First measurement of gravitational lensing by cosmic voids in SDSS

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\textbf{ABSTRACT}
We report the first measurement of the diminutive lensing signal arising from matter underdensities associated with cosmic voids. While undetectable individually, by stacking the weak gravitational shear estimates around 901 voids detected in SDSS DR7 by Sutter et al. (2012a), we find substantial evidence for a depression of the lensing signal compared to the cosmic mean. This depression is most pronounced at the void radius, in agreement with analytical models of void matter profiles. Even with the largest void sample and imaging survey available today, we cannot put useful constraints on the radial dark-matter void profile. We invite independent investigations of our findings by releasing data and analysis code to the public at https://github.com/pmelchior/void-lensing.

\textbf{Key words:} gravitational lensing: weak – cosmology: observations

1 INTRODUCTION

Voids are low-density environments, interesting both as probes of cosmology via their shape and size distributions (e.g., Bos et al. 2012; Sutter et al. 2012b) as well as laboratories for studying galaxy formation and modified gravity via their internal structure (e.g., Goldberg & Vogeley 2004; Platen et al. 2008; Li et al. 2012; Spolyar et al. 2013). While the existence of voids has been known since the earliest galaxy redshift surveys (Gregory & Thompson 1978; Kirshner et al. 1981), it is only recently with the advent of high-density large-volume spectroscopic surveys such as the Sloan Digital Sky Survey (SDSS; Abazajian et al. 2009) that reliable void catalogs have become available (Pan et al. 2012; Sutter et al. 2012a).

Despite the large numbers of voids detected in spectroscopic surveys, our knowledge is limited by the fact that observed voids are defined by biased, sparse tracers of the underlying matter density, namely galaxies. On the other hand, our understanding of the evolution and characteristics of voids comes from analytical estimation (Furlanetto & Piran 2006; Tinker & Conroy 2009) and dark matter \textit{N}-body simulation (e.g., Colberg et al. 2005; Kreckel et al. 2011; Lavaux & Wandelt 2012), which require semi-analytic modeling (De Lucia 2009) or halo occupation distributions (Berlind & Weinberg 2002; Mandelbaum et al. 2006; Zehavi et al. 2011) to reproduce the statistics of observed galaxy populations. Thus, the link between our theoretical understanding of dark matter underdensities and the voids observed in redshift surveys is tenuous at best. One way to strengthen this link is to measure the matter underdensities directly by means of weak gravitational lensing. While the shear signal from all but the largest single voids will be undetectably small (Amendola et al. 1999), theoretical calculations by Krause et al. (2013) and numerical simulations by Higuchi et al. (2013) suggest that stacking many voids will not only enable the detection of the effect with high significance, it will also constrain the radial profile of voids. By measuring the radial shear profile around a sufficient number of voids, we can therefore directly study their interior structure and the material that surrounds them, thus testing the predictions from numerical simulations.

Our work ties in with a growing list of studies of cosmic density fluctuations based on the comparison of large-scale galaxy and lensing maps (e.g. Planck Collaboration 2013; Van Waerbeke et al. 2013). Whereas such studies exploit the statistical correlation between baryonic tracers and lensing fields, we aim for the lensing signal of a spectroscopically preselected sample of voids, rendering our approach equivalent to stacked cluster lensing. While stacking CMB temperature maps at void locations has already been utilized to detect the imprint of the integrated Sachs-Wolfe effect (Granett et al. 2008), it is the first time that lensing measurements seek to constrain the matter distribution within voids.

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We start with a review of the lensing formalism and void model. In Section 2 we describe the void and weak-lensing catalogs, in Section 3 our novel stacking approach, the resulting void lensing signal, and its significance. We discuss our findings, point out the limitations of our approach, and conclude in Section 4.

Approach

Voids are underdense regions in the matter density distribution $\rho$. Gravitational lensing probes this matter field along some range $D_a$ in redshift,

$$\Sigma(\theta) = \int_{D_a} dz \, \rho(\theta, z) - \bar{\rho},$$

being sensitive only to deviations from the cosmic mean density $\bar{\rho}$—may they be overdense like galaxy clusters or underdense like voids. As we will deal with spherical voids, we adopt a radial coordinate frame, in which $\theta = 0$ specifies the spatial location of the void center. In the weak-lensing limit of small perturbations of the matter distribution, which is certainly justified for voids, the shear

$$\gamma(\theta) = \frac{\Delta \Sigma}{\Sigma_c} = \frac{\Sigma(<\theta) - \Sigma(\theta)}{\Sigma_c}$$

(2)

traces the deviation of the projected surface mass density $\Sigma$ from the average surface density $\Sigma_c(<\theta)$ of all matter inside of a cylinder of radius $\theta$. The critical density $\Sigma_c$ is a function of $D_a$, the redshift distribution of the lensed background galaxies, and the angular-diameter distances between lenses and background. By measuring the gravitational lensing effects of voids we are therefore able to directly constrain the matter field, but only in projection along the line of sight. Derivations of the previous equations and details on weak gravitational lensing can be found in e.g. Bartelmann & Schneider (2001).

In this work we will compare our shear measurements to void shapes found in cosmological simulations. Lavaux & Wandelt (2012, later called LW12) determined an average radial profile,

$$\rho(r|R_v) \approx \bar{\rho} \left[0.13 + 0.70 \left(\frac{r}{R_v}\right)^3\right],$$

(3)

in simulations comprising only dark matter particles. The void radius $R_v$ sets the characteristic scale and is the only free parameter in this self-similar model. The parameters of the LW12 model are determined for voids of $R_v \approx 8$ $h^{-1}$Mpc, but found to describe larger voids similarly well (LW12). By extending this model with a compensation region outside of $R_v$, Krause et al. (2013) calculated the observable shear profile for several spherical void models, including LW12. We will use this as the baseline for our comparison to observed shear profiles around voids in the SDSS footprint.

2 DATA SETS

We use voids from the 2012.11.17 release of the public cosmic void catalog of Sutter et al. (2012a), based on the ZOBOV algorithm (Neyrinck 2008). This catalog identified voids in the SDSS DR7 main sample (Blanton et al. 2005) and luminous red galaxy (LRG, from Kazin et al. 2010) redshift catalogs, spanning a redshift range of $z = 0$–$0.45$ and yielding void sizes from $5$ to $120$ $h^{-1}$Mpc. To avoid systematics induced by truncated profiles of voids near the survey edges and masks, we take the “central” sample, which corresponds to 1031 voids. For 901 of these voids, we can follow the radial shear profiles up to a maximum distance of at least $70$ $h^{-1}$Mpc, unobstructed by the survey edges. Their size and redshift distribution is shown in Figure 1.

The weak-lensing measurements are based on SDSS DR8 imaging (Aihara et al. 2011). Only the r-band was used for this study. The shear catalog is very similar to that used in Sheldon et al. (2009) and is an implementation of the “re-gaussianization” method from Hirata & Seljak (2003). We have verified that the shears in this new catalog are consistent with those used in Sheldon et al. (2009) for objects found in both catalogs.

The re-gaussianization method performs well for high S/N galaxies, with expected calibration errors at less than a percent (Reyes et al. 2012), which we have confirmed with a set of simplified simulations. However, we find that the shears can have “noise bias” (e.g. Melchior & Viola 2012) of several percent for low-S/N galaxies (S/N < 20, cf. also Reyes et al. 2012). Such a calibration bias is multiplicative, and is not expected to be a function of scale when averaged over the full SDSS survey area. Thus this type of bias would not affect the shape of the void signal, but would result in a misestimation of the density in the voids. To alleviate this effect somewhat, we have used a more conservative magnitude cut ($r < 21.5$) for this new catalog. Galaxies at $r = 21.5$ have a median S/N of about $10$, for which we expect a few percent calibration error. Each galaxy receives a weight in the final analysis of

$$\sim 1/(0.32^2 + \sigma_e^2),$$

where $\sigma_e$ is the error in the measured shape and $0.32^2$ is the variance in intrinsic shapes; thus, these galaxies get relatively small weight. Yet they are numerous, so there is certainly some remaining calibration bias in the catalog at the few percent level. However, this error is small compared to the Poisson noise in the void lensing measurements, as we will show below.

We also apply a small “de-trend” to the ellipticities in the catalog, subtracting the mean ellipticity as a function of location in the CCD array and resolution factor $R$, which characterizes the size of the galaxy compared to the PSF width (see Hirata et al. 2004, eq. 8). For the bulk of objects, this mean ellipticity is negligible. However, for galaxies with $R$ values near $1$, and also for those falling on certain areas of the focal plane for a subset of our data, the mean ellipticity can be a few percent (Huff et al. 2011).
for details). We find that, after de-trending, the catalog successfully passes various null tests, such as B-mode tests and mean shear around random points, even on the scales probed in the void analysis carried out here. Results for B-mode test and random points are shown in Appendix A. We also trim the catalog to the footprint of the Baryon Oscillation Spectroscopic Survey (Dawson et al. 2013, BOSS) and remove galaxies near bright stars and in bad fields (where the processing was unsuccessful).

Finally, only those galaxies with a good photometric redshift estimate are used in the shear analysis (see Sheldon et al. 2012) for a description of the catalog selection). We use the photometric redshift distributions from Sheldon et al. (2012) and expect calibration errors associated with these \( p(z) \) to be less than one percent for the redshifts of interest. We think that analysis is sufficient to characterize the errors for the current work, but note a more realistic treatment of these errors will be needed for more precise studies.

3 RESULTS

As our void catalog comprises voids with size from 5 to about 120 \( h^{-1} \text{Mpc} \), stacking the measured shear profiles as a function of the physical distances \( r \) would wash out the lensing signal considerably. We therefore rebin the measurements in units of the void radius, \( r/R_v \), to take full advantage of the entire void sample, despite the large variations in \( R_v \) in the entire sample or any reasonably sized subsample.

In Figure 2 we show the inferred radial surface density contrast rescaled by the void radius, \( \Delta \Sigma / R_v \) as a function of the normalized radius \( r/R_v \) together with the analytical LW12 void model.\(^{4}\) For the whole sample, the E-mode shows a depression in three consecutive bins around the void. The depression fades towards smaller and larger radii, consistent with the model, but with decreasing significance. The expected depression is also present in the main void sample, albeit at even larger statistical errors. The LRG sample also shows the depression leading towards \( r = R_v \), but returns to a null signal quickly thereafter.

To assess the significance of a lensing detection, we compute the likelihood ratio

\[
K \equiv \frac{\mathcal{L}_m}{\mathcal{L}_0} = \exp \left[ -\frac{1}{2} (\chi_m^2 - \chi_0^2) \right],
\]

where \( \mathcal{L}_{m/0} \) is the probability of the data being described by the void model or by the null hypothesis, respectively, and \( \chi^2 \) is the usual residual sum-of-squares. In this case the null hypothesis can be interpreted literally as signal that is consistent with zero at all scales. Because neither void nor null model have any free parameters to fit, the likelihood ratio is equivalent to the Bayes factor with fair priors, so that our approach is a Bayesian model comparison between two scenarios: One, in which the locations on the sky specified by the void catalog correspond to actual void centers; and the alternative, in which the given locations are entirely random.

To determine the likelihoods of either hypothesis, we have to make – and ideally verify – additional assumptions about the signal and the noise distributions. First, we adopt a Gaussian for the functional form of the likelihoods, which we found to be a good approximation to the distribution of bootstraps in all bins, although the distributions have substantial power in the tails.

Secondly, as we look for a signal with a specific shape across several bins, it is crucial to determine the amount of covariance between bins. We claim that our measurement is mostly affected by the shape noise of galaxies, for which no bin-to-bin covariance for any single void can exist, as no lensed background galaxy can be counted at more than one radius from the void center. But as we stack many voids, a single galaxy can potentially contribute to more than one void profile, so that a small bin-to-bin covariance in the stacked profile is expected. As long as the voids in our sample are sparse and do not strongly cluster, this correlation will be suppressed due to the randomness of the void center coordinates, so that multiple inclusion of the same galaxy shape would contribute to the stacked profile with effectively randomized orientations. Beyond that, the signal is also contaminated with lensing caused by other large-scale structure along the line of sight. While this contribution was properly accounted for in Krause et al. (2013), performing the bootstraps per void (as opposed to within bins as shown in Figure 2), we have verified numerically that the off-diagonal entries

\(^{4}\) As this model is self-similar, \( \Delta \Sigma / R_v \) is a unique function of \( r/R_v \), which means the void model has zero degrees of freedom in our stacking approach. Hence, the model curve in Figure 2 is not a fit to the data.
of the covariance matrix of the lensing measurements are too small to affect the analysis presented here. These off-diagonal elements are also very noisy in the bootstrap resamples, so that we restrict ourselves to the fairly robust diagonal entries. To account for the fact that the inverse of an thus estimated covariance matrix (even when only diagonal terms are considered) is not an unbiased estimate of the inverse covariance matrix, we corrected the \( \chi^2 \) values following Hartlap et al. (2007),

\[
\chi^2 \to \chi^2 \frac{N - B - 2}{N - 1},
\]

where \( N \) denotes the number of voids in the sample and \( B \) denotes the number of bins of the profile.

Thirdly, one has to bear in mind that the void model, and hence \( \chi^2_m \), additionally assumes that the void radii are accurate and that the void lensing signal is characterized by the LW12 profile. We will discuss the implications for our analysis in Section 4.

For the full sample, we find \( K = 17.44 \) and a model error of \( \chi^2_m = 8.35 \). With nine independent bins, the model constitutes an excellent fit to the data and is clearly preferred over the null. Due to substantially larger errors, the main sample is less decisive with \( K = 7.94 \). As mentioned above, there is only weak evidence of lensing in the LRG sample, reflected in a rather poor \( \chi^2_m = 13.34 \) and a likelihood ratio \( K = 2.33 \).

As a measure of potential systematic contamination, we also show the B-mode in Figure 2. While it is not zero at all scales, its largest deviations for the full sample occur on the small-scale side where the errors are largest due to a small number of background galaxies. Overall, we find \( \chi^2(B) = 6.83 \), clearly consistent with a null signal. As the B-mode fluctuations appear uncorrelated with the E-mode, we do not expect them to drive the lensing signal.

Taken at face value, we have substantial evidence of lensing in the sample of all voids, and weaker evidence in the main sample. But there are some aspects worthwhile mentioning about the validity of this result. It seems odd that the whole sample has much larger \( K \) without having much smaller \( \chi^2_m \) compared to the main sample, but this is in fact characteristic of the likelihood ratio, whose power to reject the null grows exponentially with sample size if the alternative is true. It is also counter-intuitive that the errorbars for the LRG sample are substantially smaller than for the main sample, so that the larger voids appear in principle better suited to pick up the lensing signal. Because the noise in this measurement primarily stems from the shape scatter of the background galaxies, one would assume that voids at lower redshift with more galaxies behind them should have smaller errors. There are also more voids in the main sample than in the LRG sample. But the profiles in Figure 2 are plotted as \( \Delta \Sigma / R_e \), which also rescales the per-void errors. In addition, the distances \( r \) are also rescaled so that radial bins correspond the annulli on the sky, whose area is proportional to \( r^2 \). For the lensing data at hand, these two effects more than make up for the lower number of voids and background galaxies of the LRG sample. We will point out in Section 4 why this is still not enough to allow for a clear measurement of the lensing signal.

Finally, the most problematic aspect of the analysis lies in the choice of the binning in Figure 2. Precisely because we deal with a weak signal compared to the noise, it matters a lot in which bins the lensing measurements happen to fall. Given a particular binning, such as the one shown above, statistical fluctuations may or may not cancel each other to yield a good or a poor estimate of the mean in any of the bins. So we need to expect a substantial variation of \( L_{m,0} \) and consequently \( K \). Indeed, when varying the binning scheme within plausible limits of \( \max(r/R_e) = \{0.0, 0.1, ..., 0.5\} \) and \( \min(r/R_e) = \{1.6, 1.7, ..., 2.5\} \) using a number of bins \( B + 1 = \{6, 8, ..., 10\} \), Figure 3 shows a very broad distribution of \( K \) in particular for the whole sample. It does, however, qualitatively support the findings we made earlier: The whole sample has substantial evidence for lensing with a median \( K = 11.4 \), while the main sample has weaker evidence with a median \( K = 6.3 \). None of these two distribution has considerable power between \( K 

\[0 \leq K 

\]

\[K < 1 \], which would favor the null hypothesis.

4 DISCUSSION AND CONCLUSION

While we see a substantial depression of the density contrast around the void radius, there are several potential limitations to our analysis, all of which concerns different aspects of the void model being the correct description of actually observed voids.

First, the assumption of self-similarity. It allowed us to rebin the lensing data in terms of \( \Delta \Sigma / R_e \) and \( r / R_e \) and thus stack all voids, irrespective of their actual size, on top of each other. Given the small size of the void sample and the low number density of background sources in the lensing data, this rebinning turned out to be crucial for a significant lensing detection.

If, on the other hand, voids are not self-similar, the approach we took would mix voids with different profiles and hence reduce the statistical power of the test. The same happens when the estimates of the void radius are inaccurate or the stack of voids is not perfectly spherical due to sample variance. Taken to the extreme, our assumed model could be such an inaccurate description of the measured stacked lensing signal that the likelihood ratio \( K \) would be in favor of the null, even if the void catalog provides valid void locations and these voids properly act as (anti-)lenses. This is not the case here. While we do not claim that voids are necessarily self-similar in nature or that the void radii are precisely estimated or that there is no residual deviation from average sphericity, these conditions seem to be fulfilled well enough to enable our approach.

Secondly, in addition to self-similarity we have adopted a particular, perfectly compensated void model, which means that when
integrating out to a sufficiently large radius, the enclosed matter will have exactly average cosmic density. Specifically, Krause et al. (2013) assumed a constant density compensation region between 1.0921 \( R_c \) and 2 \( R_c \) such that the void is compensated at this outer radius. This has implications for the shape of the expected shear profiles, mainly for the value of the profile around \( r = R_c \). (see their Fig. 3 for details). In fact, recent studies of the galaxy distribution in SDSS and in realistic mock simulations have revealed that small voids (especially with \( R_v \leq 10 h^{-1} \text{Mpc} \)) tend to be overcompensated, whereas large voids have very little compensation at all (Ceccarelli et al. 2013; Hamaus et al. 2014). Given the limited significance of our results, we can neither support nor reject such a dichotomy on the basis of a lensing measurement, which would be free of galaxy bias.

Finally, we address the lack of a significant lensing signal in the LRG sample. Because large voids should not be fully compensated, they should generate an even more negative lensing signal than predicted by our compensated model, also extending far beyond 2\( R_v \), which would make it easier to pick up than in the main sample. However, the sparse sampling of the galaxy field by LRGs introduces additional uncertainties for the void finding algorithm. In a dedicated simulation study, we found that optically detected voids in the LRG sample still correspond to underdensities in the dark matter distribution, but the average level of underdensity is reduced by a factor 2, at least within \( \frac{1}{2} R_v \) (Sutter et al. 2014). As a consequence, the lensing signal would suffer a similar degradation. Due to the sparse sampling, we also expect larger uncertainties in the void radius estimate, which would reduce the significance of the data in our rebinning approach even further. These two issues can make LRG voids suboptimal for a lensing detection and hence render their lensing signal undetectable given the limited statistical power of our data set.

In summary, by using a spectroscopically selected void sample and a well-tested shear catalog together with a novel rebinning technique, we were able to detect the lensing signal arising from the underdensities of cosmic voids in the SDSS DR7 footprint with a median likelihood ratio of 11.4:1 over a random null signal. Even if the statistical power of our data set is reduced by about a factor 2, at least within \( \frac{1}{2} R_v \), the void is compensated at this outer radius (Sutter et al. 2014).

ACKNOWLEDGEMENTS

The authors thank Guilhem Lavaux and Christopher Bonnet for useful discussions. The manuscript profited considerably from suggestions by Uroš Seljak and the anonymous referee. PM is supported by the U.S. Department of Energy under Contract No. DE-FG02-91ER40690. PMS and BDW acknowledge support from NSF Grant AST-0908902. BDW acknowledges funding from an ANR Chaire dExcellence, the UPMC Chaire Internationale in Theoretical Cosmology, and NSF grants AST-0908902 and AST-0708849. ES was supported in part by the U.S. Department of Energy under Contract No. DE-AC02-98CH10886.

REFERENCES

Abazajian K. N., et al., 2009, ApJS, 182, 543
Blanton M. R., et al., 2005, AJ, 129, 2562
Colberg J. M., Sheth R. K., Diaferio A., Gao L., Yoshida N., 2005, MNRAS, 360, 216
Furlanetto S. R., Piran T., 2006, MNRAS, 366, 467
Goldberg D. M., Vogele M. S., 2004, ApJ, 605, 1
Kirshner R. P., Oemler A., Schechter P. L., Sheth R. K., 1981, ApJ, 248, L57
Kreckel K., Ryan Joung M., Cen R., 2011, ApJ, 735, 132
Platen E., van de Weygaert R., Jones B. J. T., 2008, MNRAS, 387, 128
Spyolar D., Sahli M., Silk J., 2013, Phys. Rev. Lett., 111, 241103
Tinker J. L., Conroy C., 2009, ApJ, 691, 633
Aihara H., et al., 2011, ApJS, 193, 29
Amendola L., Frieman J. A., Waga I., 1999, MNRAS, 309, 465
Bartelmann M., Schneider P., 2001, Phys. Rep., 340, 291
Berlind A. A., Weinberg D. H., 2002, ApJ, 575, 587
Bos E. G. P., van de Weygaert R., Dolag K., Pettorino V., 2012, MNRAS, 426, 440
Ceccarelli L., Poz D., Lores M., Padilla N., Lumbas D. G., 2013, MNRAS, 434, 1455
Dawson K. S., et al., 2013, AJ, 145, 10
De Lucia G., 2009, in Gioiibi G., Tornambe A., Raimondo G., Limongi M., Antonelli L. A., Menci N., Brocato E., eds, American Institute of Physics Conference Series Vol. 1111, pp 3–10
Granet B. R., Neyrinck M. C., Szapudi I., 2008, ApJ, 683, L99
Gregory S. A., Thompson L. A., 1978, ApJ, 222, 784
Hamaus N., Wandelt B. D., Sutter P. M., Lavaux G., Warren M. S., 2014, Physical Review Letters, 112, 041304
Hartlap J., Simon P., Schneider P., 2007, A&A, 464, 399
Higuchi Y., Oguri M., Hanama T., 2013, MNRAS, 432, 1021
Hirata C., Seljak U., 2003, MNRAS, 343, 459
Hirata C. M. et al., 2004, MNRAS, 353, 529
Huff E. M., Eifler T., Hirata C. M., Mandelbaum R., Schlegel D., Seljak U., 2011, arXiv:1112.3143
Kazin E. A. et al., 2010, ApJ, 710, 1444
Krause E., Chang T.-C., Doré O., Umetsu K., 2013, ApJ, 762, L20
Lavaux G., Wandelt B. D., 2012, ApJ, 754, 109
Li B., Zhao G.-B., Koyama K., 2012, MNRAS, 421, 3481
Mandelbaum R., Seljak U., Kauffmann G., Hirata C. M., Brinkmann J., 2006, MNRAS, 368, 715
Melchior P., Viola M., 2012, MNRAS, 424, 2757
Neyrinck M. C., 2008, MNRAS, 386, 2101
Pan D. C., Voegle M. S., Hoyle F., Choi Y.-Y., Park C., 2012, MNRAS, 421, 926
Planck Collaboration 2013, arXiv:1303.5078
Reyes R., Mandelbaum R., Gunn J. E., Nakajima R., Seljak U., Hirata C. M., 2012, MNRAS, 425, 2610
Sheldon E. S., et al., 2009, ApJ, 703, 2217
Sheldon E. S., Cunha C. E., Mandelbaum R., Brinkmann J., Weaver B. A., 2012, ApJS, 201, 32
Sutter P. M., Lavaux G., Wandelt B. D., Weinberg D. H., 2012a, ApJ, 761, 44
Sutter P. M., Lavaux G., Wandelt B. D., Weinberg D. H., 2012b, ApJ, 761, 187
Sutter P. M., Lavaux G., Lavaux B. D., Weinberg D. H., Warren M. S., 2014, MNRAS, 438, 3177
Van Waerbeke L. et al., 2013, MNRAS, 433, 3373
Zehavi I. et al., 2011, ApJ, 736, 59

APPENDIX A: B-MODE AND RANDOM POINT SIGNAL

In Figure A1 we compare the B-mode signal (labeled \( \Delta \Sigma_{\perp} \)) to the E-mode signal (labeled \( \Delta \Sigma_{\parallel} \)). We expect no B-mode if the signal is created by gravitational lensing. The stacked signal is shown for voids drawn from two separate redshift ranges \([0.02, 0.10]\) and
Figure A1. Lensing E-mode ($\Delta \Sigma_e$, black) and B-mode ($\Delta \Sigma_b$, red) for voids in two separate redshift bins. The signal was averaged as a function of physical radius rather than in units the void radius.

Figure A2. Lensing signal around voids (black) to the signal around random points (red) for two different redshift ranges. The signal was averaged as a function of physical radius rather than in units the void radius. In contrast to Figure 2, the radial binning is in physical units rather than units of the void radius. We find no significant detection of B-modes in either redshift bin, consistent with our findings in Section 3.

In Figure A2 we compare the signal the E-mode signal measured around voids to that measured around random points in the survey footprint. The footprint is taken directly from the SDSS DR7 public release and is the same used to identify boundaries when finding voids (see Fig. 3 of Sutter et al. 2012a). Redshifts were drawn uniformly in the volume from redshift zero to redshift 0.3 and weighted to match the redshift histogram of the voids. There is a detection of a small signal around random points at small and large scales, but the amplitude is not large enough to account for the detected void signal. We subtract the mean random points signal from the signal around voids in the analysis presented above.