Survey geometry and the internal consistency of recent cosmic shear measurements


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Accepted 2018 June 11. Received 2018 April 27; in original form 2018 June 7

ABSTRACT

We explore the impact of an update to the typical approximation for the shape noise term in the analytic covariance matrix for cosmic shear experiments that assumes the absence of survey boundary and mask effects. We present an exact expression for the number of galaxy pairs in this term based on the survey mask, which leads to more than a factor of three increase in the shape noise on the largest measured scales for the Kilo-Degree Survey (KiDS-450) real-space cosmic shear data. We compare the result of this analytic expression to several alternative methods for measuring the shape noise from the data and find excellent agreement. This update to the covariance resolves any internal model tension evidenced by the previously large cosmological best-fitting $\chi^2$ for the KiDS-450 cosmic shear data. The best-fitting $\chi^2$ is reduced from 161 to 121 for 118 degrees of freedom. We also apply a correction to how the multiplicative shear calibration uncertainty is included in the covariance. This change shifts the inferred amplitude of the correlation function to higher values. We find that this improves agreement of the KiDS-450 cosmic shear results with Dark Energy Survey Year 1 and Planck results.

Key words: gravitational lensing: weak – methods: data analysis – methods: statistical.
1 INTRODUCTION

Cosmic shear, the weak lensing of background galaxies by large-scale structure in the Universe, has been viewed for many years as a promising cosmological probe (Albrecht et al. 2006; Peacock & Schneider 2006; Weinberg et al. 2013). With the current generation of surveys, among them the Kilo-Degree Survey (KiDS; Kuijken et al. 2015) and the Dark Energy Survey (DES); we are beginning to see this promise fulfilled. Alongside the improved statistical power of these latest data sets, weak lensing methodology has developed rapidly over the last several years (Fenech Conti et al. 2017; Huff & Mandelbaum 2017; Sheldon & Huff 2017; Zuntz et al. 2017; Mandelbaum et al. 2017, 2018). With DES and KiDS, we have two independent weak lensing data sets of comparable constraining power — capable of constraining the amplitude of the two-point correlation function at better than the 3 per cent level in the case of DES. If the reduction of systematic uncertainties can keep pace, these constraints will rapidly improve over the coming years as the area of sky analysed and the integrated exposure time of each survey increases by several factors. A third, similarly powerful constraint from the Hyper-Suprime Cam (HSC) survey is also expected on a similar time-scale. The Canada–France–Hawaii Telescope Lens Survey (Joudaki et al. 2017) and the Deep Lens Survey (Jee et al. 2016) have also published cosmic shear constraints of similar precision to those of KiDS-450.

Cosmic shear is one of several low-redshift probes that tightly constrain the amplitude of structure in the standard cosmological model (ΛCDM). By comparing these constraints to high-redshift measurements like the cosmic microwave background (CMB), we can determine the parameters of ΛCDM (cold dark matter) to predict observations of the structure in and geometry of the Universe across its entire history. One important question is whether the amplitude of structure measured by lensing surveys is compatible with predictions based on CMB observations, extrapolated to late times assuming ΛCDM. In order to interpret these results as strong, reliable tests of ΛCDM, we must first subject them to stringent tests of internal and mutual consistency.

Internal consistency tests the agreement of different subsets of the weak lensing data vector with predictions based on the same ΛCDM parameters. The most simple such test is whether the best-fitting ΛCDM model describes these measurements (i.e. the correlation function) at an acceptable goodness-of-fit (χ²). Both the latest DES (Troxel et al. 2017, T17) and KiDS (Hildebrandt et al. 2017, H17) cosmic shear analyses reported a large χ² per degree of freedom (dof) in their initially released results.

Tests of mutual consistency, e.g. between the fully independent cosmic shear real-space two-point correlation function from KiDS-450 (H17) and DES Y1 (T17) are a unique opportunity to validate the entire analysis process. We note that mutual consistency between the DES and KiDS cosmic shear results is not a new result in this work — good agreement between the two data sets has already been demonstrated in T17.

In this paper we make a substantial progress on these critical consistency tests by implementing three improvements:

(i) to the shape noise component of analytic cosmic shear covariance matrix estimates relative to the initial implementation in both analyses. This correction was first implemented during the referee process for the DES Y1 analyses (Krause et al. 2017, K17; DES Collaboration 2017; T17). We find that it non-trivially impacts current generation weak lensing surveys, especially in the regime of small fields or non-compact footprints, including the KiDS-450 covariance matrix (H17) described below;

(ii) to the treatment of uncertainty in the multiplicative shear bias calibration in the H17 covariance matrix; and

(iii) to the angular separation estimate used for evaluating the predicted shear signal for the bins in H17 (see footnote 1 of Joudaki et al. 2018).

The first update has a significant impact on the interpretation of internal consistency in both H17 and T17, particularly the best-fitting χ². The shape noise component of the covariance in such surveys should thus be treated with care beyond the simple geometric approximations commonly used in earlier work. Each of the latter two updates shift the H17 contours only by a fraction of the 68 per cent CL, but jointly are relevant for evaluating mutual consistency and consistency with CMB observations.

This paper is organized as follows: In Section 2, we describe the cosmic shear data used and our analysis framework relative to H17 and T17. We describe methods to appropriately describe the shape noise term in non-compact or disjoint survey geometries and discuss how this impacts the H17 results in Section 3, including also update (ii) mentioned above. We then discuss the resulting interpretation of the cosmic shear results from H17 and T17 in Section 4. We conclude and discuss the outlook for future such studies in Section 5.

2 COSMIC SHEAR DATA AND ANALYSIS

We infer cosmological model constraints from the KiDS-450 and DES Y1 datasets using measurements of the cosmic shear real-space correlation function between tomographic bins i and j.

\[ \xi_h^i(\theta) = \frac{1}{2\pi} \int d\ell d\theta J_{i/(\ell)}(\theta) P_{i,\ell}(\theta). \]  

(1)

The corresponding convergence power spectrum is \( P_{\ell}^{(i)} \) and \( J_\theta \) is a Bessel function of the first kind. \( P_{\ell}^{(i)} \) is related to the non-linear matter power spectrum \( P_{\text{NL}} \) via the Limber approximation as

\[ P_{\ell}^{(i)}(\ell) = \int_0^{\ell H} d\chi q'(\chi) q''(\chi) P_{\text{NL}}(\ell H^2, \chi). \]  

(2)

Here \( \chi \) is the radial comoving distance and \( \chi_H \) is the horizon distance. The lensing efficiency function \( q' \) depends on the matter density \( \Omega_m \), Hubble constant \( H_0 \), scale factor \( a \), and the effective number density distribution of galaxies \( n'(\chi) \), normalized such that \( \int d\chi n'(\chi) = 1 \).

\[ q'(\chi) = \frac{3}{2} \Omega_m \left( \frac{H_0}{c} \right)^2 \int_{\chi}^{\chi_H} d\chi' n'(\chi') \frac{\chi - \chi'}{\chi'}. \]  

(3)

\(^1\)http://kids.strw.leidenuniv.nl
\(^2\)http://www.darkenergysurvey.org
\(^3\)http://hsc.mtk.nao.ac.jp/ssp/
\(^4\)We note that these effects were known to the KiDS collaboration before the publication of this paper, with updates already incorporated for their upcoming 9-band cosmological analysis (Hildebrandt et al in prep., KiDS Collaboration private communication).

\(^5\)By compact, we simply mean regions that minimize boundary length relative to area, i.e. a circle.
We assume a multivariate Gaussian in the correlation function measurements
\[ \ln \mathbf{L}(\mathbf{p}) = -\frac{1}{2} \sum_{ij} \left[ D_i - T_i(\mathbf{p}) \right] C^{-1}_{ij} \left[ D_j - T_j(\mathbf{p}) \right], \]
where \( \mathbf{p} \) is the set of parameters and \( T_i(\mathbf{p}) \) are the theoretical predictions for \( \xi_{\pm} \). We utilize the data vector \( \xi_{\pm}^D(\Delta) \), redshift distributions, covariance matrix, scale cuts, and priors from the cosmic shear analyses of both \( \text{H17} \) and \( \text{T17} \), wherein further modelling details can be found. In these works, the analysis of the cosmic shear signal is distinct in almost every step of the shape measurement and subsequent modelling process. This includes the shape measurement algorithms, shear calibration methods, redshift calibration strategy, theory modelling choices, and cosmological model parametrization. The KiDS data\(^4\) are processed by \text{THELI} (Erben et al. 2013) and \text{ASTRO-WISE} (Begeman et al. 2013; de Jong et al. 2015). Shapes are measured using \text{LENSFIT} (Miller et al. 2013), and photometric redshifts are obtained from PSF-matched photometry and calibrated using external overlapping spectroscopic surveys (see Hildebrandt et al. 2017). Details of the DES Y1 data processing can be found in Zuntz et al. (2017), Drlica-Wagner et al. (2018), and references therein.

Because the analysis choices are fundamentally different between \( \text{H17} \) (their table 3) and \( \text{T17} \) (their table 2), we test conclusions in this work using both analysis configurations. These differences include parameter choices, such as sampling over \( \Omega_m \) versus \( \Omega_b h^2 \), as well as changes to the angular scales used and non-linear matter power spectrum modelling. The impact of baryonic effects, which are modelled in the \( \text{H17} \) analysis configuration, are mitigated in the \( \text{T17} \) analysis configuration by removing additional angular scales from the data vector that may be biased by baryonic effects.

We make two important changes to the \( \text{H17} \) data vector and covariance: (1) we update the measured \( \xi_{\pm} \) to use the weighted mean pair separation as noted in footnote 1 of Joudaki et al. (2018), and (2) we update the covariance matrix as discussed below in Section 3. Rather than use the multiple realizations of the redshift distribution as in \( \text{H17} \), we sample over four photometric redshift bias parameters on the mean redshift of each tomographic bin with Gaussian widths \( \sigma(\Delta z) \times 10^5 = (2.376, 1.380, 0.945, 0.594). \) The priors for these parameters are taken from the error on the mean redshift given in Appendix A3 of \( \text{H17} \). We verify that this last change, as well as that due to sampling over \( \mathcal{H}_0 \) instead of \( \theta_{\text{MC}} \) (100 times the ratio of the sound horizon to the angular diameter distance), results in only a slight difference from the original published \( \text{H17} \) contour in \( S_8 \) and \( \Omega_m \), which does not impact any conclusions in this work.\(^5\)

The results in this work are derived using \text{COSMOSIS}\(^6\) (Zuntz et al. 2015), which uses \text{CAMB} (Lewis, Challinor & Lasenby 2000; Howlett et al. 2012) and has been compared in detail with \text{COSMOLIKE} (Krause & Eifler 2017) in \( \text{K17} \). For the \( \text{H17} \) analysis configuration, we model the non-linear matter power spectrum and baryonic effects using Mead et al. (2015, 2016). In the \( \text{T17} \) analysis configuration, we use \text{HALOFIT} (Smith et al. 2003) with updates from Takahashi et al. (2012). The impact of neutrino mass on the matter power spectrum is implemented in \text{HALOFIT} from Bird, Viel & Haehnelt (2012). We sample the parameter spaces, with dimensionality ranging from 11 to 16, using \text{MULTINEST} (Feroz et al. 2013).

### 3 MODELLING THE COVARIANCE

The fiducial analyses of \( \text{H17} \) and \( \text{T17} \) both use covariances obtained from analytic models (see \( \text{H17}, \text{K17} \) for model details and validation). These analytic covariances consist of a Gaussian part, which combines cosmic variance and shape noise, and a non-Gaussian part, which includes contributions from the connected matter trispectrum as well as supersample covariance (Takada & Hu 2013).

The pure shape noise contribution to the covariance \( \text{Cov}_{\text{SN}} \) is given by (Schneider et al. 2002; Joachimi, Schneider & Eifler 2008)
\[ \text{Cov}_{\text{SN}} \left( \xi_{\pm}(\theta_1), \xi_{\pm}(\theta_2) \right) = \frac{\sigma_{\epsilon 1}^2}{N^{ij}_{\pm}(\theta_1, \Delta \theta_1)} \delta_{\theta_1 \theta_2} \left( \delta_{ij} \delta_{jj} + \delta_{ii} \delta_{kj} \right), \]
where \( i, j, k, l \) are tomographic bin indices, \( \sigma_{\epsilon} \) is the ellipticity dispersion, \( N^{ii}_{\pm}(\theta, \Delta \theta) \) is the number of galaxy pairs in angular bin \( \Delta \theta \), and \( \delta \) is the Kronecker delta function. While Schneider et al. (2002) provide an exact expression for \( N_p \) as the weighted sum over all galaxy pairs, this is commonly approximated with \( N^{ii}_{\pm}(\theta) = 2\pi \theta \Delta \theta A_{\text{ns}}^2 n^{g g} \), where \( A_\text{ns} \) is the survey area, \( n^{gg} \) the mean effective projected galaxy density, and \( 2\pi \theta \Delta \theta A_{\text{ns}}^2 n^{g g} \) the approximate area of the annular bin. We will refer to this approximation as the geometric shape noise estimate.

In all cases, we use \( n^{gg} \) and \( \sigma_{\epsilon} \) as defined in Heymans et al. (2013), where the total \( \sigma_{\epsilon} \) is used.

We revisit this common geometric estimate and instead calculate \( N_p \) directly for an infinitesimal annular bin element, taking into account both the survey geometry, as well as the clustered distribution of source galaxies. This calculation starts from the continuous limit of the discrete pair count expression of Schneider et al. (2002),
\[ N^{ii}_{\pm}(\theta, \Delta \theta) = \int_{\Delta \theta} d^2 \theta' \int_{[\theta', \theta + \Delta \theta]} d^2 \theta'' W(\theta') n^{g}(\theta') \times W(\theta'' - \theta') n^{g}(\theta'' - \theta') \]
\[ = 2\pi \theta A_{\text{ns}}^2 n^{g g} \left( W(\theta') d^{g}(\theta') W(\theta'' - \theta') d^{g}(\theta'' - \theta') \right)_{\theta'' = \theta} d \theta'. \]
\[ = 2\pi \theta A_{\text{ns}}^2 n^{g g} \left[ w_w(\theta) \left( 1 + w_{\ell}(\theta) \right) \right] d \theta. \]
Here \( W(\theta) \) is the angular survey mask, normalized such that \( \int_{\Delta \theta} d^2 \theta W(\theta) = A, n^{g}(\theta) \) is the projected galaxy density field, and \( d^{g}(\theta) = 1 + d^{g}_{\text{ns}}(\theta) \), where \( d^{g}_{\text{ns}}(\theta) \) is the projected source galaxy density contrast. \( W \) and \( d^{g} \) have angular correlation functions \( w_w \) and \( w_{\ell} \). The latter is often called source clustering, and we define \( w_w \) to be normalized such that \( w_w(0) = 1, w_w(\theta \rightarrow \infty) = 0 \). We compute \( w_w \) from the power spectrum of the mask, which is evaluated using the \text{HEALPIX}\(^8\) sphere pixelization software package Górski et al. (2005).

In addition to the analytic calculation of \( N_p \), we can also measure components of the shape noise in equation (5) directly from the data. One way to do this is by counting the effective number of pairs of galaxies in the survey footprint, \( N_p(\theta) \), to use in equation (5). This approach still makes use of the measured \( \sigma_{\epsilon} \). The total shape noise contribution \( \text{Cov}_{\text{SN}} \) can also be measured from the variance of \( \xi_{\pm} \) across many random realizations of the shape catalogue (e.g. Gruen et al. 2017; Sánchez et al. 2017; Murata et al. 2018), where \( \xi_{\pm} \) is

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\(^4\)KiDS is an ESO public survey. All data products are available to download from \url{http://kids.strw.leidenuniv.nl/DR3/index.php}.

\(^5\)For example, we reproduce the peak 1D value of \( S_8 \) in the original \( \text{H17} \) chain to 0.03 per cent, while finding a 0.14\( \sigma \) weaker constraint.

\(^6\)\url{https://bitbucket.org/joezuntz/cosmosis/wiki/Home}

\(^7\)\url{http://healpix.sf.net}
measured in each realization after randomly rotating the measured galaxy shapes.

We note that survey geometry also affects the other terms in the analytic covariance: the impact of the KiDS footprint on supersample variance is included in the analytic covariance from H17, and the impact of survey geometries on the cosmic variance and mixed covariance terms is explored in Kilbinger & Schneider (2004) and Joachimi et al. (2008). Singh et al. (2017) also considers the effect of masking on the covariance of several large-scale structure probes. However, in this analysis we focus on the shape noise, which is the leading contribution to the total variance on most angular scales included in the cosmology analysis.

3.1 Impact of survey geometry on shape noise

We demonstrate that the three approaches to measuring Cov\textsuperscript{SN} give consistent results for KiDS-450 in Fig. 1, while they disagree substantially with the geometric estimate. We find that in surveys that are non-compact or composed of multiple small fields, like KiDS-450, measurements of the shape noise that account for survey geometry can be significantly boosted relative to the standard geometric shape noise estimate. This update slightly reduces the impact of shape noise on scales < 2 arcmin due to source clustering, while significantly boosting the measured shape noise on large scales, up to a factor of 3.5 relative to the geometric estimate at 300 arcmin. We also show the impact of the shape noise update to the total covariance in KiDS-450 in the lower panel of Fig. 1.

This is less of an issue for survey geometries that are more contiguous and compact, such as DES Y1, but there is still a non-negligible difference in the shape noise term compared to the geometric estimate. This is shown in the top panel of Fig. 1, which compares the effect on the KiDS-450 and DES Y1 analyses. This difference will diminish as survey footprints become larger and for those that are designed to be contiguous and compact.

3.2 Treatment of multiplicative shear bias uncertainty

In addition to the shape noise update described in Section 3.1, we also modify how the multiplicative shear bias uncertainty σ\textsubscript{m} is added to the covariance, which is given in equation (12) of H17:

\[
\text{Cov}_{\text{cal}} = 4\sigma^2_{m} \xi_{-} + \text{Cov}^{ij}. \tag{7}
\]

This approach of analytic marginalization should be equivalent for a single m value to what was done in T17 (e.g. Bridle et al. 2002) when the predicted (theory) template (\(\bar{\xi}_{+}\)) is used.

In the original H17 covariance, the measured \(\bar{\xi}_{-}\text{,meas.}\) was used as a fixed template to evaluate this term, which can lead to a bias in the inferred amplitude of the signal (e.g. \(S_{h}\)) due to part to there being a larger relative addition to the covariance for the data points that scatter high due to noise than for those that scatter low. The use of the data vector as a template in equation (7) was identified by accident while separating terms in the H17 covariance to update the shape noise component described in the previous section. We agreed that this should be changed if it non-negligibly biased the resulting cosmological inference prior to knowing its precise impact on the inferred parameter values.

After we decided to change the template used in equation (7), we further explored the impact of using various fixed templates for the analytic marginalization of m relative to the formally correct, cosmology-dependent template in equation (7). This included (1) using a fixed predicted \(\bar{\xi}_{+}\) as the template at the cosmology used to evaluate the other covariance terms, (2) using fixed predicted templates at two cosmologies with amplitudes of \(\pm 10\%\) per cent, (3) marginalizing over m by including it as an additional model parameter in the inference, and (4) using the data vector as a fixed template. The first three approaches were found to give consistent results to the cosmology-dependent approach, while using the data as a fixed template biases the inferred \(S_{h}\) low, as shown in Fig. 2.

For consistency, we implement equation (7) using a fixed template at the same cosmology used to evaluate the results of the covariance in the results described below.

4 INTERPRETATION OF COSMIC SHEAR RESULTS

The accuracy of the cosmic shear covariance can significantly impact the interpretation of measurements. We have found in the recent DES Y1 analyses that underestimating the shape noise by 10–
This shift improves agreement in the MNRAS, reduced from 268 to 227 for 211 dof, with DES Y1 and in the parameter space of T17, the best-fitting 122 to 78 for 67 dof. The interpretation of the DES cosmic shear we find significant improvements to the $\Lambda$CDM best-fitting model parameters. In the original parameter space of H17, increasing in the best-fitting the KiDS-450 cosmic shear data due to the shape noise update. The best-fitting $\chi^2$ is reduced from 161 to 121 for 118 dof, corresponding to an increase in the $p$-value from $5 \times 10^{-3}$ to 0.4. Similarly, in the parameter space of T17, the best-fitting $\chi^2$ is reduced from 122 to 78 for 67 dof. The interpretation of the DES cosmic shear best-fitting $\chi^2$ in this parameter space is similar, with a $\chi^2$ that was reduced from 268 to 227 for 211 dof, with $p = 0.21$.

The update to the way $\sigma_m$ is included in the covariance in equation (7) more strongly impacts the inferred cosmology, while not significantly modifying the $\chi^2$. This is shown in the left-hand panels of Fig. 2 combined with the shape noise update (solid contour), along with the impact of updating the reported $\theta$ value in the data vector (dotted), both relative to the analysis of the original data vector and covariance (dashed). Both of these updates increase the inferred $S_8$. We find a similar shift in $S_8$ in the $\Omega_m$ parameter space. This shift improves agreement in the $S_8-\Omega_m$ plane compared to both DES Y1 (T17) and Planck (TT+lowP; Ade et al. 2016) results. However, the complete overlap of the KiDS and DES constraints found here is likely coincidental and not necessarily expected statistically.

We compare the final parameter constraints from KiDS-450 and DES Y1 in the right-hand panels of Fig. 2, finding complete overlap of the KiDS-450 and DES Y1 cosmic shear contours in $S_8$ and $\Omega_m$, with constraints of $S_8 = 0.782^{+0.027}_{-0.031}$ for DES Y1 and $S_8 = 0.772^{+0.037}_{-0.031}$ for KiDS-450 in the T17 analysis configuration.

Beyond the primary cosmological parameters, it is also important to recognize (as recently highlighted in Efstathiou & Lemos 2018) the impact that the major astrophysical systematic in cosmic shear, the intrinsic alignment of galaxies (IA) (see Joachimi et al. 2015; Troxel & Ishak 2015, and references therein), can have on the interpretation of cosmological results. One diagnostic of potential residual systematics is an inconsistent model fit for the IA signal, up to a potential difference in the effective amplitude due to the use of different shape measurement methods. We also find excellent agreement here, with an amplitude for the intrinsic alignment model of $A_{IA} = 1.0^{+0.4}_{-0.5}$ (DES Y1) and $A_{IA} = 0.9^{+0.5}_{-0.6}$ (KiDS-450) in the T17 analysis configuration, marginalizing over a free redshift power-law evolution that is also strongly consistent. This is a powerful demonstration of consistency between the cosmic shear analyses of these two surveys, which lends credence to the robustness of constraints shown here from cosmic shear.

### 5 CONCLUSIONS AND OUTLOOK

We have demonstrated that using an exact measurement [e.g. the actual $N_0(\theta)$] of the shape noise component of analytic cosmic shear covariance matrix estimates is critical for ongoing and future analyses where the survey footprint is non-compact or disjoint. In the case of KiDS-450, we have demonstrated that this correction increases the shape noise term in the covariance by up to a factor of 3.5 on the largest scales. This shape noise correction is sufficient to completely resolve the large best-fitting reduced $\chi^2$ for $\Lambda$CDM from the original analysis of H17, and the first preprint version of T17. With these updates, there is no longer any evidence for a lack of internal model consistency in this basic test for these cosmic shear analyses. We find that these changes can also relieve previously discussed tensions in other internal consistency tests, such as those.
performed in Efstathiou & Lemos (2018). For the six subsets of the data considered in that work, we find that all subsets are now consistent using the statistical methods described therein.

We find that two additional updates in (1) the addition of $\sigma_m$ to the covariance matrix described in Section 3.2 and (2) the determination of the effective angular values for the data vector both shift the inferred $S_8$ from KiDS-450 to slightly larger values. This improves the mutual consistency in cosmological constraints between the KiDS-450 and DES Y1 cosmic shear data sets found in T17, while also bringing the KiDS-450 and Planck results into better agreement in the $S_8-\Omega_m$ plane. These results are an important step forward in the mutual validation of cosmic shear results. A more complete comparison of the DES and KiDS weak lensing results and a full investigation of the impact of survey geometry on the mixed and cosmic variance covariance terms is warranted and is left to future work. An extended study of the internal and mutual consistency between several existing weak lensing surveys, including KiDS-450, will be presented in Chang et al. (in preparation).

Our results weaken evidence that $\Lambda$CDM cannot consistently describe both low-redshift cosmic shear and the CMB, given the agreement shown here between DES Y1, KiDS-450, and Planck. With the next releases of DES, HSC, and KiDS weak lensing results and CMB results from Planck, $\Lambda$CDM will face a much stronger test. These upcoming results will determine whether the current agreement converges further, or whether we begin to see evidence of new fundamental physics needed to describe the evolution of the Universe from the surface of last scattering to the low redshifts probed by weak lensing.

ACKNOWLEDGEMENTS

We thank O. Doré, G. Efstathiou, C. Heymans, H. Hildebrandt, C. Hirata, B. Joachimi, S. Joudaki, and K. Kuijken for comments, and the KiDS Collaboration for providing their weak lensing data to the community. Part of this research was supported from NASA grant 15-WFIRST15-0008 Cosmology with the High Latitude Survey WFIRST Science Investigation Team. Part of the research was carried out at the Jet Propulsion Laboratory, California Institute of Technology, under a contract with the National Aeronautics and Space Administration and is supported by NASA ROSES ATP 16-ATP16-0084 grant and by NASA ROSES 16-ADAP16-0116. PL acknowledges support from an Isaac Newton Studentship at the University of Cambridge and from the Science and Technologies Facilities Council.

EK thanks the Centro de Ciencias Pedro Pascual in Benasque, Spain for hospitality during the workshop Understanding cosmological observations, where some of this work originated. Funding for the DES Projects has been provided by the DOE and NSF(USA), MEC/MICINN/MINECO(Spain), STFC(UK), HEFCE(UK), NCSA(UIUC), KICP(U. Chicago), CCAPP(Ohio State), MIFPA(Texas A&M), CNPQ, FAPERJ, FINEP(Brazil), DFG(Germany) and the Collaborating Institutions in the DES.

The Collaborating Institutions are Argonne Lab, UC Santa Cruz, University of Cambridge, CIEMAT-Madrid, University of Chicago, University College London, DES-Brazil Consortium, University of Edinburgh, ETH Zürich, Fermilab, University of Illinois, ICE (IEEC-CSIC), IFAE Barcelona, Lawrence Berkeley Lab, LMU München, and the associated Excellence Cluster Universe, University of Michigan, NOAO, University of Nottingham, Ohio State University, University of Pennsylvania, University of Portsmouth, SLAC National Lab, Stanford University, University of Sussex, Texas A&M University, and the OzDES Membership Consortium.

This work is based in part on observations at Cerro Tololo Inter-American Observatory, National Optical Astronomy Observatory, which is operated by the Association of Universities for Research in Astronomy (AURA) under a cooperative agreement with the National Science Foundation.

The DES Data Management System is supported by the NSF under Grant numbers AST-1138766 and AST-1536171. The DES participants from Spanish institutions are partially supported by MINECO under grants AYA2015-71825, ESP2015-66861, FPA2015-68048, SEV-2016-0588, SEV-2016-0597, and MDM-2015-0509, some of which include ERDF funds from the European Union. IFAE is partially funded by the CERCA program of the Generalitat de Catalunya. Research leading to these results has received funding from the European Research Council under the European Union’s Seventh Framework Program (FP7/2007-2013) including ERC grant agreements 240672, 291329, and 306478. We acknowledge support from the Australian Research Council Centre of Excellence for All-sky Astrophysics (CAASTRO), through project number CE110001020, and the Brazilian Instituto Nacional de Ciências Tecnologia (INCT) e-Universes (CNPq grant 465376/2014-2).

This manuscript has been authored by Fermi Research Alliance, LLC under Contract No. DE-AC02-07CH11359 with the U.S. Department of Energy, Office of Science, Office of High Energy Physics. The United States Government retains and the publisher, by accepting the article for publication, acknowledges that the United States Government retains a non-exclusive, paid-up, irrevocable, world-wide license to publish or reproduce the published form of this manuscript, or allow others to do so, for United States Government purposes.

Based on data products from observations made with ESO Telescopes at the La Silla Paranal Observatory under programme IDs 177.A-3016, 177.A-3017, and 177.A-3018.

This research used resources of the National Energy Research Scientific Computing Center, a DOE Office of Science User Facility supported by the Office of Science of the U.S. Department of Energy under Contract No. DE-AC02-05CH11231. This work also used resources at the Ohio Supercomputing Center (OSC 1987). Some of the results in this paper have been derived using the healpix (Górski et al. 2005) package.

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