWISE galaxies. We adopt the priors and scale cuts consistent with our ACT analysis and estimate the covariance as described



Figure 40. Shown here are the Monte Carlo norm corrections for the cross-correlation of the unWISE galaxies with *Planck* PR4 (Carron et al. 2022) and PR3 (Planck Collaboration et al. 2020b) lensing reconstructions. We also show the equivalent norm correction for the ACT DR6 lensing reconstruction used in this work. That correction is much larger due to the application of Fourier space masking prior to lensing reconstruction that aims to reduce contamination from instrumental effects such as ground pick-up and features in the scan strategy. Moreover, we show the fractional 1σ errors on our $C_{\ell}^{\kappa g}$ (here shown for the Green galaxy sample) for comparison.



Figure 41. As with ACT we recover the input C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ on *Planck* simulations to better than 1% on all scales. The lightly coloured errorbars indicate our measurement errors, while the dark errorbars show the error on the mean of 480 Gaussian simulations.

in Sec. 6. We find $S_8 = 0.805 \pm 0.018$ from the combination of $C_{\ell}^{\kappa g}$ and C_{ℓ}^{gg} alone and $\sigma_8 = 0.810 \pm 0.018$ when adding BAO information.

J. IMPACT OF GALAXY SYSTEMATICS WEIGHTING

Figures 43 and 44 show the correlation of the unWISE number counts with various systematic tracers discussed in Sec. 3.1.3 before and after correction for the trends with stellar density and W2 limiting magnitude (corrected correlations are shown as red lines in the relevant panels).

In this work, we include systematics weights which had not been applied previously, but we also no longer filter out the low- ℓ ($\ell < 20$) modes in the unWISE map (in contrast to what was done in Krolewski et al. 2019, 2021). The large-scale filtering has a similar effect to systematics weighting, also reducing large-scale power by removing correlations between systematics and the true galaxy density. However, removing large scales in harmonic space complicates the use of the MASTER algorithm Hivon et al. (2002) to obtain unbiased bandpowers through mode decoupling. Hence, we

UNWISE X ACT DR6 COSMOLOGY

	Ω_m		σ_8		S_8		$S_8^{ imes}$	
			Planck	$PR4 \times u$	nWISE only			
Blue	$0.425\substack{+0.054\\-0.088}$	(0.390)	$0.703\substack{+0.052\\-0.059}$	(0.714)	$0.828\substack{+0.022\\-0.033}$	(0.814)	$0.815\substack{+0.021\\-0.030}$	(0.804)
Green	$0.259^{+0.022}_{-0.025}$	(0.250)	0.861 ± 0.032	(0.865)	0.798 ± 0.023	(0.790)	0.804 ± 0.022	(0.797)
Joint	$0.273\substack{+0.020\\-0.024}$	(0.279)	0.846 ± 0.032	(0.836)	0.805 ± 0.018	(0.807)	0.809 ± 0.017	(0.810)
			Planck P	$R4 \times un$	WISE + BAO			
Blue	0.3102 ± 0.0078	(0.311)	0.795 ± 0.026	(0.793)	0.808 ± 0.025	(0.808)	0.807 ± 0.025	(0.806)
Green	0.3042 ± 0.0077	(0.304)	0.3042 ± 0.0077	(0.814)	0.823 ± 0.020	(0.820)	0.822 ± 0.020	(0.820)
Joint	0.3050 ± 0.0075	(0.305)	0.810 ± 0.018	(0.808)	0.816 ± 0.017	(0.816)	0.816 ± 0.016	(0.815)

Table 11. Here we summarise the one dimensional marginalise posteriors on the key cosmological parameters from our reanalysis of the cross-correlation between unWISE galaxies and the *Planck* lensing reconstructions. We use the newest *Planck* PR4 lensing maps (Carron et al. 2022) and our improved model detailed in this paper. We provide best fit values in parentheses following the marginalised constraints.



Figure 42. Updated parameter constraints from the cross-correlation of *Planck* lensing and unWISE galaxies. From the combination of C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ only for both samples of unWISE galaxies (left) we find $S_8 = 0.805 \pm 0.018$. With additional information on from BAO (right) we obtain $\sigma_8 = 0.810 \pm 0.018$.

no longer adopt this method. Nevertheless, the net effect of replacing the low- ℓ filtering with the angular systematics weights is at most 0.5% (Fig. 45).

K. SPECTROSCOPIC SAMPLES USED IN CROSS-CORRELATION REDSHIFTS

We follow the methodology described in Section 5.4 of Krolewski et al. (2019) with updated spectroscopic galaxy samples and revised magnification bias measurements for the spectroscopic samples. We split the spectroscopic samples into northern and southern galactic cap (NGC and SGC) regions, allowing for different spectroscopic biases between the hemispheres (Alam et al. 2017; Ross et al. 2017) and to test their consistency. We summarise the spectroscopic samples used in Table 12 and Figures 46 and 47.



Figure 43. We show here the correlations of the unWISE number density from the Blue sample with the imaging systematics templates discussed in Sec. 3.1.3. Prior to applying any corrections the number density is significantly correlated with all templates. All correlations are reduced dramatically after applying corrections for the correlation with W2 depth and stellar density. Because many of the templates are correlated the improvement also extends to those templates for which we did not explicitly correct. To guard against overfitting and because the residual impact on the scales of interest is small we do not correct for further contamination even though some of the correlations are still significant even after the correction.



Figure 44. Similar to Fig. 43 but here shown for the Green sample of unWISE galaxies. As with the blue sample we correct the data for correlations with W2 depth and stellar density, improving the correlations with all of the imaging templates used.



Figure 45. Ratio of galaxy-galaxy and galaxy-CMB lensing spectra after weighting against W2 5 σ limiting magnitude and Gaia stellar density (as described in Sec. 3.1.3) to the same spectra without such weighting. The correction for these observational effects reduces the power in the auto-correlation by up to 3% on large scales, while it does not significantly impact the measured cross-correlation spectra. We also show for comparison the spectra used in Krolewski et al. (2019, 2021); these filtered our large scales ($\ell < 20$) in harmonic space as an alternative method for mitigating such observational effects. The difference between the spectra using the two methods is about 0.5%. In this comparison we have corrected for the simulation derived transfer function described in Sec. 5.

Sample	~ .	~	N	Jackknife	Area
Sample	~min	~max	1 V	regions	(deg^2)
LOWZ NGC	0.0	0.5	273547	34	5661
LOWZ SGC	0.0	0.5	120585	13	2393
CMASS NGC	0.1	0.8	544308	37	6676
CMASS SGC	0.1	0.8	193899	13	2427
eBOSS DR16 LRG NGC	0.6	1.0	107500	63	2898
eBOSS DR16 LRG SGC	0.6	1.0	67316	38	1812
eBOSS DR16 QSO NGC	0.8	2.2	218209	61	2927
eBOSS DR16 QSO SGC	0.8	2.2	125499	38	1825
BOSS DR12 QSO NGC	2.0	4.0	69204	36	6569
BOSS DR12 QSO SGC	2.0	4.0	16590	12	2176

Table 12. Properties of the spectroscopic samples used for cross-correlation redshifts. Areas are defined using the overlap of the spectroscopic survey mask and the unWISE mask.



Figure 46. Maps of the spectroscopic samples used to determine b(z)dN/dz of the unWISE galaxies. Top row is NGC and bottom row is SGC. Together the spectroscopic samples cover a representative subset of the unWISE footprint (as can also be seen in Fig. 48).

We use both unWISE and spectroscopic sample systematics weights in computing the Davis-Peebles (Davis & Peebles 1983) estimator for $\bar{w}_{sp}(\theta)$. The unWISE weights are described in Section 3.1.3 above and the spectroscopic sample weights are the standard systematic weights for BOSS and eBOSS (Reid et al. 2016; Eftekharzadeh et al. 2015).

The eBOSS LRGs are targeted at lower priority than eBOSS quasars. Unlike BOSS, which used multiple passes to eventually reach lower priority objects, eBOSS did not have enough passes to reach lower priority objects within the fiber radius of higher priority objects. Therefore, no LRGs were observed within the fiber scale (62") of any quasar target. This poses a problem for comparing LRG-unWISE and quasar-unWISE cross-correlations in the region where both spectroscopic tracers overlap (0.8 < z < 1.0), since the LRGs and quasars will be anti-correlated on small scales due to fiber collisions. We therefore remove any unWISE galaxies and randoms within 62" of an eBOSS quasar target, as these unWISE galaxies will also be anti-correlated with the eBOSS LRGs due to their positions near the quasars.

We revise the magnification bias measurements presented in Appendix C of Krolewski et al. (2019) and update them for the eBOSS LRGs. Magnification bias of the spectroscopic samples can create spurious correlations between widely separately photometric and spectroscopic redshift bins, thus biasing the clustering dN/dz in the tails; the spectroscopic magnification bias is relevant when the spectroscopic sample lies at higher redshift than the photometric sample.

First, Krolewski et al. (2019) did not correctly consider the impact of magnification bias on fixed-aperture flux measurements. Lensing magnification preserves surface brightness, brightening galaxies by making them larger on the sky. The impact of magnification on a fixed-aperture flux measurement therefore depends on the light profile: if it is spatially constant, then magnification has no effect, but in the more realistic case where it rises towards the galaxy's center, lensing magnification will brighten the galaxy, but not by as much as if the aperture contained all of the galaxy's light. CMASS and LOWZ use four different measurements of magnitude: model magnitudes, consisting



Figure 47. Here we show the redshift distribution of spectroscopic samples used to measure the cross-correlation redshifts for the unWISE samples (NGC and SGC are shown as solid and dashed lines respectively). The spectroscopic tracers used in this work span the full redshift range of the unWISE samples up to $z \sim 3$.

of the best-fit exponential or deVaucoleurs profile; cmodel magnitudes, a linear combination of the exponential and deVaucoleurs profiles; PSF magnitudes, which fit a PSF model to the source and therefore only consider flux within the PSF solid angle; and fib2 magnitudes, which only count flux within the 2" fibers. We approximate the model and cmodel magnitudes as measuring the full galaxy flux (therefore being maximally affected by lensing magnification). For PSF magnitudes and fib2 magnitudes, we magnify the galaxy profile (assuming an isotropic linear combination of deVaucoleurs and exponential profiles with effective radius given in the catalog) and truncate the fluxes at either 2" or the mean PSF size in a given band (1.32" in r, 1.26" in i, 1.29" in z).

Second, in a spectroscopic survey, lensing magnification affects the redshift success rate, the probability with which a redshift can be estimated from a observed spectrum, which is lower for fainter galaxies. LOWZ galaxies are sufficiently bright that redshift failures and successes have very similar magnitude distributions (Fig. 7 in Reid et al. 2016), i.e. the faintest LOWZ galaxies are considerably brighter than the "limiting magnitude" of the spectrograph and therefore increasing or decreasing their flux by a small amount will not change the success rate. CMASS galaxies are fainter, and we estimate from Fig. 7 in Reid et al. (2016) that $\frac{dlnN}{difib2} = 0.32$ due to the change in the redshift success rate (by differencing the $i_{\rm fib2}$ distribution at the faint end of CMASS galaxies, comparing $i_{\rm fib2} = 21.3$ and 21.4). Since $i_{\rm fib2}$ has a fixed aperture, we further correct for its reduced sensitivity to lensing magnification compared to total magnitudes. For eBOSS LRGs, we also correct for changes in redshift success rate using the measurement of redshift success against spectral signal-to-noise in Fig. 5 of Ross et al. (2020). We translate signal-to-noise to magnitudes assuming background-limited sources and account for the finite aperture (fiber size) in which the spectral SNR is measured. For eBOSS quasars, we re-run the XDQSO quasar probability code to measure the response of the quasar selection to small magnifications (the change in the redshift success rate with magnitude is negligible; see Fig. 4 of Ross et al. 2020). We also account for the reduced efficiency of star/galaxy separation at fainter magnitudes due to magnification (where a point source can be demagnified and then spuriously considered an extended source Strauss et al. 2002)³⁴.

³⁴ https://classic.sdss.org/dr7/products/general/stargalsep.html

UNWISE X ACT DR6 COSMOLOGY

		LOWZ				CMASS				eBOS	S DR16	LRG				
z_{\min}	$z_{\rm max}$	$z_{{ m eff},ss}$	$z_{{\rm eff},sp}$	s	$b_{ m NGC}$	$b_{\rm SGC}$	$z_{{ m eff},ss}$	$z_{\mathrm{eff},sp}$	s	$b_{ m NGC}$	$b_{\rm SGC}$	$z_{{ m eff},ss}$	$z_{\mathrm{eff},sp}$	s	$b_{ m NGC}$	$b_{\rm SGC}$
0.00	0.05	0.042	_	1.975	1.326	1.244										
0.05	0.10	0.079	0.070	1.069	1.368	1.305										
0.10	0.15	0.124	0.123	0.912	1.518	1.491										
0.15	0.20	0.175	0.175	1.049	1.683	1.682	0.179	0.177	0.756	3.078	1.735					
0.20	0.25	0.225	0.223	1.015	1.892	1.755	0.226	0.223	0.663	1.204	1.448					
0.25	0.30	0.276	0.273	0.972	1.978	2.051	0.276	0.273	0.982	1.775	1.650					
0.30	0.35	0.327	0.325	0.923	1.984	2.062	0.324	0.323	1.121	1.765	1.798					
0.35	0.40	0.372	0.369	1.014	2.051	2.077	0.382	0.371	1.161	2.273	2.078					
0.40	0.45	0.420	0.419	1.179	2.211	2.109	0.439	0.429	0.972	2.055	2.048					
0.45	0.50	0.460	0.463	1.511	2.541	2.187	0.479	0.475	0.937	2.065	2.000					
0.50	0.55						0.524	0.524	1.040	2.034	2.101					
0.55	0.60						0.572	0.573	1.148	2.136	2.073					
0.60	0.65						0.620	0.617	1.352	2.233	2.147	0.627	0.620	0.539	2.137	2.114
0.65	0.70						0.669	0.672	1.553	2.400	2.236	0.675	0.675	0.595	2.094	2.107
0.70	0.75						0.717	0.715	1.957	2.509	2.434	0.725	0.718	0.721	2.202	2.066
0.75	0.80						0.765	0.772	2.239	2.666	2.918	0.773	0.777	0.798	2.363	2.333
0.80	0.85											0.821	0.825	0.884	2.458	2.514
0.85	0.90											0.871	0.868	0.890	2.572	2.400
0.90	0.95											0.919	0.922	0.949	2.517	2.405
0.95	1.00											0.969	0.971	0.947	2.211	2.382

Table 13. The table summarises the effective redshifts for the spectroscopic auto-correlation, $z_{\rm eff,ss}$, and the cross-correlation with Green unWISE sample $z_{\rm eff,sp}$. We also summarise the measured magnification (s) and clustering biases ($b_{\rm NGC}$ and $b_{\rm SGC}$ for the NGC and SGC respectively) for the spectroscopic samples. The effective redshifts and magnification bias are determined from the NGC only, whereas the clustering bias is measured separately for NGC and SGC. Estimates of the magnification biases on the SGC are consistent with the NGC but with increased uncertainties. Therefore, we use the magnification biases measured on the NGC throughout. We allow for a different clustering bias between the hemispheres. We do not report a value for $z_{\rm sp}$ at z < 0.05, because there are zero unWISE-Green galaxies matched to COSMOS photometric redshifts in this redshift range.

Third, we correctly account for the impact of lensing magnification at both the bright and faint ends of the sample (i.e. if the sample has a bright cut, de-magnified objects near the threshold are more likely to constitute the sample). This particularly matters for the eBOSS LRGs, which have a bright cut to prevent overlap with BOSS CMASS. We follow the method of Elvin-Poole et al. (2023) and Samuroff et al. (2023), in which the change in number counts is computed both when magnifying and de-magnifying the galaxies, and the total change is the difference of the two: $\delta N = \delta N_+ - \delta N_-$. We use a small finite step size of $\delta m = 0.02$ to compute magnification biases.

Magnification bias measurements are shown in Table 13 and 14. These results are reasonably consistent with external estimates: von Wietersheim-Kramsta et al. (2021) find s = 0.77 for LOWZ at z = 0.35 and s = 1.05 for CMASS at z = 0.65; (Samuroff et al. 2023) find s = 0.21 for CMASS at z = 0.5 and s = 0.81 for eBOSS LRG at z = 0.8, and Leauthaud et al. (2022) find $s \sim 1.2$ at z = 0.65. The agreement between these measurements is hindered by differences in the exact samples used (i.e. redshift binning) and methodology adopted (measuring the slope of the luminosity function in a single band, or properly taking into account the complex selection function). During preparation of this paper, we compared results to Wenzl et al. (2023), who also account for the full complexities of the selection function (with some methodological improvements over our results). We find very similar results for magnification bias of CMASS and LOWZ in redshift bins, agreeing to $\Delta s \sim 0.05$, well within our assumed systematic error of $\Delta s = 0.2$.

For quasars, Wang et al. (2020) finds $s \sim 0.1-0.2$ for quasars following the eBOSS quasar luminosity function with an absolute magnitude cutoff; Lepori et al. (2020) and Iršič et al. (2016) find s = 0.295 for BOSS quasars; Myers et al. (2003, 2005) find s = 0.29 for 2dF quasars.

We propagate errors on the magnification bias estimates into our uncertainty on the clustering redshift by assuming that magnification bias errors are dominated by a systematic error of $\Delta s = 0.2$. This is motivated by the level of

			eBOS	S DR16	BOSS	5 DR12	QSO		
z_{\min}	$z_{\rm max}$	$z_{{ m eff},ss}$	$z_{{\rm eff},sp}$	s	$b_{ m NGC}$	$b_{ m SGC}$	$z_{{\rm eff},ss}$	$z_{\mathrm{eff},sp}$	s
0.00	0.20								
0.20	0.40								
0.40	0.60								
0.60	0.80								
0.80	1.00	0.910	0.910	0.126	1.563	1.840			
1.00	1.20	1.107	1.115	0.153	2.062	2.088			
1.20	1.40	1.300	1.279	0.151	2.055	2.095			
1.40	1.60	1.499	1.482	0.171	2.406	2.398			
1.60	1.80	1.699	1.674	0.163	2.315	2.520			
1.80	2.00	1.895	1.894	0.180	2.789	3.019			
2.00	2.20	2.090	2.059	0.163	3.269	2.928	2.176	2.112	0.494
2.20	2.40						2.307	2.302	0.213
2.40	2.60						2.481	2.461	0.234
2.60	2.80						2.671	2.616	0.387
2.80	3.00						2.904	2.820	0.793
3.00	3.20						3.095	3.150	0.821
3.20	3.40						3.270		1.047

Table 14. Same as Table 13, but for the quasar samples. For the BOSS DR12 QSOs, we do not measure the spectroscopic bias, but rather use the fit from Laurent et al. (2017). We do not report a value for z_{sp} at 3.2 < z < 3.4, because there are zero unWISE-Green galaxies matched to COSMOS photometric redshifts in this redshift range.

agreement between our results and past results, and by Elvin-Poole et al. (2023), who find agreement to $\Delta s \sim 0.2$ between the approximate method that we use and more sophisticated methods based on image simulations. We find the magnification bias measurements are consistent between NGC and SGC, so we use NGC measurements due to its higher density and smaller uncertainties. We assume the magnification bias error is entirely correlated between NGC and SGC for each tracer.

We place the measurement in each bin at its effective redshift rather than the mean redshift of the bin. This is a very minor effect except for the $\Delta z = 0.2$ quasar bins in the tail of the unWISE redshift distribution. We compute the effective redshift for the unWISE cross-correlations as

$$z_{\text{eff},sp} = \frac{\sum_{p} z_{p,i} \ 1/\chi_{p,i}^{2} \ dN_{s}/dz}{\sum_{p} \ 1/\chi_{p,i}^{2} \ dN_{s}/dz}$$
(K13)

and for the spectroscopic auto-correlations as

$$z_{\text{eff},ss} = \frac{\int dz \, z \, 1/\chi^2 dN_s/dz}{\int dz \, 1/\chi^2 dN_s/dz} \tag{K14}$$

We can only obtain estimates of the redshifts of individual photometric galaxies, $z_{p,i}$, in the unWISE sample for the small number of objects that have counterparts in the 2 deg² COSMOS field. Because the number of objects is small we can not reliably estimate a smooth dN_p/dz and therefore explicitly write the integral as a sum over the photometric sample in Eq. K13.

With an improved measurement of eBOSS quasar clustering, we switch from using the parameterised form of the quasar bias from Laurent et al. (2017) to our own measurements of the quasar bias from its auto-correlation, similar to the method adopted for CMASS and LOWZ (Fig. 6 in Krolewski et al. 2019). We continue to use the Laurent et al. (2017) bias function for the DR12 QSOs, which have a noisier autocorrelation measurement. Our quasar bias measurements are consistent with Laurent et al. (2017) within the errorbars, but they allow for deviations from a smooth trend and take advantage of the larger sample size, especially compared to the measurement of Laurent et al. (2017) which uses a preliminary release of eBOSS quasars (DR14). Spectroscopic biases are given in Table 13 and 14.

As in Krolewski et al. (2019), we verify that our cross-correlation redshifts are consistent between different spectroscopic tracers and between hemispheres. We find good agreement (p > 0.05) for all overlapping tracers with the same

Spec. sample 1	Spec. sample 2	Blue χ^2/dof	Green χ^2/dof
DR12 CMASS N	DR12 CMASS S	7.4/14	17.7/14
DR12 CMASS N	DR12 LOWZ N	5.6/7	6.6/7
DR12 CMASS N	DR12 LOWZ S	6.2/7	5.1/7
DR12 CMASS N	DR16 LRG N	3.6/4	2.5/4
DR12 CMASS N	DR16 LRG S	1.4/4	1.2/4
DR12 CMASS S	DR12 LOWZ N	4.9/7	10.3/7
DR12 CMASS S	DR12 LOWZ S	3.7/7	10.7/7
DR12 CMASS S	DR16 LRG N	5.1/4	3.3/4
DR12 CMASS S	DR16 LRG S	3.6/4	1.4/4
DR12 LOWZ N	DR12 LOWZ S	5.7/9	1.7/9
DR16 LRG N	DR16 LRG S	2.9/8	8.5/8
DR16 QSO N	DR16 QSO S	10.2/7	7.4/7
DR16 QSO N	DR12 QSO N	_	3.0/1
DR16 QSO N	DR12 QSO S	_	1.5/1
DR16 QSO S	$\rm DR12~QSO~N$	_	0.0/1
DR16 QSO S	DR12 QSO S	_	2.7/1
DR12 QSO N	DR12 QSO S	_	11.3/8

Table 15. Consistency of cross-correlation estimates of b(z)dN/dz for different spectroscopic tracers. We show the χ^2 between the clustering redshift estimates obtained using different spectroscopic subsamples and the number of degrees of freedom, i.e. the number of overlapping redshift bins between the samples. The Blue sample of unWISE galaxies has no significant overlap with the high redshift BOSS DR12 quasars (compare Fig. 47).

redshift binning ($\Delta z = 0.05$ for galaxies and $\Delta z = 0.20$ for quasars; see Table 15). Furthermore, since the eBOSS LRGs overlap with the eBOSS quasars in the redshift range 0.8 < z < 1.0 we have an additional consistency check. We find that for both the Blue and Green unWISE samples, the quasar cross-correlation is slightly higher than the LRG cross-correlation (see Fig. 50). To quantitatively compare the two, we create both a $\Delta z = 0.2$ bin for the LRGs and a $\Delta z = 0.05$ bin for the quasars. For Blue, in the single $\Delta z = 0.2$ bin, the quasars are higher by 1.77σ . In the four $\Delta z = 0.05$ bins, the quasars are higher by 1.25σ , 0.77σ , 2.35σ and 1.65σ . For Green, in a single $\Delta z = 0.2$ bin, the quasars are higher by 1.06σ . Therefore, while a discrepancy certainly exists (especially for Blue), it is not large enough to be particularly concerning, especially given the large number of comparisons in Table 15. Future spectroscopic surveys such as DESI will significantly help the situation-both by improving the statistics and by offering multiple tracers for improved systematics testing.

The unWISE sample exhibits some inhomogeneity, particularly in ecliptic latitude (see Sec. 7.2), likely due to the variable unWISE coverage depth. This inhomogeneity may also result in spatially varying dN/dz. Given that the dN/dz is based on cross-correlations measured over only a fraction of the unWISE-ACT footprint (see Fig. 46), we desire that this area is representative of the full footprint. In Fig. 48, we show that the unWISE-ACT and unWISE-BOSS footprints match very well in (absolute value) ecliptic and Galactic latitude. The absolute value is most relevant here because the WISE depth is greatest at the poles and shallowest at the equator; likewise, Galactic contamination is small at both poles. Therefore, the sign of the latitude (hemisphere) matters much less than the distance from the equator. In Table 16, we summarise the mean absolute ecliptic and Galactic latitude of the BOSS and eBOSS surveys. The combined spectroscopic regions are similar in mean absolute ecliptic and Galactic latitude to the unWISE-ACT footprint. Lizancos & White (2023) showed that as long as the redshift distribution assumed in the model matches the mean redshift distribution of the sample used, anisotropies in the dN/dz across the survey footprint do not bias the inferred cosmology.

In Fig. 49, we show the variations in dN/dz considered as part of our null tests: restricting to the NGC; restricting to the SGC; and using a larger minimum scale cut in the clustering redshift measurement ($s_{\min} = 3.99 \ h^{-1}$ Mpc rather than the default $s_{\min} = 2.52 \ h^{-1}$ Mpc).

In Fig. 50, we compare the new cross-correlation b(z)dN/dz to the old ones, breaking the changes into two parts: the changes or additions to the dataset (adding SGC tracers, eBOSS LRGs, and exchanging the eBOSS DR14 quasars



Figure 48. Distribution of absolute value of ecliptic and Galactic latitude for ACT-unWISE footprint that we measure C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ on; and unWISE-BOSS footprint that we use for cross-correlation redshifts. The fact that the unWISE-BOSS footprint is a good match to the ACT-unWISE footprint means that our estimate of b(z)dN/dz from the cross-correlation of unWISE with BOSS is likely to be representative of the bias weighted redshift distribution on the ACT-unWISE footprint.

Region	Galactic $\langle b \rangle$ (°)	Ecliptic $\langle \beta \rangle$ (°)
ACT-unWISE	53.8	25.3
eBOSS N	55.4	39.2
eBOSS S	47.0	15.1
eBOSS N+S	52.1	29.5
BOSS N	55.2	29.5
BOSS S	45.7	14.3
BOSS N+S	52.9	25.8

 Table 16. Mean absolute value of ecliptic and Galactic latitude for various footprints. The combined spectroscopic regions are similar in mean absolute ecliptic and Galactic latitude to the unWISE-ACT footprint.

for the final eBOSS DR16 quasars), and a change in method (using unWISE weights in the cross-correlation as well as the spectroscopic weights). For Blue, the mean redshift changes from 0.714 to 0.690 when adding more data, and finally to 0.697 when adding the unWISE weights. This is a shift of 0.74σ compared to the standard deviation of 0.023 as measured from 1000 samples of dN/dz, and this large shift is the dominant source for the upwards shift in S_8 in the Planck reanalysis. For Green, the mean redshift changes from 1.352 to 1.357 when adding new data, and to 1.355 when adding the unWISE weights (compared to an uncertainty of 0.022: a shift of 0.14σ). The new dN/dz show clearly reduced uncertainty, especially in the eBOSS quasar redshift range 0.8 < z < 2.0, where we have used almost four times as much area. Moreover, the new dN/dz have improved redshift resolution at 0.8 < z < 1.0 from the eBOSS LRG, which reduces the uncertainty on dN/dz in this range although it is not apparent from the plot. It is also clear that most of the change in dN/dz comes from the addition of the southern footprints of the spectroscopic tracers and the additional eBOSS data; adding unWISE weights makes little difference. Interestingly, the weights seem to make the largest difference at low redshift, where they suppress dN/dz, but make little difference at z > 0.7. This suggests that there may have been some small residual systematic power in the cross-correlation redshifts with BOSS CMASS and LOWZ. The changes in dN/dz are largest at z > 0.8, where the statistical fluctuations related to adding more data. The changes in dN/dz are largest at z > 0.8, where the statistical error is biggest.

L. DETAILED DISCUSSION OF POST-UNBLINDING CHANGES

As discussed in Sec. 9.2 we made four minor corrections to the analysis after unblinding. This includes a modification to the treatment of the mask induced mode coupling of the shot noise, a small correction to the transfer function obtained from simulations, the inclusion of the cross-covariance between the two unWISE samples, and a bug-fix affecting the implementation of the higher order biases. The impact of these corrections partially offsets and they



Figure 49. Comparison of b(z)dN/dz used for different null tests: NGC only, SGC only, and increasing the minimum scale used in the clustering redshifts. The estimates are consistent within the uncertainties and we show in Sec. 9.7 that our cosmological inference is stable with respect to these choices.

results in a net shift of -0.1σ in S_8 compared to our initially unblinded results. Here we provide some more detail on these corrections.

In order to compare our theory predictions of C_{ℓ}^{gg} and $C_{\ell}^{\kappa g}$ with the observed spectra, we convolve them with a bandpower window which captures the effect of the approximate mode decoupling performed on the data. As discussed in Sec. 4 the expectation value of the observed pseudo- C_{ℓ} is related to the true underlying C_{ℓ} through a mode coupling matrix $M_{\ell\ell'}$,

$$\left\langle \tilde{C}_{\ell} \right\rangle = \sum_{\ell'} C_{\ell'} M_{\ell\ell'}. \tag{L15}$$

For the binned power spectra \hat{C}_b this relation can only approximately be inverted. The decoupled observed power spectrum is then

$$\hat{C}_{b}^{\text{decoupled}} = \sum_{b'} \widehat{M^{-1}}_{bb'} \hat{C}_{b'} \tag{L16}$$

where $\widehat{M^{-1}}_{bb'}$ is the inverse of the binned mode coupling matrix obtained under the assumption that the power spectrum is piecewise constant across each bin³⁵. The expectation value of the decoupled power spectrum is therefore given in terms of the true underlying C_{ℓ} by

$$\left\langle \hat{C}_{b}^{\text{decoupled}} \right\rangle = \sum_{b'} \widehat{M^{-1}}_{bb'} \sum_{\ell} w_{\ell}^{b'} \sum_{\ell'} M_{\ell\ell'} C_{\ell'} \tag{L17}$$

where w_{ℓ}^{b} are the weights associated with each multipole in bin b. In order to compare a given theory prediction with the observed and decoupled bandpowers, one thus needs to convolve it with a bandpower window.

$$F_{b,\ell'} = \sum_{b'} \widehat{M^{-1}}_{bb'} \sum_{\ell} w_{\ell}^{b'} \sum_{\ell'} M_{\ell\ell'}.$$
 (L18)

³⁵ This is the default approximation implemented in NaMaster (Alonso et al. 2019). An alternative approach suggested in Hivon et al. (2002) is to assume the $\ell(\ell+1)C_{\ell}$ is constant instead.



Figure 50. Comparison between the old cross-correlation b(z)dN/dz (from Krolewski et al. 2019, 2021) and the updated b(z)dN/dz used in this work. We show both the data points (i.e. cross-correlation measurement, averaged over scale, divided by spectroscopic sample bias) and the smooth best fits. The b(z)dN/dz used in Krolewski et al. (2021) are shown in blue (with the smooth best fit shown as a dashed line). The orange data points and dashed line show the estimated bdN/dz after adding the southern spectroscopic samples, the eBOSS LRGs, and replacing the eBOSS DR14 quasars with the DR16 quasars (this also includes the correction to the treatment of the spectroscopic magnification bias). Finally, the green data points and dotted line shows estimates which additionally include unWISE weights in the cross-correlation (final version used in paper). Points are offset in redshift for clarity.

For a purely white power spectrum no mode coupling is induced. This is because, as was pointed out in Hivon et al. (2002), $\sum_{\ell'} M_{\ell\ell'} = w_2$. So no mode decoupling is required on any white noise component of the signal. However, since the mode coupling matrix is computed in practice only to finite ℓ , the operation of masking the noise and convolving the noise power spectrum with the finite mode coupling matrix are not equivalent ($\sum_{\ell'=0}^{\ell_{max}} M_{\ell\ell'} \neq w_2$). Consequently, while the white noise present in the map is in fact scaled simply by w_2 the operation applied to the white noise component of the theory spectrum does not reflect this. In principle the mode coupling matrix is also inaccurate for the signal component of the spectrum for the same reason, however, because the signal falls off quickly with ℓ the effect is strongly suppressed. On our simulations we find that power spectrum is most accurately recovered if we instead adopt the following approach

$$C_b^{\text{th,bpw conv.}} = \sum_{b'} \widehat{M^{-1}}_{bb'} \sum_{\ell} w_{\ell}^{b'} \left[\sum_{\ell'} \left(M_{\ell\ell'} \left(C_{\ell'}^{\text{th}} - N_{\text{shot}} \right) + M_{0,\ell'} N_{\text{shot}} \right) \right].$$
(L19)

The inaccuracy in our previous treatment, which did convolve the shot noise with the mode coupling matrix, due to this effect is very small (< 0.1%). In the process of correcting this effect, however, we noticed that because our Gaussian realisations of the lensing field are band limited to $\ell_{\max}^{Gaussian \kappa} = 5100 < 3 \times \text{NSIDE}$, our correlated realisations of the galaxy field were not generated with the correct power at $\ell > \ell_{\max}^{Gaussian \kappa}$. This was originally not taken into account when testing the power spectrum recovery and computing the transfer function (see Sec. 5). Since the mode coupling matrix mixes even widely separated scales, the incorrect power at small scales also affects some scales within our analysis range at the ~ 0.5% level. Given that the fractional error on C_{ℓ}^{gg} is of the order 2% this explains the ~ 0.2 σ shift in our results.

Additionally, it was noticed that the cross-covariance between the two redshift samples was erroneously omitted from the analysis. When corrected, this results in shift of $\sim -0.3\sigma$ in the value inferred for S_8 from the joint analysis of the Blue and Green sample ($\sim -0.17\sigma$ for S_8^{\times}) and a slight degradation of constraining power by about 5%.
Finally, we noticed a small bug in the implementation of the bias evolution of the higher order biases. The fiducial bias evolution for $b_{2,L}$ was erroneously adopted for $b_{s,L}$ as well. Correcting this mistake required newly determining the priors on the higher order bias offsets from our N-body simulations. The prior mean on the offset for the second order Lagrangian bias for the Blue sample, $c_{2,L}^{\text{Blue}}$, increased from 0.39 to 0.55 (approximately 0.3σ) and since we set the width of the prior to be consistent with $c_{2,L}^{\text{Blue}} = 0$ this also required an increase of the width of the prior by ~ 30%. The prior mean for the shear bias parameter decreased from $c_{2,s}^{\text{Blue}} = 0.30$ to $c_{2,s}^{\text{Blue}} = 0.17$ with the width of the prior remaining basically unchanged. For the Green sample, the mean of the prior on $c_{2,s}^{\text{Green}}$ also increased by about 30% to 0.42 with the corresponding increase in width. The prior on $c_{2,s}^{\text{Green}}$ increased in width by about 15% while the mean increased by about 30%. We reran the model verification on our N-body sims and recovered the input cosmology well within the requisite 1σ in terms of the covariance appropriate for the simulation volume. This indicates that the error introduce by this bug was at least partially absorbed by our higher order bias marginalisation.

Since the higher order terms are much more significant for the Blue sample, the impact on the value of S_8 inferred from the Blue sample alone lies at about 0.3σ . On the other hand, our joint inference is unchanged apart from a small, ~ 5%, increase in the posterior width. This suggests that in the joint inference the combined data provide sufficient constraining power to probe the higher order biases beyond the information given by the prior.