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The Atacama Cosmology Telescope: Detection of the Pairwise Kinematic Sunvaev-Zel'dovich Effect with SDSS DR15 Galaxies

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We present a 5.4σ detection of the pairwise kinematic Sunyaev-Zeldovich (kSZ) effect using Atacama Cosmology Telescope (ACT) and *Planck* CMB observations in combination with Luminous Red Galaxy samples from the Sloan Digital Sky Survey (SDSS) DR15 catalog. Results are obtained using three ACT CMB maps: co-added 150 GHz and 98 GHz maps, combining observations from 2008-2018 (ACT DR5), which overlap with SDSS DR15 over 3,700 sq. deg., and a componentseparated map using night-time only observations from 2014-2015 (ACT DR4), overlapping with SDSS DR15 over 2,089 sq. deg. Comparisons of the results from these three maps provide consistency checks in relation to potential frequency-dependent foreground contamination. A total of 343,647 galaxies are used as tracers to identify and locate galaxy groups and clusters from which the kSZ signal is extracted using aperture photometry. We consider the impact of various aperture photometry assumptions and covariance estimation methods on the signal extraction. Theoretical predictions of the pairwise velocities are used to obtain best-fit, mass-averaged, optical depth estimates for each of five luminosity-selected tracer samples. A comparison of the kSZ-derived optical depth measurements obtained here to those derived from the thermal SZ effect for the same sample is presented in a companion paper.

I. INTRODUCTION

Deciphering the origins of accelerated cosmic expansion [1, 2] is one of the central goals of modern cosmology. The effects of dark energy only manifest indirectly, through possible deviations from the predictions of General Relativity (GR) and the gravitational properties of Standard Model particles and dark matter. To determine if dark energy is a cosmological constant, a novel type of matter, or evidence that gravity deviates from GR on cosmic scales, one is principally reliant on three cosmological tracers of the gravitational field: the positions and velocities of massive objects and the distortion they create in the geodesic paths of light from more distant objects.

As Cosmic Microwave Background (CMB) photons traverse through a galaxy cluster they interact with the hot cluster gas, and the peculiar motion of the cluster relative to the CMB rest-frame creates a Doppler-shift in the CMB known as the kinematic Sunyaev-Zel'dovich effect (kSZ)[3]. Concurrently with the kSZ, the CMB photons are also heated up by the cluster gas, the thermal Sunyaev-Zel'dovich effect (tSZ). The tSZ imprint has a characteristic frequency dependence, and can be isolated through the use of multi-frequency measurements. By contrast, the kSZ effect is an order of magnitude smaller and has a thermal spectrum that makes its detection, and separation from tSZ and dust emission foregrounds, challenging.

On scales of the order of $\sim 25-50$ Mpc, the gravitational attraction between clusters (and groups) of galaxies causes them, on average, to move towards each other. This pairwise motion can be used to extract the kSZ effect. A pairwise correlation statistic [4] is a useful approach to extracting kSZ signals because of its dependence on differences of measured temperatures on the sky at the positions of clusters, averaging out contaminating signals like the tSZ signal and dust emission. The pairwise kSZ momentum, sensitive to both the cluster peculiar velocity and optical depth, has been shown to have the potential to probe the large scale structure (LSS) growth rate, providing insights into the evolution of dark energy, cosmic modifications to gravity over cosmic time, and constraints on the sum of the neutrino masses [5-11].

Extraction of the kSZ signal is aided by using galaxy surveys to provide bright tracer galaxies to identify and locate the clusters [12–16]. The first measurement of the kSZ signal was made by Hand et al. [17] (herein H12) by estimating the mean pairwise cluster momentum with the ACT data from 2008 to 2010 observing seasons and a sample of clusters traced by galaxies in the Sloan Digital Sky Survey Data Release 9 (SDSS DR9) galaxy catalog. This measurement has since been improved with a 4.1σ measurement in the mass-averaged optical depth, $\bar{\tau}$, using improved data from ACT DR3 and SDSS DR11 data [18] (herein DB17). Detections using this estimator have also been reported by the *Planck* collaboration using galaxies from SDSS [19], and the South Pole Telescope collaboration using galaxies from the Dark Energy Survey [20].

In addition to the pairwise statistics, other complementary techniques have also been applied to measure the kSZ effect [21], including velocity reconstruction [22], projected fields [23, 24], cross-correlation of angular redshift fluctuations [25] and cluster stacking [26]. Two recent papers [27, 28] focused on using velocity reconstruction and stacking of galaxy cluster samples to study the radial profiles of tSZ and kSZ signals in the ACT data. The work used the same coadded [29] and componentseparated [30] maps, as are used here, but the galaxy samples are different, with different host halo masses. As a result, the findings from these papers are not directly comparable to those in this work, nor those in the companion paper [31] (V21). We find, however, that the rough signal-to-noise ratios are comparable. Overall, these two sets of papers provide complementary ways to analyze tSZ and kSZ effects.

Our work is laid out as follows: In Section II, we describe the ACT and *Planck* CMB data and the SDSS galaxy samples used in our analysis. In Section III, we lay out the formalism for the pairwise estimator, the covariance techniques, mass-averaged optical depth fitting and signal-to-noise estimation. In Section IV, we present our results and discuss the pairwise kSZ detections and mass-averaged optical depth constraints. The findings are drawn together in the Conclusion, in Section V.

II. DATASETS

II.A. ACT data

In our analysis, we use three CMB datasets that combine ACT and *Planck* data. The first dataset is a component-separated internal linear combination map (ILC) [30], referred to as DR4 ILC, which uses nighttime ACT observations from DR4, principally from 2014 to 2015 [32, 33] as well as *Planck* data in eight bands, from 30 to 545 GHz, from the PR2 (2015) release [34]. The map is created by minimizing the variance and is dominated by CMB and kSZ signals but also has other foregrounds including thermal SZ and Cosmic Infrared Background (CIB) contributions.

The second and third datasets are the co-added ACT DR5 98 GHz and 150 GHz maps [29] which combine ACT observations from 2008-2018 seasons, including day-time data, and *Planck* PR2 [35] data release centered at 100 and 143 GHz. We refer to these two maps as DR5 f090 and DR5 f150, respectively, using the frequency naming conventions in Naess et al. [29]. The CMB maps have point source and galactic plane masks, and a noise threshold cut of 45 μ K, relative to the CMB, as discussed in more detail in V21. The companion paper also includes a map of the specific regions utilized for DR4 ILC, DR5 f150 and DR5 f090 CMB and SDSS surveys.

The use of co-added maps at two different frequencies and the multi-frequency component separated map facilitates the comparison of the extracted kSZ measurements from maps in which potential thermal SZ and other secondary foreground contributions will vary.

II.B. SDSS data

As in the previous ACT analyses, H12 and DB17, we utilize spectroscopically-selected luminous red galaxies (LRGs) as tracers of the galaxy groups and clusters in which the kSZ is to be measured. Galaxies from SDSS DR15 [36] are identified in the regions overlapping with the CMB maps, for DR5 f150 and DR5 f090 maps the overlap is 3,700 sq. deg. while for the DR4 ILC map the overlap is 2,089 sq. deg. Galaxies are selected based on multiband de-redenned model magnitudes. Full details of the selection process, including the SDSS query, are provided in V21. The query yields 602,461 galaxies for which the luminosities are calculated from the model magnitude using K-corrections made with the $k_correct^1$ software [37] using the SDSS asinh magnitude conversion [38].

Samples are selected based on luminosities, with minimum thresholds 4.3, 6.1 and $7.9 \times 10^{10} L_{\odot}$, with the latter two chosen to align with selection criteria in DB17. Additional galaxy cuts are applied based on ACT CMB map noise, point source excision, and removal of the galactic plane. The final catalog analyzed includes 343,647 galaxies with $L > 4.3 \times 10^{10} L_{\odot}$.

We consider the kSZ properties in five luminosity bins: two disjoint luminosity bins, $4.3 < L(10^{10}L_{\odot}) < 6.1$, $6.1 < L(10^{10}L_{\odot}) < 7.9$ (referred to as L43D and L61D respectively), two cumulative luminosity bins, $L(10^{10}L_{\odot}) >$ 4.3 and > 6.1 (L43 and L61, respectively) and one bin that is both disjoint and cumulative, $L(10^{10}L_{\odot}) > 7.9$ (referred to as L79). The characteristics of the samples in each luminosity-selected bin are summarized in Table I, including the host group/cluster mass ranges, the mean redshift and luminosities, and the number of galaxies included in each bin. The full redshift distributions of the galaxy samples are shown in the companion paper, V21.

III. FORMALISM

III.A. Pairwise momentum estimator

The CMB temperature shift induced by the peculiar motion of a galaxy group/cluster is given by [39],

$$\frac{\delta T_{\rm kSZ}}{T_0}(\hat{\mathbf{r}}) = -\int dl \,\sigma_T \, n_e \frac{\mathbf{v} \cdot \hat{\mathbf{r}}}{c} \tag{1}$$

where n_e is the electron number density, $T_0 = 2.726K$ is the average CMB temperature and σ_T is the Thomson cross-section. A positive peculiar velocity, **v**, relates to motion away from the observer, so induces a negative kSZ effect.

The temperature is obtained through aperture photometry (AP), in which the temperatures of pixels within a disk of aperture size Θ and an annulus of equal area, out to radius $\sqrt{2}\Theta$, are differenced around each group/cluster. We use the positions of the tracer LRGs to center the aperture, under the assumption of the Central Galaxy Paradigm [40] that the brightest galaxy within a group/cluster traces the minimum of the gravitational potential well.

The aperture temperature is calculated by analyzing a postage stamp region centered at the angular position of the i^{th} galaxy, $\mathbf{r}_i = \{\hat{\mathbf{r}}_i, z_i\}$ that includes, but extends beyond, the group/cluster in question. Within the postage stamp a finer resolution pixel grid is created, 10 times smaller than the pixel size; temperatures are assigned to the finer pixels using a Fourier domain interpolation. The average temperatures of the smaller pixels contained in the disk/ring are then used to calculate the $T_{AP}(r_i, z_i, \Theta) = \overline{T}_{\text{disk}} - \overline{T}_{\text{annulus}}$. We checked that this gives an equivalent result to taking weighted averages of the full size pixels when a fractional weighting equivalent to the area of each pixel within the disk or annulus is included.

¹ http://kcorrect.org

			DR5 f1	50 and DI	R5 f090	DR4 ILC			
Bin	Luminosity cut	Mass cut M_{200}	Ν.	$\langle L \rangle$	/~\	Ν.	$\langle L \rangle$	/~\	
Label	$(10^{10}L_{\odot})$	$(10^{13}M_{\odot})$	¹ vgal	$(10^{10}L_{\odot})$	\~/	1 vgal	$(10^{10}L_{\odot})$	\2/	
L43D	4.3 < L < 6.1	0.55 < M < 1.00	130,577	5.2	0.48	$71,\!699$	5.2	0.48	
L61D	6.1 < L < 7.9	1.00 < M < 1.66	109,911	6.9	0.48	$61,\!024$	6.9	0.48	
L43	L > 4.3	M > 0.52	$343,\!647$	7.4	0.49	$190,\!551$	7.4	0.50	
L61	L > 6.1	M > 1.00	$213,\!070$	8.7	0.51	$118,\!852$	8.7	0.51	
L79	L > 7.9	M > 1.66	$103,\!159$	10.6	0.53	$57,\!828$	10.9	0.54	

TABLE I: Summaries of the five luminosity-determined samples analyzed in this paper along with the bin labels with which we will refer to them throughout. The host halo mass ranges, M_{200} , the number of galaxies, N_{gal} , the mean redshift, $\langle z \rangle$, and mean luminosity, $\langle L \rangle$, are given for the samples that overlap with the DR5 f150, DR5 f090 and DR4 ILC maps. These galaxy selection and halo mass estimates are derived in the companion paper V21.

We use an aperture size of $\Theta = 2.1'$, aligned with the anticipated angular size of groups/clusters in the redshift ranges we are analyzing (2.1' at z = 0.5 relates to a comoving scale of ~1.1Mpc for the cosmological model assumed in the analysis, as described in Sec. III.B).

We estimate the kSZ temperature by calculating the temperature decrement around each tracer galaxy,

$$\delta T_i(r_i, z_i, \sigma_z, \Theta) = T_{AP}(r_i, z_i, \Theta) - \bar{T}_{AP}(r_i, z_i, \Theta, \sigma_z),$$
(2)

where, following H12 and DB17, we subtract a redshiftsmoothed aperture temperature, \bar{T}_{AP} , to remove potential redshift dependent contamination that could mirror a pairwise signal when differencing aperture temperatures from objects separated in redshift. A Gaussian smoothing is applied for each pair using a redshift smoothing parameter, σ_z :

$$\bar{T}_{AP}(r_i, z_i, \Theta, \sigma_z) = \frac{\sum_j T_{AP}(r_i, \Theta) \exp\left(-\frac{(z_i - z_j)^2}{2\sigma_z^2}\right)}{\sum_j \exp\left(-\frac{(z_i - z_j)^2}{2\sigma_z^2}\right)} (3)$$

We use $\sigma_z = 0.01$, as used by the *Planck* team [19] and in DB17. We demonstrate that the pairwise results are insensitive to the precise value of σ_z in Appendix A.

Analogous with the corrections made to the tSZ temperatures in V21, we correct for the difference in kSZ aperture temperature due to the differences in beam size between the DR5 f090 map (FWHM=2.1') and the DR4 ILC map (FWHM=1.6') versus the DR5 f150 map (FWHM=1.3')[29]. We consider a fiducial kSZ density profile for the average virial group/cluster in each luminosity bin and derive estimates of the kSZ signal when convolved with the respective beams for the DR5 f150 and DR5 f090 maps from [29], using Mop-c GT ², and using a Gaussian beam for the DR4 ILC map (see a detailed description in [28]). A resulting relative beam correction factor is applied that increases the DR5 f090 map T_{AP} measurements by 31% and reduces the DR4 ILC map by 5%. We implement the pairwise momentum estimator [14] for the correlation of the velocities,

$$\hat{p}(r) = -\frac{\sum_{i < j} (\delta T_i - \delta T_j) c_{ij}}{\sum_{i < j} c_{ij}^2}, \qquad (4)$$

where the sum is over all pairs, each separated by a distance $r = |\mathbf{r}_{ij}| = |\mathbf{r}_i - \mathbf{r}_j|$. The weights c_{ij} are geometrical factors that account for the alignment of a pairs *i* and *j* along the line of sight [18], given by

$$c_{ij} = \hat{\mathbf{r}}_{ij} \cdot \frac{\hat{\mathbf{r}}_i + \hat{\mathbf{r}}_j}{2} = \frac{(r_i - r_j)(1 + \cos\alpha)}{2\sqrt{r_i^2 + r_j^2 - 2r_i r_j \cos\alpha}}$$
(5)

where α is the angle between unit vectors \hat{r}_i and \hat{r}_j .

We analyze data in radial separation bins of width 10 Mpc centered on r = 5 up to 145 Mpc and then four unevenly spaced bins, centered on 175, 225, 282.5 and 355 Mpc (for which the maximum included separation is 395 Mpc). The latter bins have broader widths to account for increased correlation between spatial scales as one goes to larger separations, as was found in DB17, and discussed in Appendix B.

We update the kSZ pipelines used in [41], which analyzed *Planck* SEVEM maps in HEALPix format, and in DB17. The pipeline used in this work is publicly available ³ and parallelized and distributed [42, 43] in Python, and uses $Pixell^4$ subroutines to analyze the CMB map.

Our aperture photometry assumptions include some analytic differences from those in DB17: we include fractional pixel weighting, reproject the pixelation of the submap, and implement cluster-centered, instead of pixel-centered aperture photometry. We discuss these differences, and their respective implications for the signal extraction in Appendix A.

To estimate the covariance, at least four resampling strategies have been proposed in the literature, with error bar estimates that vary up a factor of two among them

² https://github.com/samodeo/Mop-c-GT

³ https://github.com/patogallardo/iskay

⁴ https://github.com/simonsobs/pixell

(see Appendix in [20] for more detail). As shown in [20] there are systematic differences in the inference method that tend to dominate the uncertainty estimation.

In cross-correlating the maps with the galaxy sample, contributions will be picked up from the residual foregrounds that are correlated with the tracer. In kSZ estimators, which are velocity weighted, the positive and negative contributions from cross-correlation with the velocity field are not mimicked by other residual foreground contributions and their effect is suppressed effectively contributing to noise, and not bias, in the pairwise signal. It is important to capture this noise contribution (for example from the residual tSZ) in the covariance calculation. Estimators that sample the maps directly will capture this more effectively than simulations. In this work we use bootstrap estimation from the maps directly to evaluate the covariance used in the analysis. In Appendix B, we summarize the findings of the covariance estimation comparison across three different methods, covariances of simulated maps, jackknife (JK) estimation, the primary method used in [18, 20], and bootstrap estimation, and motivate why we use the bootstrap derived estimates in the main analysis.

III.B. Signal to Noise and $\bar{\tau}$ estimates

A theoretical prediction for the observed pairwise momentum can be modeled in terms of a mass-averaged pairwise peculiar velocity, V

$$\hat{p}_{\rm th}(r,z) = -\frac{T_{CMB}}{c}\bar{\tau}V(r,z),\tag{6}$$

where $\bar{\tau}$ is an effective mass-averaged measure of the optical depth over the group/cluster samples and z is taken as the mean redshift of each luminosity sample as given in Table I.

The theoretical pairwise velocity, V, can be derived in terms of the correlation function [44], and is calculated here following [9, 10] using linear theory [45],

$$V(r,z) = -\frac{2}{3} \frac{f(z)H(z)r}{1+z} \frac{\xi_h(r,z)}{1+\xi_h(r,z)},$$
 (7)

where f(z) is the linear growth rate and H(z) is the Hubble rate. ξ_h and $\overline{\xi}_h$ are, respectively, the 2-point halo correlation function and volume averaged halo correlation function:

$$\xi_h(r,z) = \frac{1}{2\pi^2} \int dk k^2 j_0(kr) P(k,z) b_h^{(2)}(k), \qquad (8)$$

$$\bar{\xi}_h(r,z) = \frac{3}{r^3} \int_0^r dr' r'^2 \xi(r,z) b_h^{(1)}(k).$$
(9)

Here P(k, z) is the linear matter power spectrum, $j_0(x) = \sin(x)/x$ is the zeroth order spherical Bessel function, and

 $b_h^{(q)}$, are mass-averaged halo bias moments given by,

$$b_{h}^{(q)}(z) = \frac{\int_{M_{\min}}^{M_{\max}} dM \ M \ n(M,z) b^{q}(M) W^{2}[kR(M,z)]}{\int_{M_{\min}}^{M_{\max}} dM \ M \ n(M,z) W^{2}[kR(M,z)]}$$
(10)

with n(M, z) is number density of halos of mass M, for which we use a halo mass function in [46], and the top-hat window function is given by $W(x) = 3(\sin x - x \cos x)/x^3$. R is the characteristic scale of a halo of mass M, $R(M, z) = [3M/4\pi\bar{\rho}(z)]^{1/3}$, with $\bar{\rho}$ the background cosmological matter density. The lower mass limit, M_{\min} , is taken to be the halo mass cut given in Table I. The upper mass limit, M_{\max} , is taken to be $10^{16} M_{\odot}$. We consider in the analysis the sensitivity of the results to these specific limits, with the understanding that bias the moments are dominated by the, far more numerous, lower mass halos.

We compare our pairwise kSZ momentum measurements to theoretical peculiar velocity predictions using a modified version of the CAMB code [47] that calculates the mass-averaged pairwise velocity, V, as described in Mueller et al.[9, 10]. We assume a *Planck* cosmology for a flat universe [48]: $\Omega_b h^2 = 0.02225$, $\Omega_c h^2 = 0.1198$, $H_0 = 67.3 \text{kms}^{-1} \text{Mpc}^{-1}$, $\sigma_8 = 0.83$, $n_s = 0.964$. We translate the observational galaxy luminosity cuts to group/cluster mass cuts using mass-luminosity relationship described in V21.

We determine the likelihood of $\bar{\tau}$ using the χ^2 ,

$$\chi^2(\bar{\tau}) = \sum_{ij} \Delta \hat{p}_i(\bar{\tau}) \hat{C}_{ij}^{-1} \Delta \hat{p}_j(\bar{\tau}), \qquad (11)$$

with $\Delta \hat{p}_i(\bar{\tau}) = \hat{p}_{i,\text{th}}(\bar{\tau}) - \hat{p}_{i,\text{obs}}$, where $\hat{p}_{i,th}(\bar{\tau})$ is the theoretical kSZ pairwise momentum estimate at cluster separation r_i for an assumed mass-averaged optical depth, $\bar{\tau}$, and \hat{p}_i^{obs} are the measurements obtained from the ACT and SDSS data.

For the best-fit model, with $\chi^2=\chi^2_{\rm min}$, we calculate the Probability-To-Exceed (PTE), the probability of obtaining a higher χ^2 value,

$$PTE = \int_{\chi^2_{\min}}^{\infty} \chi^2_m(x) dx, \qquad (12)$$

where χ_m^2 is the χ^2 distribution for *m* degrees of freedom [49]. Unlikely events, or those in tension with theory given the experimental uncertainties, are signified by a low PTE. Consistently high PTEs might imply experimental uncertainties have been overestimated.

Finally, we compute the signal-to-noise ratio (SNR), inferred by assuming the signal is given by the best-fit theoretical model,

$$\operatorname{SNR}(\bar{\tau}) = \sqrt{\sum_{ij} \hat{p}_{i,\text{th}}(\bar{\tau}) \hat{C}_{ij}^{-1} \hat{p}_{j,\text{th}}(\bar{\tau})}.$$
 (13)

IV. ANALYSIS

IV.A. kSZ pairwise momentum results

The kSZ pipeline was tested using two methods respectively employing the DR4 ILC noise simulations and the DR5 f150 map. Firstly, we use the DR4 ILC noise simulations to apply the aperture photometry extraction, and compute the average pairwise kSZ signal and sample covariance over the 560 realizations. The second approach, testing the pipeline on the DR5 f150, calculates the aperture temperature decrements for all galaxies in each luminosity-based tracer sample and then shuffles them while keeping the sky positions of the galaxies and redshifts fixed. The average and covariance of the resulting signals are calculated over 1,000 realizations. Fig. 1 shows the results of these two tests, which both show the expected effect, that the null tests remove the pairwise kSZ signal and leave a signal around zero with correlated uncertainties encapsulated in the covariance.

In Fig. 2 we present the pairwise kSZ measurements for DR4 ILC, DR5 f150 and DR5 f090 CMB maps with the SDSS DR15 luminosity-selected galaxy tracer samples along with the uncertainty estimates obtained from the bootstrap resampling of the same maps.

The L43, L61 and L79 samples show a pairwise momentum profile with a negative signal amplitude reaching a maximum at separations around 25-50 Mpc. The negative amplitude is indicative of a gravitational infall between the cluster pairs, with the mutual gravitational attraction falling off as one moves to cluster pairs separated by larger distances. The magnitude of the signal amplitude increases as the average luminosity of the sample increases, consistent with the observed clusters being more massive halos with larger optical depths, and with deeper gravitational potentials. At small scales, for r < 20 Mpc, the pairwise velocity correlation function has nonlinear contributions, and becomes positive, rather than negative as predicted by linear theory [50-52]. In the L43D and L61D disjoint bins, the kSZ signal is less discernible, consistent with an expectation that the signal in these groups/clusters should be smaller since we expect their masses to be lower (coupled to the lower luminosities of the tracer galaxies), while the uncertainties, driven by their sample size, should be comparable to those in L79. We note that the three largest separation bins in the DR4 L79 data are positive, however these points are highly correlated (as shown in Fig. 10 in the Appendix) so that the deviation from null is not significantly anomalous.

At each luminosity-selected galaxy tracer sample, the respective pairs of kSZ signals in Fig. 2 extracted from the three complementary maps show consistency within the 1σ error bars. Given that each has a different approach to removing foreground emission, the consistency indicates that the results are robust against significant individual, distinct contamination from frequency dependent foregrounds. One concern that could arise is a po-



FIG. 1: Two null tests conducted on the pipeline for each galaxy tracer luminosity sample: [Upper] The mean and standard deviation of the pairwise estimator derived from the average sample covariance for the DR4 ILC 560 noise simulations. [Lower] A null test applied on the DR5 f150 map. Aperture photometries are taken from our science dataset for all luminosity bins, however the temperature decrement values are randomly shuffled which removes the pairwise signal. Error bars show one sigma uncertainty inferred with bootstrapping temperature decrements. The χ^2 and the probabilities to exceed it are also given for each test.

tential impact of residual thermal SZ contamination in the signal especially from the most massive, luminous clusters. To address this we undertook two additional analyses. First, we compared the signal and covariance estimation for the DR5 f150 L79 sample, which has no upper luminosity limit, with that from a subsample with an upper luminosity threshold of $L < 10^{11} L_{\odot}$ imposed. We find that there is no significant difference in the covariance estimates and no bias in the signal and the variation in the signal is at the level of a fraction of a standard deviation. Second, we analyzed a DR4 ILC map in which the tSZ signal has been deprojected [30] and compare it to the DR4 ILC map analyzed in this paper. We find no evidence of a bias in signal from the different tSZ treatments, and only that the noise is larger in the deprojected map.



FIG. 2: Pairwise velocity correlations for the DR4 ILC [black circle] DR5 f150 [blue cross] and DR5 f090 [orange square] maps for sources in the five luminosity-selected galaxy tracer samples: [from top to bottom] L43D, L61D, L43, L61 and L79. Error bars show 1σ bootstrap uncertainties.

IV.B. Comparison with theoretical pairwise velocity predictions

tions, given in (7), for a *Planck* cosmology using the code developed in [9, 10]. Using (6) we infer an effec-

We compare the observed pairwise correlations with theoretical linear pairwise velocity correlation predic-

Tracer		DR4 II	с		DR5 f090				DR5 f150			
sample	$\bar{\tau} (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR	$\bar{\tau} (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR	$\bar{\tau} (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR
L43D	0.18 ± 0.32	14	0.67	0.5	0.83 ± 0.34	12	0.81	2.2	0.46 ± 0.24	21	0.24	1.7
L61D	0.69 ± 0.34	25	0.08	1.8	1.07 ± 0.35	15	0.59	2.7	0.72 ± 0.26	11	0.85	2.5
L43	0.47 ± 0.12	22	0.20	3.6	0.65 ± 0.13	13	0.71	4.5	0.54 ± 0.09	17	0.42	5.1
L61	0.74 ± 0.15	18	0.40	4.4	0.82 ± 0.17	16	0.53	4.4	0.69 ± 0.11	10	0.92	5.4
L79	0.78 ± 0.23	21	0.21	3.0	0.79 ± 0.27	12	0.79	2.6	0.88 ± 0.18	13	0.76	4.6

TABLE II: The best-fit $\bar{\tau}$ estimates and 1σ uncertainties for the DR4 ILC [left], DR5 f090 [center] and DR5 f150 [right] maps for the five luminosity-selected galaxy tracer samples using the bootstrap uncertainty estimates. The corresponding χ^2 (for 17 degrees of freedom), SNR and PTE values are also given in each scenario.



FIG. 3: The normalized likelihood for $\bar{\tau}$ estimates for the DR5 f150 [blue, full], DR5 f090 [orange, dashed] and DR4 ILC [black, dotted] maps for each of the five luminosity-derived tracer samples using the boostrap uncertainties.

tive measure of the cluster optical depth, $\bar{\tau}$, for the observed samples using 17 spatial separation bins spanning 20 Mpc< r <395 Mpc. We exclude r < 20 Mpc as at these scales nonlinear velocity effects become significant that are not incorporated in the linear theoretical fit.

In Table II we compare the optical depth constraints for the DR4 ILC, DR5 f150 and DR5 f090 maps, using bootstrap estimates as directly obtained from the maps. The probability that a χ^2 would exceed the minimum, best-fit $\chi^2_{\rm min}$ (PTE) and the signal-to-noise ratio (SNR) for the best-fit scenario are also given.

For each map, the average optical depth obtained in each sample increases with increasing mean host halo mass in each bin (and the luminosity of the LRG used as the group/cluster center tracer). The uncertainties in the optical depth measurements increase in tandem with the signal uncertainties, principally driven by the number of galaxies in the luminosity bin. In Appendix B, we summarize how the impact on $\bar{\tau}$ fits of using the covariances derived from the JK and bootstrap methods, and from the dispersion in signals obtained from the 560 noise sims.

Our best measured detections of $\bar{\tau}$ are in the DR5 f150 map with the L61 luminosity cut, for which we obtain SNR of 5.4 with the derived best-fit mass-averaged optical depth of $\bar{\tau} = (0.69 \pm 0.11) \times 10^{-4}$ with a χ^2 of 10 for 17 degrees of freedom.

We assess the sensitivity of these results to uncertainties in the assumption about the minimum and maximum halo mass, $M_{\rm min}$ and $M_{\rm max}$ in (10) using the DR5 f150 map. The bin with the highest mass tracer sample, L79, is the sample for which the maximum mass sensitiv-

ity would be most pronounced. For the default maximum mass of $10^{16} M_{\odot}$ we obtain $\bar{\tau} = (0.88 \pm 0.18) \times 10^{-4}$. If we reduce the maximum mass by an order of magnitude, to $10^{15} M_{\odot}$, we find that the change in the $\bar{\tau}$ constraints is small, with no change to the best-fit value, a small reduction to the uncertainties $\bar{\tau} = (0.88 \pm 0.17) \times 10^{-4}$ and a change of 0.1 in the χ^2 . The small impact sensitivity to $M_{\rm max}$ is expected, as this truncates the high mass tail of the halo mass function which only contributes a small fraction of the halos over which the mass averaged optical depth is calculated. By corollary, we anticipate that the minimum mass will have a larger impact since it impacts more of the halos. Halving the minimum mass lowers the halo bias parameters in the mass averaged correlation function, and reduces the amplitude of predicted pairwise velocity signal. This, in turn, requires a larger $\bar{\tau}$ estimate to fit the theoretical velocity prediction to the pairwise momentum data. The variation introduced by a factor of two theoretical uncertainty in the minimum halo mass is found to be subdominant to the experimental uncertainties: for the DR5 f150 L43 sample, using the default minimum mass of $M = 0.52 \times 10^{13} M_{\odot}$ we obtain $\bar{\tau} = (0.54 \pm 0.09) \times 10^{-4}$. For a minimum mass of half that size we find $\bar{\tau} = (0.57 \pm 0.10) \times 10^{-4}$ and when doubled, $\bar{\tau} = (0.50 \pm 0.09) \times 10^{-4}$. The χ^2 fit changes by < 0.1, varying between 17.4 and 17.6, and the SNRs remain unchanged.

In Fig. 3 we present the likelihoods for the effective $\bar{\tau}$ value for these three maps. The figure shows how the best-fit values of $\bar{\tau}$ are consistent across the three maps, and show an increase as the minimum luminosity threshold for the galaxy tracer sample increases, congruent with an increase in the integrated line of sight number density of electrons with halo mass. Again consistent with the decrease in the sample size, the uncertainty in the $\bar{\tau}$ measurement increases when one considers sequentially higher luminosity thresholds.

The consistency in the signals between the 90GHz, 150GHz and those from the component separated map (across which the tSZ contributions vary) indicates that residual tSZ contamination is not significant. Fig 3. shows that the τ estimates obtained from each of the 3 maps for the L79 sample, focused on the most massive clusters with the largest potential residual foregrounds, are very consistent. Similarly, the L61 cumulative bin, which doesn't have an upper luminosity bound, and the L61D bin, which does, show consistent τ estimates, though the L61D results have a greater variance due to the smaller sample size, showing that uncertainties in the high luminosity clusters are not biasing the results.

In Fig. 4, we overlay the observed pairwise correlations for the DR5 f150, DR5 f090, and DR4 ILC maps, with the theoretical models using the best-fit and 1σ constraints on $\bar{\tau}$ and the *Planck* cosmology pairwise velocity predictions. In this work, we present measurements of the optical depth of clusters derived from cross-correlations of pairwise kSZ effect for three co-added maps, DR5 f150 and DR5 f090, utilizing the most recent ACT DR5 data combined with *Planck* PR2 and PR3 data, and a component separated map, DR4 ILC, using ACT DR4 and *Planck* PR2 data. The kSZ signal is obtained by correlating the maps with the SDSS DR15 galaxy catalog using luminous red galaxies as tracers of the group/cluster center in five luminosity cuts with a minimum luminosity threshold of $4.3 \times 10^{10} L_{\odot}$.

V.

We use bootstrap-derived estimates, derived from each map, for the covariance used to derive kSZ estimates of the mass averaged optical depth for the samples. As detailed in Appendix B, we use the bootstrap method after comparing covariance estimates from bootstrap and jackknife methods, and those from variances across noise sims. We find that JK uncertainties systematically overestimate the uncertainties while the bootstrap method more closely aligns with those from the averaged noise sims.

Using the bootstrap uncertainties, the highest SNR is obtained for the L61 tracer sample and DR5 f150 map, with a 5.4 σ measurement for the best-fit theoretical model relative to the null signal, with a mass-averaged optical depth of $\tau = (0.69 \pm 0.11) \times 10^{-4}$.

In this multi-frequency analysis, we find consistent results for the component separated map, DR4 ILC, and the co-added, DR5 f090 and DR5 f150, maps implying a robustness of the signals extracted, and $\bar{\tau}$ fit estimates, to potential frequency dependent contamination.

A number of refinements to the aperture photometry method have been implemented to improve the precision of signal estimation relative to previous work. This includes considering assumptions on pixel size, pixel reprojection, galaxy versus pixel centering of CMB submaps, fractional pixel weighing, and noise-weighting, as outlined in Appendix A. We leave to future work the potential impact of the brightest galaxy being displaced from the cluster center.

In a companion paper, V21, the kSZ $\bar{\tau}$ results obtained here are compared with those derived from thermal SZ measurements from the same data and theoretical predictions based on the cluster baryon content using a Navarro-Frenk-White profile [53].

This analysis paves the way for pairwise kSZ work with this pipeline on upcoming and future data, including upcoming CMB instruments, for example, Simons Observatory [54], SPT-3G [55], CMB-S4 [56] and the FYST telescope [57], in tandem with upcoming spectroscopic and photometric large scale structure surveys including the Dark Energy Spectroscopic Instrument (DESI) [58], the ESA-NASA Euclid Telescope [59], the Vera Rubin Observatory [60, 61] and the Nancy Roman Space Telescope [62, 63].



FIG. 4: The extracted pairwise signal for the DR4 ILC [black, lower] and DR5 f090 [orange, middle] and DR5 f150 [blue, upper] maps for the three cumulative luminosity-selected galaxy tracer samples, L43 [left], L61 [center], L79 [right], overlaid with the theoretical pairwise velocity model using the *Planck* best-fit cosmology corresponding to the best-fit $\bar{\tau}$ value and 1σ boostrap-derived uncertainties.

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Appendix A: Impact of analysis assumptions on kSZ signal extraction

In this section we present a study of the impact of various analysis assumptions made in this paper, as presented in section IV. As part of this we assess the impact of differences in the assumptions in this paper and in the previous ACT kSZ analysis, DB17, in which the first three seasons of ACT data [64] and the first season of ACTpol data [65] were cross-correlated with clusters identified through a color-luminosity selected SDSS DR11 galaxy sample.

We consider the impact of six specific analysis assumptions.

- *Galaxy sample selection:* As outlined in II.B, we obtain the galaxy sample using the SDSS SQL queries and a K-correct code, described in detail in V21.
- Aperture photometry Pixel size and submap precision: We found that two pixel parameters used in the DB17 analysis to determine which pixels are included in the aperture photometry for a given cluster were approximated/rounded: the "CDELT" parameter determining the pixel size and a second



FIG. 5: Comparison of the impact of a constant [red] versus redshift-varying [blue] aperture size in the pairwise momentum estimation, \hat{p} , for the DR4 ILC map and the DR15 L43 sample.

parameter determining which pixel serves as the central pixel for the submap. We found that using the full rather than approximated values led to differences in which pixels are identified in the disc/annulus and a consequent change in the predicted signal.

- Aperture photometry Reprojection: We account for the geometrical projection effects that modify the equal area treatment in aperture photometry when the cluster location is near the poles rather than the equator by using the **Pixell** reprojection subroutine.
- Aperture photometry Pixel vs. tracer galaxycentering: In DB17, the aperture photomety was centered around the center of the pixel in which the tracer galaxy is located. In this analysis, we implement a "galaxy-centered" approach rather than a "pixel-centered" approach in translating the coordinates. This means we determine which temperature values are within the disc and annulus centered on the coordinates of the tracer galaxy itself, as opposed to translating the tracer galaxy coordinates to the reference pixel and populating the disc based on the center of the pixel. As in DB17, we use postage stamps rather than full map to speed up the code. Pixell is used to create reprojected postage stamps recentered on the galaxy location.
- Aperture photometry Fractional pixel weighting: In implementing the aperture photometry, we account for cases in which pixels are only partially included in the annulus or disc. This includes both pixels centered outside of the disc or ring, and is especially important for pixels that span between the disc and annulus. This is done by creating a finer resolution pixel grid as described in section III.A.

Scenario	ACT	SDSS	Galaxy query	CDELT	Submap	Aperture	Submap
	map	sample	& K-correction		rounding	averaging	centering
1	DB17	DR11	As in $DB17$	Approx.	Approx.	Full	Pixel
2	DB17	DR11	As in DB17	Precise	Approx.	Full	Pixel
3	DB17	DR11	As in DB17	Precise	Precise	Full	Pixel
4	DR4 ILC	DR11	As in DB17	Precise	-	Fractional	Galaxy
5	DR4 ILC	DR15	V21	Precise	_	Fractional	Galaxy
6	DR5 f150	DR15	V21	Precise	-	Fractional	Galaxy

TABLE III: Summary of scenarios utilized in Figure 7 which demonstrate in a step-by-step fashion, with step changes highlighted in blue, the impact of various assumptions used in this analysis and the earlier analysis in DB17.



FIG. 6: Comparison of the impact of the redshift smoothing factor in the aperture photometry estimation for the DR4 ILC map and the DR15 L43 sample. [Upper] The uncertainties introduced in $\hat{p}_{\rm kSZ}$ when varying $\sigma_z=0.01, 0.02, 0.05$ are compared with 1σ uncertainties from noise simulations (assuming $\sigma_z = 0.01$). [Lower] The individual \bar{T}_{AP} realizations for each case.

- Aperture photometry Noise weighting: We compared flat and noise-weighting schemes for differencing the kSZ temperature decrements in the pairwise momentum estimator in (4) and found little difference in the resulting signal.
- Aperture photometry Aperture size: We compare the signal and covariance derived from aperture photometry with a fixed 2.1' aperture at all redshifts, and an aperture that varies with redshift in proportion to the angular diameter distance, D_A , calculated assuming the best fit Planck cosmology, $\Theta(z) = 2.1'D_A(z)/D_A(z = 0.5)$, scaled to be 2.1' at the sample's mean redshift. In Fig. 5 we show



FIG. 7: Comparison of the impact of step-by-step changes in the aperture photometry assumptions in pairwise momentum estimation, \hat{p} obtained from those used in DB17 to this work. The stepwise changes in analysis and data are given in Table III for samples with a $L > 7.9 \times 10^{11} L_{\odot}$ luminosity cut. The gray shaded region shows the 1σ boostrap-derived uncertainties for the DR5 f150 analysis.

that the signal and covariance, for the tracer galaxy sample used in this analysis, are minimally affected by the choice of fixed or redshift-varying aperture size.

• Pairwise estimation – Mitigating redshift evolution: In both this paper and DB17, the effects of redshift evolution in the kSZ signature within a comoving separation bin are accounted for by subtracting a redshift averaged temperature estimate, \bar{T}_{AP} , in (3). In Fig. 6 we show that the assumptions about the redshift smoothing factor σ_z do not significantly impact the signal extraction, with the differences in signal being far smaller than the statistical uncertainties in the covariance estimation process.

In Table III and Fig. 7 we present six scenarios that allow stepwise comparison of the kSZ pairwise signal obtained in DB17 to that obtained in our main results when one factors in various updates to the analysis approach. The starting point, scenario 1, utilizes the DB17 CMB map and DR11 sample of 20,000 galaxies and a luminosity cut of $L > 7.9 \times 10^{10} L_{\odot}$. The end point, scenario 6, uses the DR5 f150 map and L79 DR15 sample of 103,159 galaxies used in the analysis in this paper. The transition from scenario 1 to scenario 2 shows that using the precise value of the CDELT parameter in the aperture photometry temperature determination reduces the peak signal by ~ 40% at r = 35 Mpc/h. Scenarios 2 and 3 show the effect of rounding the pixel size and submap centering parameters is less pronounced but does still create a variation in the recovered pairwise signal. Scenarios 3 and 4 show the impact of changing from the DB17 map to the component separated map, DR4 ILC, and introducing a scheme in which the aperture photometry is centered on the galaxies themselves, not the pixel center in which the galaxy resides and in which the temperatures include fractional weighting of pixels that overlap the edges of the apertures. The comparison of scenarios 4 and 5 shows the impact of the transition from the 20,000 DR11 galaxy sample to the 57,828 galaxies in the DR15 sample that overlap with the DR4 ILC map. The parallelized Pixell Python code used in this paper is also employed. The peak amplitude shifts slightly to fall between 35 to 45 Mpc, slightly larger cluster separation than in the DB17 signal. The final transition, from scenario 5 to scenario 6, shows the difference in signal extraction between DR4 ILC and DR5 f150 maps for the L79 DR15 sample. The peak signal shifts towards a separation of 45 Mpc while the amplitude of the peak signal changes only slightly relative to the previous changes. In comparing scenarios 1 and 6, the changes in combination lead to a reduction in the kSZ signal measured in our analysis relative to that in DB17.

Appendix B: kSZ pairwise momentum covariance estimation

In this section we compare the covariances of the kSZ pairwise momentum estimates using jackknife (JK) and bootstrap techniques for the three maps (DR4 ILC, DR5 f150 and DR5 f090) and, for the DR4 ILC map, also compare them to the estimate obtained by averaging over many simulated maps.

For the JK estimation, the clusters in the luminosity bin being considered are binned into N subsamples, removing each subsample exactly once, and each time computing the pairwise estimator according to the remaining (N-1) subsamples. The covariance matrix is then given by

$$\hat{C}_{ij,JK} = \frac{N-1}{N} \sum_{\alpha=1}^{N} (\hat{p}_i^{\alpha} - \bar{p}_i) (\hat{p}_j^{\alpha} - \bar{p}_j), \qquad (B1)$$

where \hat{p}_i^{α} is the signal extracted from the α^{th} JK sample for the i^{th} separation bin, and \bar{p}_i is the mean of the N JK samples [66]. The inverse of $\hat{C}_{ij,JK}$ is a biased estimator of the true inverse covariance, and to address this one uses an additional correction factor [67],

$$\hat{C}_{ij}^{-1} = \frac{(N-K-2)}{(N-1)}\hat{C}_{JK,ij}^{-1}$$
(B2)

where K is the number of comoving separation bins used in the analysis. For our analysis N = 1,000 and K = 19separation bins in total, although we use K = 17 in the performing the optical $\bar{\tau}$ fits (excluding the two smallest separation bins), as described in section IV.B.

For the bootstrap estimation, we randomly reassign the temperature decrements of galaxy positions allowing for repeated values (sampling with replacement). We repeat this process 1,000 times, computing the pairwise kSZ estimator for each replicant sample. We compute the covariance matrix as the sample covariance of the list of pairwise kSZ curves obtained with this process. We note that while the effects of filter overlapping have been shown to cause bootstrap errors to underestimate the covariance for large apertures, the effect has been shown to be negligible for the smaller aperture size used in this analysis ([27], Appendix D and Fig. 27).

For the DR4 ILC we have access to 560 simulated maps produced in [68], which include primary CMB, lensing, and Gaussian but spatially inhomogeneous extragalactic foregrounds and noise due to detector correlations and scan strategy, as generated using the pipeline described in [32] (specifically, we use simulation version v1.2.0). Equivalent simulated realizations are not, however, available for the DR5 f150 and DR5 f090 maps. We use one simulated map (to keep computation times bounded), to compute JK or bootstrap uncertainties using 1,000 random catalog resamplings and compare these estimates to the average sample covariance obtained from the 560 simulations.

Figure 8 shows the standard deviation for the signals obtained from the 560 noise sims, from the JK and bootstrap sampling of the a single noise sim. The purpose of this comparison is to test the covariance inference technique by checking against simulations. We find that at smaller separations the JK variance exceeds that from noise simulations by 10-25% with the effect most pronounced for the higher luminosity bin, which has comparatively fewer galaxies. We note that the number of galaxy pairs is also comparatively smaller at small versus larger spatial separations. We conjecture that the trend may be due to the inadequacy of the JK prefactor in (B1) to account for the double counting in pair space in the resampling strategy needed for this statistic, when resampling is done in catalog space, especially when fewer pairs are present. We leave further investigation of the origins of these effects, and possible remedies, to a future study. By comparison, the uncertainties obtained from the bootstrap technique are more consistent with the standard deviations for the noise simulations.

In Fig. 9, we show the distributions of the χ^2 obtained from the 560 noise sims, and the 1000 resamplings in the JK and bootstrap methods respectively for the 19 spatial separation bins for the L43 and L79 luminosity bins. We



FIG. 8: Comparison of the standard deviation of 560 DR4 ILC noise simulations ('sim') [green] to the uncertainties obtained from the jackknife ('jk') [blue] and bootstrap ('bs') [orange] resampling of a single noise simulation for [Upper] L43 and [Lower] L79 galaxy tracer samples.

find the noise sims well match the expected theoretical χ^2 distribution for 19 degrees of freedom (dof), and that the bootstrap method is comparable, with the best-fit theory having 19.6 and 20.4 dof for L43 and L79 respectively. The JK method however is found to consistently overestimate the uncertainties leading to the underestimation of the χ^2 . The effect is found to be more pronounced in the L79 bin, which has only 30% as many galaxies as L43; the L79 fit is consistent with a theoretical distribution with 13 dof while the L43, is slightly better, but still inconsistent, with a theoretical fit of 15 dof.

In Fig. 10 we show the correlations between the signal across the galaxy separation bins for the L43 sample of the DR4 ILC maps estimated from the bootstrap and JK resampling of a single noise sim and from the covariance of the 560 noise sims. We find the correlations obtained have similar forms across the three techniques, but that the bootstrap resampling better captures the correlations beyond adjacent bins found in the correlation matrix from 560 noise sims, while the JK correlation matrix predicts smaller correlations for these pairs.

In Table IV we summarize how the $\bar{\tau}$ fits vary with the different covariance estimation methods. The PTE values for the JK covariances consistently skew high, especially for the DR5 maps, implying that the JK uncertainties are overestimated. The differences between the



FIG. 9: Comparison of the distribution of χ^2 values obtained from 560 simulation realizations evaluated on covariances inferred using: sims [Black], jackknife ('jk') [Blue] and bootstrap ('bs') [Orange] resampling of a single sim for the DR4 ILC [Upper] L43 and [Lower] L79 samples versus the theoretical prediction for χ^2 for 19 degrees of freedom (equal to the number of spatial separation bins).

JK covariance and that obtained from the 560 sims and bootstrap for the DR4 ILC map are most pronounced for the L43D, L61D and L79 samples, in which the sample sizes are the smallest. The uncertainties in the $\bar{\tau}$ estimate decrease by 10% and the best-fit $\bar{\tau}$ values can shift by up to a half a standard deviation. The larger, cumulative samples, L43 and L61, are far less impacted by the differences in covariance across the methods, with the best-fit $\bar{\tau}$ values largely unchanged, and the uncertainties in $\bar{\tau}$ reduced by ~8%. The SNR is also affected with the JK-derived SNR being lower across the board. The SNR for the DR5 f150 map is similar for the L43 and L61 samples, however for the JK covariances the L43 tracer sample has a slightly higher SNR, while for the bootstrap method the best measured signal is for the L61 sample.

Tracer	Simulated noise covariance									
galaxy	DR4 ILC									
sample	$\bar{\tau} (\times 10^{-4})$ $\chi^2_{\rm min}$ PTE SNR									
L43D	0.27 ± 0.32	14	0.65	0.8						
L61D	0.68 ± 0.32	25	0.10	1.8						
L43	0.48 ± 0.13	20	0.26	3.5						
L61	0.70 ± 0.16	18	0.37	4.1						
L79	0.75 ± 0.23	24	0.13	2.9						

Tracer	JK covariance											
galaxy		DR4 IL	C		DR5 f090				DR5 f150			
sample	$\bar{\tau} (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR	$\bar{\tau} (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR	$\bar{\tau} \; (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR
L43D	0.19 ± 0.35	10	0.92	0.5	0.78 ± 0.39	7	0.98	1.8	0.53 ± 0.27	14	0.67	1.7
L61D	0.82 ± 0.37	18	0.38	2.0	1.06 ± 0.40	10	0.92	2.3	0.71 ± 0.29	8	0.97	2.2
L43	0.46 ± 0.13	17	0.46	3.2	0.66 ± 0.15	11	0.84	4.2	0.54 ± 0.09	14	0.68	5.1
L61	0.77 ± 0.16	14	0.64	4.2	0.78 ± 0.18	10	0.92	3.8	0.68 ± 0.13	9	0.95	4.8
L79	0.67 ± 0.25	16	0.53	2.3	0.72 ± 0.31	8	0.97	2.1	0.88 ± 0.21	9	0.94	3.8

Tracer	Bootstrap covariance											
galaxy		DR4 II	LC		DR5 f090				DR5 f150			
sample	$\bar{\tau} \; (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR	$\bar{\tau} \; (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR	$\bar{\tau} (\times 10^{-4})$	$\chi^2_{ m min}$	PTE	SNR
L43D	0.18 ± 0.32	14	0.67	0.5	0.83 ± 0.34	12	0.81	2.2	0.46 ± 0.24	21	0.24	1.7
L61D	0.69 ± 0.34	25	0.08	1.8	1.07 ± 0.35	15	0.59	2.7	0.72 ± 0.26	11	0.85	2.5
L43	0.47 ± 0.12	22	0.20	3.6	0.65 ± 0.13	13	0.71	4.5	0.54 ± 0.09	17	0.42	5.1
L61	0.74 ± 0.15	18	0.40	4.4	0.82 ± 0.17	16	0.53	4.4	0.69 ± 0.11	10	0.92	5.4
L79	0.78 ± 0.23	21	0.21	3.0	0.79 ± 0.27	12	0.79	2.6	0.88 ± 0.18	13	0.76	4.6

TABLE IV: A comparison of the best-fit $\bar{\tau}$ estimates and 1σ uncertainties for the DR4 ILC [left], DR5 f090 [center] and DR5 f150 [right] maps for the five luminosity-selected galaxy tracer samples using the noise simulation, bootstrap and jackknife (JK) uncertainty estimates. The upper table shows the DR4 ILC results using the covariance obtained across the 560 noise realizations. The center table shows the results using the uncertainties estimated from the JK resampling of the maps, while the lower table shows those using the bootstrap-derived uncertainties (also presented in Table II). The corresponding χ^2 (for 17 degrees of freedom), SNR and PTE values are also given in each scenario.



FIG. 10: The pairwise correlation matrix, for the DR4 ILC map across the 19 spatial cluster separation bins for the L43 galaxy tracer sample, derived from [top] the covariance across 560 noise sims, [center] the bootstrap and [lower] jack-knife resampling.

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