MEASURING THE GEOMETRY OF THE UNIVERSE FROM WEAK GRAVITATIONAL LENSING BEHIND GALAXY GROUPS IN THE HST COSMOS SURVEY

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ABSTRACT

Gravitational lensing can provide pure geometric tests of the structure of spacetime, for instance by determining empirically the angular diameter distance–redshift relation. This geometric test has been demonstrated several times using massive clusters which produce a large lensing signal. In this case, matter at a single redshift dominates the lensing signal, so the analysis is straightforward. It is less clear how weaker signals from multiple sources at different redshifts can be stacked to demonstrate the geometric dependence. We introduce a simple measure of relative shear which for flat cosmologies separates the effect of lens and source positions into multiplicative terms, allowing signals from many different source–lens pairs to be combined. Applying this technique to a sample of groups and low-mass clusters in the COSMOS survey, we detect a clear variation of shear with distance behind the lens. This represents the first detection of the geometric effect using weak lensing by multiple, low-mass groups. The variation of distance with redshift is measured with sufficient precision to constrain the equation of state of the universe under the assumption of flatness, equivalent to a detection of a dark energy component ΩX at greater than 99% confidence for an equation-of-state parameter −2.5 ≤ w ≤ −0.1. For the case w = −1, we find a value for the cosmological constant density parameter ΩX = 0.85^{+0.04}_{−0.02} (68% CL) and detect cosmic acceleration (q0 < 0) at the 98% CL. We consider the systematic uncertainties associated with this technique and discuss the prospects for applying it in forthcoming weak-lensing surveys.

Key words: cosmology: observations – dark energy – distance scale – galaxies: groups: general – gravitational lensing: weak – large-scale structure of Universe

Online-only material: color figures

1. INTRODUCTION

The current evidence for a dominant dark energy component in the universe (e.g., Percival et al. 2010; Larson et al. 2011; Sullivan et al. 2011) leaves cosmology in an uncomfortable situation. Given the concurrent evidence for cold dark matter (CDM) and an additional field driving inflation, it seems several radical new components of physics are required to explain the present-day state of the universe, with little detailed observational information so far as to their precise nature. Clarifying the nature of dark energy is particularly challenging. Observationally, the effect of dark energy on the equation of state is very close to that of a cosmological constant Λ. To determine anything else about this component requires very precise tests and a rigorous elimination of systematics. The simplest evidence for dark energy comes from measurements of the geometry of spacetime, or equivalently distance as a function of redshift, either from supernovae (Sullivan et al. 2011), which measure luminosity distance, or baryon acoustic oscillations (BAOs; Percival et al. 2010; Beutler et al. 2011) or the cosmic microwave background (CMB; Larson et al. 2011), which measure angular diameter distance. Gravitational lensing also provides tests of cosmology, measuring both the matter distribution and how gravity deflects light on large scales. There has been much emphasis on cosmological lensing tests using cosmic shear (e.g., Massey et al. 2007a; Fu et al. 2008; Schrabback et al. 2010; Semboloni et al. 2011), which is sensitive both to the matter distribution and to spacetime geometry. While the theory of CDM structure formation makes fairly robust predictions as to the matter distribution, and thus the lensing potential, on scales larger than galaxies, it also assumes additional physics associated with inflation, such as Gaussian distribution of initial fluctuations with an almost-scale invariant power spectrum. By separating out the geometric effects of
lensing from the properties of the density field, one can, in principle, derive more general constraints on geometry independent of these assumptions.

The underlying idea of geometric lensing tests is straightforward. First, the strength of lensing is measured behind an object as a function of redshift. Then, taking ratios of the lensing signal at different redshifts, the dependence on redshift gives a measure of spacetime geometry via the angular diameter distance–redshift relation. The attraction of this technique is that (at least in principle) any uncertainties in the exact form of the lens potential cancel out. A cosmological constant (or dark energy with a similar equation of state) manifests itself by increasing the distance to an object at a given redshift. In geometric lensing tests, the amplitude of the lensing distortion provides an independent estimate of the distance that the source lies behind the lens. Thus, by measuring average distortion versus offset in redshift, one can constrain the value of $\Lambda$ or more generally $\Omega_X$, the dark energy density parameter.

In practice, the specific implementations of this idea fall into two broad categories. The most developed tests use one or a few massive clusters to provide the lensing signal and measure the position of strongly lensed arcs to determine the strength of lensing as a function of redshift behind the cluster. This method was first discussed in detail by Link & Pierce (1998; although the idea is much older, e.g.,Refsdal 1966) who assumed the simplest singular isothermal sphere potential for the cluster. The method was subsequently revisited by Golse et al. (2002), who considered several sources of systematic error, in particular the effect of substructure and irregular cluster mass distributions. They applied the test to the galaxy clusters AC114 and A2218, obtaining constraints $0 < \Omega_M < 0.33$ on the matter density parameter and $w < -0.85$ on the dark energy equation-of-state parameter (Soucail et al. 2004). The method was also used by Sereno (2002) on the cluster CL 0024+1654, providing evidence for a flat accelerating cosmology. Most recently, Jullo et al. (2010) have derived tight constraints using 28 images from 12 multiple image families in A1689.

A related method uses weak lensing to measure the amplitude of the lensing signal as a function of redshift behind a cluster (see Lombardi & Bertin 1999; Gautret et al. 2000 for early forms of this test, and Clowe 1998 for an early photometric detection). Wittman et al. (2001) first used the weak-lensing signal to determine the redshift to a cluster by lensing alone, and it has subsequently been used by Gavazzi & Soucail (2007) to estimate redshifts to a dozen clusters in the CFHTLS Deep fields. Most recently Medezinski et al. (2011) have demonstrated the effect for three massive clusters using rough photometric redshifts (photo-zs) to determine mean redshifts to different samples of lensed galaxies. This work should provide interesting constraints on dark energy when extended to their full sample of 25 clusters.

A second category of tests uses the combined signal from many less massive halos as the source of the lensing effect. Jain & Taylor (2003) proposed the first such test, the “cross-correlation tomography” method. Here the mass distribution is inferred statistically from the foreground galaxy distribution, while the lensing signal is measured with weak shear in two background samples; ratios of the galaxy-shear cross-correlation functions for the two samples then probe geometry. Variants of this method have been developed by Bernstein & Jain (2004) and Zhang et al. (2005). Taylor et al. (2007) proposed applying this technique behind clusters using ratios of individual shear measurements, rather than correlation functions. Their revised method was demonstrated together with three-dimensional cosmic shear in Kitching et al. (2007) and systematics and error forecasts were discussed in detail in Kitching et al. (2008). If the stacked signal from lenses in a narrow redshift range is sufficiently strong, the geometric dependence behind the sample can be demonstrated directly. For instance, Hoekstra et al. (2005) show that the lensing signal increases with source photometric redshift for a sample of galaxy-scale lenses at $z = 0.2–0.4$.

The two kinds of tests have different advantages and disadvantages. Tests using a small number of massive clusters benefit from significant lensing signal and external constraints on the form of the mass distribution from optical or X-ray data, but suffer from cosmic sample variance and uncertainties in the cluster profile. Since these tests probe only one or a few lines of sight, real structures in front of or behind the cluster—voids or other halos—will enhance or reduce the signal at particular redshifts. Although ideas for modeling the line of sight are emerging (e.g., Hoekstra et al. 2011), the results are not yet satisfactory, and the current practical consensus is to average the signal coming from different clusters. Simulations suggest that at least ~10 massive clusters with many multiple-image systems each are needed to overcome cosmic sample variance (Dalal et al. 2005; Gilmore & Natarajan 2009; D’Aloisio & Natarajan 2011). Furthermore, massive clusters are relatively rare, so only a fraction of the total lensing signal from all cosmic structure can be used. Tests using a larger number of less powerful lenses sample more of the total lensing signal but require large survey areas with accurate photometric redshifts and may be subject to more subtle uncertainties in the foreground mass distribution. Furthermore, since the geometric term in the lensing equation depends on source and lens redshifts separately, it is not immediately clear how to stack the results from large samples of lenses in a simple way. The analysis is thus less intuitive, making it harder to spot unanticipated systematics in the results.

The COSMOS survey provides an interesting data set with which to explore geometric lensing tests. COSMOS has an unusual combination of a high density of sources with accurate lensing shape measurements and accurate photo-zs for a large fraction of these sources. Unfortunately, the COSMOS field has no really massive clusters in it; the largest cluster has an estimated mass of $2.5 \times 10^{14} M_\odot$ (Finoguenov et al. 2007), 10 times less than the largest strong-lensing clusters, and is at a redshift of $z = 0.73$ (Guzzo et al. 2007) where lensing is past the peak in sensitivity, given the source redshift distribution. The lensing signal in the COSMOS field comes instead from many low-significance, group-sized halos (Finoguenov et al. 2007; Leauthaud et al. 2010). Collectively these systems could still produce a large lensing signal to test geometry, however, provided the signal could be stacked.

Here we propose a simple method for stacking the signal from multiple lenses into a single measure of geometry, in effect the relation between comoving or angular diameter distance and redshift. Applying this new “stacked shear ratio test” to X-ray-selected groups in the COSMOS field, we obtain a clear detection of the geometric signal and derive significant constraints on the dark energy density parameter $\Omega_X$. While the COSMOS field is probably too small to overcome sample variance limitations, the magnitude of our statistical errors illustrates the future promise of this technique.

The outline of paper is as follows. In Section 2, we present the basic data, including the sample of lensing groups, and the source redshifts and shape measurements of the COSMOS.
lensing catalog. In Section 3, we introduce the stacking technique and discuss optimal weighting for this technique. In Section 4, we use the stacked shear ratio test to derive parameter constraints on the density of dark energy \( \Omega_X \) and the equation-of-state parameter \( w \), and discuss possible systematics for this test. In Section 5, we summarize our results and discuss future prospects for applying the stacked shear ratio test to other weak-lensing surveys. Throughout the paper, we refer our results to the *Wilkinson Microwave Anisotropy Probe* (WMAP) seven-year mean parameter values derived in Larson et al. (2011), taking a flat cosmology with \( \Omega_m = 0.73 \), \( \Omega_{\Lambda} = 0.27 \), \( H_0 = 70 \, h_{70} \, \text{km s}^{-1} \, \text{Mpc}^{-1} \) (WMAP7 hereafter) as our baseline.

2. DATA

2.1. The COSMOS Survey and Group Sample

The COSMOS survey (Scoville et al. 2007a) brings together panchromatic imaging from X-ray to radio wavelengths, including the largest contiguous area observed with the *Hubble Space Telescope* (HST), and deep optical spectroscopic observations. The field covers an area of 1.64 deg\(^2\) centered at 10:00:28.6, +02:12:21.0 (J2000) and contains identified groups, clusters, and larger structures spanning a wide range in redshift (Scoville et al. 2007b).

We consider the gravitational lensing signal behind a sample of galaxy groups selected originally via their X-ray emission (Finoguenov et al. 2007) and updated using a combined mosaic of imaging from *XMM-Newton* (1.5 Ms; Hasinger et al. 2007; Cappelluti et al. 2009) and the *Chandra* observatories (1.8 Ms; Elvis et al. 2009). Groups are detected from the combined X-ray mosaic using a wavelet filter, which can result in centering uncertainties of up to 32". The distribution of galaxies along the line of sight to each X-ray detection is searched for a red sequence overdensity to determine the group redshift, with spectroscopic redshifts used for subsequent refinement (Finoguenov et al. 2007). Group members are selected based on their photometric redshift and proximity to the X-ray centroid, using an algorithm tested extensively on mock catalogs and spectroscopic subsamples (George et al. 2011). Stellar masses of the member galaxies are determined from multiwavelength data (see Leauthaud et al. 2012 for details). From an initial list of members, group centers are then redefined around the most massive group galaxy within the Navarro–Frenk–White (NFW) scale radius of the X-ray centroid (\( M_{\text{X-ray, scale}} \)), which optimizes the weak-lensing signal at small radii (M. George et al., in preparation). For the majority of our groups, this gives centers that agree with the X-ray centroid; a minority (approximately 20%) of groups show significant offsets between the most massive galaxy and the X-ray centroid. These offsets could be due to observational problems (such as low signal to noise (S/N) in the X-ray or optical data), or they might indicate unrelaxed, low-concentration groups with poorly defined physical centers, such as recent mergers. We will consider below both the full set of groups and a "restricted" set which excludes the systems with significant offsets. The centering algorithm will be discussed further in a forthcoming paper (M. George et al., in preparation).

The full X-ray group sample, together with derived properties, is available from the NASA/IPAC Infrared Science Archive (IRSA) Web site\(^\text{13}\) (see George et al. 2011 for details).

We restrict the lens sample to groups at \( z < 1 \) to ensure the reliability of X-ray detections and optical associations, as well as good photometric redshifts for identifying members and centers. We further cut out of the sample poor groups, groups with centroids affected by masking, and possible mergers (this corresponds to taking only groups with FLAGINCLUDE = 1 as defined in George et al. 2011). Our final sample consists of 129 systems (105 in the restricted set) spanning a rest-frame 0.1–2.4 keV luminosity range between \( 10^{41} \) and \( 10^{44} \) erg s\(^{-1}\), with estimated virial masses of \( 0.8 \times 10^{13} \)–\( 2 \times 10^{14} \) \( h_{70}^{-1} \) \( M_\odot \), virial radii of 0.4–0.8 \( h_{70}^{-1} \) Mpc, and projected angular sizes of 1′–6′. Figure 1 shows the mass, physical size, and angular size for the groups in the sample (note that units have been converted from the value \( h_0 = 72 \, \text{km s}^{-1} \, \text{Mpc}^{-1} \) used in the catalog to \( h_0 = 70 \, \text{km s}^{-1} \, \text{Mpc}^{-1} \)). The virial radius is taken to be \( R_{200c} \), the radius within which the mean density is equal to 200 times the critical density \( \rho_c(z) \) at the redshift of the group, and the virial mass is taken to be \( M_{200c} \), the mass enclosed within \( R_{200c} \). These masses and radii are estimated from the X-ray data, using X-ray scaling relations calibrated with lensing data (Finoguenov et al. 2007; Leauthaud et al. 2010; George et al. 2011).

2.2. Weak-lensing Galaxy Shape Measurements

High-resolution imaging of the COSMOS field was obtained with the *HST* between 2003 October and 2005 June (Scoville et al. 2007c; Koekemoer et al. 2007). The main program consisted of 575 slightly overlapping pointings of the Advanced Camera for Surveys (ACS) Wide-Field Camera taken with the F814W (approximately I band) filter. At each pointing, four slightly dithered exposures of 507 s were obtained. Any cosmetic defects and reflection ghosts were carefully masked by hand. Using version 2.5.0 of the SExtractor photometry package (Bertin & Arnouts 1996), in a Hot–Cold configuration on the stacked images, we detected compact objects in a 0′.15 diameter

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\(\text{13} \) http://irsa.ipac.caltech.edu/COSMOS/tables/groups
aperture down to $F814W_{AB} = 26.6$ at $5\sigma$ (Leauthaud et al. 2007).

We measure the shapes of galaxies in this catalog using the RRG method (Rhodes et al. 2000), largely following the analysis pipeline of Leauthaud et al. (2007) and calibrated against simulated ACS images generated with the Simulate package (Massey et al. 2004, 2007b). However, we now include two significant improvements on this earlier work.

First, we correct trailing in the ACS images due to charge transfer inefficiency via a physically motivated readout model (Massey et al. 2010) that acts at the pixel level, rather than a parametric scheme at the catalog level. This moves electrons in the raw exposures back to where they should have been read out. The method achieves a 97% level of correction and is robust to variety in galaxy morphology, local galaxy density, and sky background level. After correction, residual shears are well below statistical measurement precision.

Second, we model the HST’s point-spread function (PSF) as a function of chip position, telescope focus offset, and velocity aberration factor. The latter two parameters reflect HST’s thermal condition during each exposure: slight expansion and contraction changes the PSF. Following Rhodes et al. (2007), we measure the focus offset (the distance between the primary and secondary mirrors) with a precision of $1 \mu m$ by comparing the apparent shapes of $\sim 10$ stars in each exposure to TinyTim models (Krist 2003). Jee et al. (2007) found that focus offset correlates with the first Principal Component of PSF variation and accounts for 97% of the power, while Schrabback et al. (2010) found that VAFOCUS correlates with the next Principal Component. We measure the shapes of all stars in the COSMOS imaging, then interpolate between them using all four measured parameters. This improves residuals compared to Massey et al. (2007a), and we retain this physically motivated approach rather than relying solely on Principal Component Analysis.

We have also revised our method for determining the variance of the tangential shear slightly. This is now determined empirically, as described in Section 3.5 of Leauthaud et al. (2012). Galaxies are binned by $S/N$ and magnitude, and the total variance of the shear components $\gamma_1$ and $\gamma_2$ is measured directly in each bin. This empirical derivation of the shear dispersion includes both the scatter due to intrinsic variations in galaxy shape and the additional scatter due to shape measurement errors. We find that the shear dispersion varies from $\sigma_\gamma \sim 0.25$ for bright galaxies with high $S/N$ to $\sigma_\gamma \sim 0.4$ for faint galaxies with low $S/N$. These measured values may be very slightly overestimated, however, as suggested by the reduced $\chi^2$ of the profile fit discussed in Section 3.2 below.

### 2.3. Photometric Redshift Measurements

Of the 129 groups in our full sample, 95% contain two or more spectroscopically confirmed members, 3% have one spectroscopically confirmed member, and the remainder have redshifts determined photometrically from the red sequence of member galaxies (as in Finoguenov et al. 2010). The average redshift error for the group ensemble is $\sim 0.0017$, only slightly larger than their typical velocity dispersions of $300-450$ km s$^{-1}$.

Multicolor ground-based imaging in over 30 bands (Capak et al. 2007) also provides photometric redshift information for all of the source galaxies along the same line of sight. We use photometric redshift estimates from the LePhare $\chi^2$ template-fitting code, which are updated from those published in Ilbert et al. (2009) by the addition of deep $H$-band data, and small improvements in the template-fitting technique. We have compared photo-zs from 10801 galaxies at $z \sim 0.48$, 696 at $z \sim 0.74$, and 870 at $z \sim 2.2$ to spectroscopic redshift measurements with the Very Large Telescope Visible Multi-Object Spectrograph (Lilly et al. 2007) and the Keck Deep Extragalactic Imaging Multi-Object Spectrograph. The rms dispersion in the offset $\sigma_\Delta$ between photometric and spectroscopic redshift is $0.007(1+z)$ at $i^*_g < 22.5$ and $0.02(1+z)$ at $i^*_g < 24$ and $z < 1.3$ (or $0.06(1+z)$ for $i^*_g \sim 24$ and $z \gtrsim 1.3$).

To mitigate against catastrophic failure in estimated photo-zs, for example due to confusion between the Lyman and 4000Å breaks, we reject from the sample all-source galaxies with a secondary peak in the redshift peak in the redshift probability distribution function (i.e., galaxies where the parameter $zp_{SEC}$ is greater than zero in the Ilbert et al. 2009 catalog). The rejected $zp_{SEC} > 0$ galaxy population is expected to contain a large fraction of catastrophic errors (roughly 40%-50%; Ilbert et al. 2006, 2009). For the purposes of cosmological constraints, we further exclude from the sample objects with relative redshift uncertainties $\Delta z/(1+z) \gtrsim 0.05$, taking the average redshift error to be $\Delta z \equiv (zu68_{gal} - z168_{gal})/2.0$, where $zu68_{gal}$ and $z168_{gal}$ are the 68% confidence limits on the redshift, based on the photo-z probability distribution (Ilbert et al. 2009). Our final source sample consists of all galaxies passing these cuts that lie within 6' of a group center. Individual galaxies may enter into the final sample multiple times if they lie within 6' of more than one peak. The photo-z quality cuts reduce the number density of source galaxies to 26 galaxies arcmin$^{-2}$, for a total of $3.7 \times 10^5$ galaxies (3.1 $\times 10^5$ in the restricted sample). The mean redshift of the final sample is $\langle z \rangle = 0.95$ and the mean relative error in redshift is $\Delta z/(1+z) = 0.018$, while the mean magnitude is $\langle I_{F814W} \rangle \sim 24$.

### 3. METHOD

#### 3.1. The Stacking Method

If we consider a source galaxy (or “source” hereafter) at redshift $z_s$ being lensed by a foreground group (or “lens” hereafter) at redshift $z_L$ and observed at $z_o$, in the weak limit the tangential shear induced by the lens will be

$$\gamma_t(r) = \frac{|\langle \Sigma(x) - \langle \Sigma(x) \rangle \rangle \Delta \Sigma(r)/\Sigma_c}{\Sigma_c} = \frac{c^2D_LD_LD_{LS}}{4\pi G D_z D_L D_{LS}},$$

where $\langle \Sigma(x) \rangle$ is the mean surface density interior to projected (physical) radius $r$, $\langle \Sigma(x) \rangle$ is the azimuthally averaged surface density at $r$, and $\Sigma_c$ is the critical surface density defined as

$$\Sigma_c \equiv \frac{c^2}{4\pi G D_z D_L D_{LS}}.$$  

Here $D_z$ denotes the angular diameter distance from the observer to the source

$$D_z = f_z(\omega[z_o, z_s])a(z_o),$$

where $\omega[z_o, z_s]$ is the comoving (or coordinate) distance along a radial ray between the observer and the source

$$\omega[z_o, z_s] = \int_{z_o}^{z_s} dz = \int_{\alpha_s}^{\alpha_o} cdt = \int_{\alpha_s}^{\alpha_o} cda a^2H(a),$$

$D_L$ and $D_{LS}$ are angular diameter distances from the observer to the lens and from the lens to the source respectively, given by

$$D_L = f_L(\omega[z_o, z_L])a(z_L),$$

and

$$D_{LS} = f_L(\omega[z_s, z_L])a(z_L).$$
\[ D_{LS} = f_k(\omega[z_L, z_S])a(z_S), \]  

(e.g., Bartelmann & Schneider 2001).

If we consider the case of a flat cosmology \((k = 0)\), \(f_k(\omega) = \omega\) and thus we can rewrite \(\Sigma_c\) more simply in terms of the comoving distances \(\omega_L \equiv \omega[z_D, z_L]\) and \(\omega_S \equiv \omega[z_D, z_S]\)

\[ \Sigma_c \equiv \frac{c^2}{4\pi G} \frac{\omega_S}{\omega_L}(1 + z_L). \]  

The critical density incorporates all the geometric dependence of lensing; it is the nonlinear dependence of \(\Sigma_c\) on \(z_S\) and \(z_L\) that makes it hard to stack the signal from different source–lens pairs in a straightforward way. We can simplify the dependence, however, by defining the comoving distance ratio \(x \equiv \omega_S/\omega_L\). Note that \(x > 1\) for sources chosen to be behind the lens; we will consider objects with both \(x > 1\) and \(x \leq 1\) below. In terms of \(x\)

\[ \Sigma_c \equiv \frac{c^2}{4\pi G} \frac{(1 + z_L)}{\omega_L} \left(\frac{1}{1 - 1/x}\right). \]  

We can also write this in terms of \(\Sigma_{c,\infty}\), the value of the critical density in the limit \(x \to \infty\), which is

\[ \Sigma_c = \frac{1}{(1 - 1/x)} \Sigma_{c,\infty}, \]  

where

\[ \Sigma_{c,\infty} \equiv \frac{c^2}{4\pi G} \frac{(1 + z_L)}{\omega_L} \]  

depends only on the lens properties, not on the properties of the source galaxy.

From Equation (1), the geometric dependence of all source–lens pairs now takes on a universal form

\[ \gamma_t(r) \Sigma_{c,\infty}/\Delta\Sigma(r) \equiv \Gamma(x) = \left(1 - \frac{1}{x}\right). \]  

\(\Gamma(x)\) corresponds, e.g., to the lensing efficiency \(E\) defined by Golse et al. (2002). In as much as the measured tangential ellipticity \(\epsilon_t\) of each source galaxy is an estimator \(\tilde{\gamma}_t\) of the true tangential shear \(\gamma_t\), we can construct a weighted sum of estimates from individual source galaxies \(j\) with respect to lensing centers \(i\) to recover the universal geometric dependence:

\[ \left(1 - \frac{1}{x}\right) = \sum_{i,j} w_{ij} \tilde{\gamma}_{ij}(x) \]

\[ = \sum_{i,j} w_{ij} \tilde{\gamma}_{ij} \Sigma_{c,\infty}/\Delta\Sigma(r_{ij}) \]

\[ \sum_{i,j} w_{ij} \]  

with weights \(w_{ij}\) chosen to maximize the S/N or sensitivity to cosmological parameters, as discussed below.

Since we are just fitting the data to a fixed function, cosmology appears to have disappeared from Equation (12). In fact, it is hidden in the conversion from measured source and lens redshifts to inferred source and lens distances. For a given cosmology we convert redshifts to comoving or angular diameter distances, construct the weighted sum in Equation (12), and calculate the \(\chi^2\) of \(\Gamma(x)\) with respect to the theoretical expectation \((1 - 1/x)\). This gives us the relative likelihood of that particular set of cosmological parameters; iterating over this process then allows parameter constraints. The only remaining problems are to determine the optimal weights \(w_{ij}\) and surface mass density contrast \(\Delta\Sigma(r)\). We discuss these calculations in the next section.

Before proceeding we should note that our simple stacking analysis ignores several complications. First, it ignores the distinction between true and reduced shear (e.g., Shapiro 2009). In the weak shear limit the two are identical, and for the groups considered here the surface mass density is low enough that the contribution from non-weak shear corrections is unimportant outside \(\sim 50–100\ h^{-1}_{70}\ \text{kpc}\). The effect of the second-order correction term is illustrated in Figure 4 of Leauthaud et al. (2010). Its contribution is roughly comparable to that of the stellar mass in the central group galaxy (see Figure 2 below) at large projected radii and always less than the stellar contribution at radii less than \(\sim 50–100\ h^{-1}_{70}\ \text{kpc}\). Given only that \(\sim 2\%\) of our sources lie at such small projected radii, the effect of these contributions on our fits should be negligible (excluding from the sample all sources within 15’ of group centers, for instance, has no effect on the final results).

Second, we have also assumed flatness in separating the dependence on the lens distance and the source distance. While current cosmological constraints indicate an almost completely flat universe (e.g., Larson et al. 2011 find \(|\Omega_k| \lesssim 0.01\) from various sets of constraints), it would be nice to be able to relax this assumption. Unfortunately, while it is still possible to fit shear ratios between individual pairs in the general case, there is no simple way of stacking all measurements together into a single functional form since the dependence on the two redshifts

14 Note that in order to evaluate \(\Delta\chi^2\) smoothly as we vary the cosmological parameters, the sample selection has to be independent of cosmology, and thus our cuts have to be in an observable such as angular separation, rather than a cosmology-dependent quantity such as physical separation.
can no longer be factored out of $f_s(\omega(z_L, z_S))$ in a simple way. We can estimate the effect of curvature by considering the series expansion for $f_s$ in the limit $|\Omega_\Lambda| \ll 1$. For typical values of $\omega \sim c/H_0$, the next term in the series is smaller by a factor $1/6(\omega/R_0)^2 \sim 1.6 \times 10^{-3}$, where $R_0 \equiv c/(H_0\sqrt{|\Omega_\Lambda|})$. Thus in realistic non-flat cosmologies, we expect a correction of the order of $1.6 \times 10^{-3}$ to our values of $\Gamma(x)$. Compared to our cosmological sensitivity $\Delta \Gamma \sim 0.05$ (see Section 4.2 below), this represents a 3\% correction to our derived parameters. This correction is smaller than errors on $\Omega_X$ we obtain below, although it would quickly become important in larger surveys. In what follows we will ignore the complication of non-zero spatial curvature.

Third and lastly, our analysis assumes a specific functional form for the surface mass density contrast $\Delta \Sigma(r)$, namely a projected NFW profile. We will show below that this functional form is in fact an excellent fit to the stacked data. We could use instead an empirical profile based directly on the data itself, but given the agreement between the NFW model and the data, this would not affect our results significantly.

### 3.2. Radial Dependence and Optimal Weighting

Our goal is to measure the redshift dependence of the group lensing signal. The radial dependence of the surface mass density contrast $\Delta \Sigma(r)$ around groups, although intrinsically very interesting, is essentially a nuisance parameter in this calculation. We need to determine $\Delta \Sigma(r)$, however, in order to weight measurements of individual source galaxy shapes optimally when estimating $\Gamma(x)$. The density contrast around groups was studied in detail in Leauthaud et al. (2010); we reproduce the same calculation here, stacking with respect to physical radius the signal from all groups with well-determined redshifts and centers.

The density contrast profile includes contributions from four main terms:\(^{15}\): the weak shear contribution of the main halo (the “one-halo” term), the average weak shear contribution from nearby halos (the “two-halo” term), a weak shear contribution from the stellar mass of the central group galaxy, and the second-order corrections to the shear in the center of the main halo. Of these, only the one-halo term is important here; the two-halo term only becomes significant at large projected radii ($r \sim 4 h_{70}^{-1}$ Mpc), while the stellar and second-order terms are only significant at small projected radii ($r \leq 50–100 h_{70}^{-1}$ kpc) where we have very few galaxies in the source sample, as discussed in Section 3.1.

We expect that the one-halo term for a single group should follow a projected NFW profile $\Sigma(r)$, whose form $f_{\text{NFW}}(r/r_s)$ is given, e.g., in Wright & Brainerd (2000). The profile has two free parameters, a scale radius $r_s$ and a normalization $\Sigma_0$, or alternately it can be defined in terms of a virial radius $r_{\text{vir}}$ and a concentration $c \equiv r_{\text{vir}}/r_s$. The expected values of these parameters can be inferred from X-ray fluxes, X-ray scaling relations, and theoretical concentration–mass relations. Using the concentration relations of Zhao et al. (2009), for instance, Leauthaud et al. (2010) predicted concentrations in the range 3.6–4.6 for the COSMOS groups (these values assume the definition $r_{\text{vir}} = R_{200c}$; thus they correspond to the values $c_{200c}$ from Zhao et al. 2009). Simulations suggest that individual halos will have significant (\sim 50\%) scatter around these mean values (Zhao et al. 2009). Finally, we note that here we are considering an average profile for all groups in the sample, where the averaging is weighted by surface mass density contrast $\Delta \Sigma$. At a fixed redshift and fixed concentration, $r_s$ and $r_{\text{vir}}$ for each group should scale as $M^{1/3}$ and our stacked profile would be similar to a mass-weighted average. Variations in concentration and redshift complicate this behavior, but we can still use mass-weighted averaging to guide our expectations as to the final values for the concentration or scale radius. Using the concentration relations of Zhao et al. (2009) and assuming a WMAP7 cosmology, for instance, we predicted a mass-weighted average scale radius of $r_s = 154 h_{70}^{-1}$ kpc for our groups.

The surface mass density contrast $\Delta \Sigma(r)$ is related to the tangential shear by

$$\Delta \Sigma(r) = \Sigma_c \times \gamma_t(r).$$

(13)

Thus, $\Delta \Sigma(r)$ can be estimated as in Equation (8) of Leauthaud et al. (2010)

$$\Delta \Sigma(r) = \sum_{ij} w_{ij} \tilde{\gamma}_{t,ij} \Sigma_{c,ij}. \sum_{ij} w_{ij}.$$

(14)

Figure 2 shows the surface mass density contrast for our stacked sample of groups, assuming a WMAP7 cosmology. Points with error bars indicate the mean value inferred from Equation (14), binned logarithmically in radius. The thin dotted lines indicate contributions from a projected NFW profile (middle blue line), the stellar mass of the central galaxy in the group (lower red line), and the sum of these two components (upper green line). The normalization of the stellar contribution is based on the mean stellar mass of the central group galaxy, as inferred from photometry (see Leauthaud et al. 2012 for details).

The main part of the profile is well fit by a projected NFW profile $\Sigma(r)$ with a scale radius $r_s \sim 160–180 h_{70}^{-1}$ kpc and a normalization $\Sigma_0 \sim 200 h_{70} M_\odot$ pc$^{-2}$ at $r_s$. The best-fit values of these two parameters are strongly correlated, so we choose instead to normalize the profile at a radius $r_{\text{fix}} = 250 h_{70}^{-1}$ kpc where the measurement errors are small and the amplitude is less dependent on $r_s$. Thus the profile can be written

$$\Sigma(r) = \Sigma_0 \frac{f_{\text{NFW}}(r/r_s)}{f_{\text{NFW}}(r_{\text{fix}}/r_s)}.$$

(15)

This gives us constraints on $\Sigma_0 = \Sigma(r_{\text{fix}})$ and $r_s$ which are more independent of each other.

We determine values for $\Sigma_0$ and $r_s$ by calculating $\chi^2$ with respect to a projected NFW model. We do not bin the data in radius, since the bin boundaries would change with cosmology or scaling, but sum the contribution from each galaxy individually. For the full group sample and assuming WMAP7 cosmology, we obtain the lowest reduced $\chi^2$ by considering all sources at $x > 1.1$, with no other cut on redshift errors. This gives a slightly high value for the scale radius $r_s = 187_{-29}^{+34} h_{70}^{-1}$ kpc, however. Considering only the restricted sample and/or sources with $\Delta z/(1 + z) \leq 0.05$, we obtain values of $r_s$ closer to the expected value. This suggests that the remaining 20\% of the group sample may be affected by centering problems or that it may include many unrelaxed or disturbed groups with lower mean concentrations. The dependence on photo-z cuts could indicate that the lensing signal in all groups is diluted by group member contamination in the source population when the limits on photo-z errors are relaxed. For the restricted group sample
with the cut on source redshift errors, we obtain best-fit values $\Sigma_0 = 98.8 \pm 11 \, h_{70}^{-1} \, M_\odot \, \text{pc}^{-2}$ and $r_s = 158_{-25}^{+35} \, h_{70}^{-1} \, $ kpc. The reduced $\chi^2$ is marginally higher for this sample than for the much larger uncut sample, but the best-fit value of $r_s$ closer to the expected value, so we will take this as our fiducial profile, and marginalize over values of $r_s$ and $\Sigma_0$ in this range for our cosmological calculations. The best-fit value of $r_s$ also places some constraints on possible centering errors for the groups. We have tested the effect of centering errors by adding random offsets to the individual group centers, with rms values of $6''$, $12''$, and $24''$ in each coordinate. The resulting profiles are still well fit by our model, but the best-fit value of $r_s$ increases to 220, 260, and 340 $h_{70}^{-1}$ kpc for the three cases, respectively. This suggests average centering errors are $\lesssim 6'' \sim 25-50 \, h_{70}^{-1}$ kpc in each coordinate, consistent with other estimates of the centering uncertainty (M. George et al., in preparation).

We have also investigated other forms of stacking. In principle we could correct for the predicted variations in concentration, for instance, stacking in $r/r_s$, or we could stack in comoving rather than physical coordinates. Testing scaling the radius by $r_s$, $v_{\text{vir}}$, or $(1+z_s)$, we find little or no significant improvement in the $\chi^2$ of the fit to the radial profile. In particular, we find only a marginal indication of any trend in concentration with mass or redshift. Given that the halo-to-halo scatter measured in simulations is comparable or larger than to the average trends over the mass and redshift range spanned by our group sample, this is perhaps unsurprising. Furthermore, since the concentration relations are themselves dependent on cosmology, we would have to account for this dependence in our marginalization over cosmological parameters, so we will not attempt to correct for variations in concentration. We can also consider other analytic fits to the profile. We find that NFW is preferred over a cored isothermal profile at the 95% confidence level and preferred over a singular isothermal ($\Delta r \propto r^{-1}$) profile at 97%-98% confidence. Thus, our stacked profile provides empirical confirmation of the NFW model in agreement with other recent high-precision measurements of cluster density profiles (Umetsu et al. 2011; Okabe et al., in preparation).

Finally, we note that the fit to the radial profile gives us an independent check of our empirical shear variance estimates. Because we fit the profile without binning, we have large number of degrees of freedom and thus a narrow range of expected scatter in the reduced $\chi^2$. The best-fit NFW profile has a reduced $\chi^2$ of 0.931 with a very small (0.0027) expected variance-weighted average:

$$\tilde{\chi}^2 \leq 0.931$$

where the variance of the tangential shear is determined empirically, as described in Section 2.2.

4. RESULTS

4.1. The Geometric Signal

Given a functional form for the radial dependence of the surface mass density contrast $\Delta \Sigma(r)$, we can proceed to estimate $\Gamma(x)$ via Equation (12). The weights in the sum can be calculated as the inverse variance of the $\Gamma_{ij}$

$$w_{ij} = (\text{var}[\Gamma_{ij}])^{-1} = \left( \frac{\Delta \Sigma(r_{ij})}{\Sigma_{\infty}} \right)^2 \text{var}[\gamma_{\text{t},ij}]^{-1},$$

where the variance of the tangential shear is determined empirically, as described in Section 2.2.

![Figure 3](https://example.com/figure3.png)

**Figure 3.** Geometric dependence of the lensing signal, plotted vs. the distance ratio $x = u_S/u_L$, for a WMAP7 cosmology with $\Omega_M = 0.27$, $\Omega_L = 0.73$. The solid (red) curve shows the theoretical prediction $\Gamma(x) = 0$ for $x < 1$; $\Gamma(x) = 1 - 1/x$ for $x > 1$. Note that the data have not been weighted for cosmological sensitivity (as in Section 4.2); a fit to the unweighted data favors a slightly shorter length scale for $x$ (i.e., lower value of $\Omega_L$) but with very large uncertainties.

(A color version of this figure is available in the online journal.)

The “model” here, the geometric sensitivity function $(1-1/x)$, is fixed, while the data vary as we change $\Sigma_0$ and $r_s$, which both change $\Delta \Sigma(r)$, and the cosmological parameters, which map the redshifts ($z_S$, $z_L$) onto $x$-values and also determine $\Sigma_{\infty}$. If we restrict ourselves to flat cosmologies with two components, matter and dark energy, then the goodness of fit depends on $\Sigma_0$, $r_s$, $\Omega_X$, and $w$.

Figure 3 shows $\Gamma(x)$ over the range $x = [0, 5]$ for a WMAP7 cosmology with $\Omega_M = 0.27$, $\Omega_L = 0.73$. The density contrast profile parameters are fixed to the best-fit values $r_s = 158 \, h_{70}^{-1}$ kpc and $\Sigma_0 = 98.8 \, h_{70} \, M_\odot \, \text{pc}^{-2}$. The points are weighted averages in bins of 0.3 in $x$, while the solid (red) curve is the theoretical expectation:

$$\Gamma(x) = \begin{cases} 0 & \text{for } x < 1; \\ 1 - 1/x & \text{for } x \geq 1. \end{cases}$$

Weights here are inverse variance, as in Equation (16). The error bars on the data points are calculated as usual for an inverse-variance-weighted average:

$$\sigma_{\Gamma,\text{bin}} = \left( \sum_{\text{bin}} w_{ij} \right)^{-1/2},$$

where the sum is over all pairs $(i, j)$ with values of $x$ in the bin.

Clearly the geometric signal is present in the COSMOS data and measured to reasonable significance over a broad range of distance ratio $x$. Given the possible systematics in the measurement discussed below, the excellent agreement between theory and data illustrates the potential of the method. On the other hand, “$\chi^2$-by-eye” is somewhat misleading for this figure, as the binning in $x$ may hide systematics at particular distance ratios. (There is a 2.1σ indication of positive signal in one
bin at \( x < 1 \), for instance. This could indicate that photo-z errors are scattering sources to lower redshifts, but given the number of bins there is a 26% chance that this is simply a random statistical fluctuation.) As discussed below, in a small field structures at a few redshifts can dominate the lensing signal, introducing excesses or deficits of mass along the line of sight that dominate the signal at particular values of \( x \). The bins chosen here are broad enough to smooth out many of these features, but clearly a goodness-of-fit measurement over the whole data set is required to determine the statistical significance of the apparent agreement in Figure 3.

### 4.2. Parameter Constraints

We can estimate goodness of fit by calculating

\[
\chi^2 = \sum_{ij} w_{ij}[(1 - 1/x_{ij}) - \Gamma_{ij}]^2. \tag{19}
\]

This sum depends on \( \Sigma_0, r_s \), and the cosmological parameters, so marginalizing over the first two parameters gives constraints on the dark energy density \( \Omega_X \) and the equation-of-state parameter \( w \). While the resulting value of \( \chi^2 \) will tell us whether the data are a good fit to the model, this is not necessarily the most sensitive way of determining cosmological parameters. In particular, for flat (\( k = 0 \)) cosmological models with a cosmological constant with density parameter \( \Omega_X = \Omega_{\Lambda} \), as the value of \( \Omega_{\Lambda} \) increases all distances will increase; and therefore so will \( x \) and \( (1 - 1/x) \). Thus, the signal from a given source–lens redshift pair will be compared to \( \Gamma(x) \) at a value of \( x \) which is larger by some factor. The most sensitive probes of this rescaling will be points at large \( x \). To constrain \( \Omega_{\Lambda} \) more precisely, individual measurements should be weighted by this sensitivity.

The exact sensitivity to cosmology itself depends on the cosmological parameters. We can estimate a sensitivity factor and thus a weighting that will be close to optimal over the whole range of \( \Omega_{\Lambda} \), however, by calculating

\[
\Delta \Gamma(z_S, z_L) = \Gamma_{10} [x(z_S, z_L)] - \Gamma_{00} [x(z_S, z_L)] = [x_{10}(z_S, z_L)]^{-1} - [x_{00}(z_S, z_L)]^{-1}, \tag{20}
\]

where \( \Gamma_{10}, \Gamma_{00} \) are the model \( \Gamma(x) \) evaluated for cosmologies with \( \Omega_{\Lambda} = 0.1 \) and 0.9, respectively (with \( w = -1 \) in each case). Figure 4 shows this sensitivity function versus \((z_S, z_L)\). (Note that this sensitivity function has been calculated previously, e.g., in Figure 3 of Golse et al. 2002.) For \( z_S \leq z_L \), the sensitivity function is zero since the model value \( \Gamma(x) = 0 \) independent of cosmology. For \( z_S \geq z_L \), sensitivity generally increases with source or lens redshift.

To maximize S/N in our cosmological constraints, we apply this weighting quadratically in our previous expression for \( \chi^2 \):

\[
\chi_w^2 = \sum_{ij} \Delta \Gamma_{ij}^2 w_{ij}[(1 - 1/x_{ij}) - \Gamma_{ij}]^2 / \sum_{ij} \Delta \Gamma_{ij}^2. \tag{21}
\]

This \( \chi^2 \) can be converted to a likelihood by assuming the error distribution is Gaussian. This is only approximately true in our case, but determining more accurate error distributions would require a significantly more complex error analysis, so we will leave this to future work. Figure 5 shows the likelihood function for \( \Omega_{\Lambda} \), calculated using the restricted group sample and normalized so the area under the curve is 1. The dashed vertical line indicates the value \( \Omega_{\Lambda} = 0.848 \) where the likelihood peaks. Dotted vertical lines indicate 68.2%, 95.4%, and 99.7% (1\( \sigma \), 2\( \sigma \), 3\( \sigma \)) confidence regions. The solid vertical line and shading indicate the mean WMAP seven-year value \( \Omega_{\Lambda} = 0.727 \) and 68% confidence range from Larson et al. (2011). The vertical green line indicates the value \( \Omega_{\Lambda} = 1/3 \), the lower limit for cosmic acceleration (\( \dot{q}_0 < 0 \)).

(A color version of this figure is available in the online journal.)

Figure 4. Sensitivity function \( \Delta \Gamma(z_S, z_L) \). Points with high source or lens redshifts are the most sensitive to changes in cosmology.

Figure 5. Likelihood function for \( \Omega_{\Lambda} \) with a uniform prior on \( r_s \) over the range 120–200 \( h^{-1} \) kpc, using the restricted group sample. (The likelihood has been normalized so the area under the curve is 1.) The dashed vertical line indicates the value \( \Omega_{\Lambda} = 0.848 \) where the likelihood peaks. Dotted vertical lines indicate 68.2%, 95.4%, and 99.7% (1\( \sigma \), 2\( \sigma \), 3\( \sigma \)) confidence regions. The solid vertical line and shading indicate the mean WMAP seven-year value \( \Omega_{\Lambda} = 0.727 \) and 68% confidence range from Larson et al. (2011). The vertical green line indicates the value \( \Omega_{\Lambda} = 1/3 \), the lower limit for cosmic acceleration (\( q_0 < 0 \)).
small values of \( r_s \) are allowed. The average scale radius is predicted to be \( r_s \sim 160 h_{70}^{-1} \) kpc for our sample, based on theoretical concentration relations. Simulations show \( \sim 50\% \) scatter in concentration from halo to halo (e.g., Zhao et al. 2009; Reed et al. 2011 and references therein), but most of these variations should average out in the set of \( \sim 100 \) objects considered here, provided they represent a reasonably unbiased sample. On the other hand the conversion from angular to physical radius will vary by \( \sim 10\%–15\% \) for the range of cosmologies considered, and the range of uncertainty on our fitted value of \( r_s \) is \( \sim 130–210 h_{70}^{-1} \) kpc. Thus, we take a uniform prior on \( r_s \) over the range 120–200 \( h_{70}^{-1} \) kpc, allowing for a variation of \( \pm 25\% \) around the fiducial value \( r_s = 160 h_{70}^{-1} \) kpc. Extending the range of our priors to higher values of \( r_s \) has little effect on the results, while extending it to lower values of \( r_s \) decreases the lower limit on \( \Omega_X \).

Overall, we obtain the estimate \( \Omega_X(w = -1) = \Omega_\Lambda = 0.848^{+0.043}_{-0.047} \), corresponding to a detection of dark energy at more than 99\% confidence. This value is consistent with the most recent WMAP analysis of CMB anisotropies, which finds a mean value \( 0.72^{+0.05}_{-0.03} \) (Larson et al. 2011). Our 68\% confidence range is approximately four times wider than that of WMAP; given the small size of the field considered here (1.64 deg\(^2\)), however, this level of precision demonstrates the power of the geometric test. We note, however, that our error estimates do not include systematic effects. We estimate the magnitude of some of the possible systematics in the next section.

In a cosmology with multiple components with equation-of-state parameters \( w \) and density parameters \( \Omega_i \), cosmic dynamics can be characterized by the deceleration parameter:

\[
q = \frac{1}{2} \sum_i \Omega_i w_i (1 + 3w_i).
\] (22)

For a flat universe with components \( \Omega_m \) and \( \Omega_\Lambda \), we find a value \( q_0 = -0.77^{+0.28}_{-0.06} \), at the present day. Present-day acceleration, which corresponds to \( q_0 < 0 \) or \( \Omega_\Lambda > 1/3 \) if \( w = -1 \), is detected at greater than the 98\% CL (solid [green] vertical line in Figure 5).

We can also extend the constraints to more general dark energy models with \( w \neq -1 \). Figure 6 shows the likelihood surface for models with a dark energy component \( \Omega_X \) with an equation-of-state parameter \( w \). Our current results provide an upper bound of \( w < -0.4 \) (68.2\% CL). They do not provide a lower bound, although there is some information in the constraints on \( w \) as a function of \( \Omega_\Lambda \). The shape of the confidence regions is similar to those derived by geometric tests using strong lensing (e.g., Jullo et al. 2010), although our contours are shifted upward to less negative values of \( w \), perhaps because of the redshift distribution of our lenses. The shape of the confidence regions also differs from those derived from observations of the CMB (Larson et al. 2011), supernovae (Sullivan et al. 2011), or BAOs (Percival et al. 2010; Beutler et al. 2011), providing interesting complementarity with these other methods.

4.3. Cosmic Sample Variance and Systematics

4.3.1. Basic Noise Sources

There are several basic sources of noise in our measurement of \( \Gamma(x) \). Individual galaxy shapes are noisy estimators of the tangential shear, and shape measurements themselves are not perfect, particularly for small or faint galaxies. The redshift errors in our lens population are negligible (\( \sim 0.0017 \) on average), but the photometric redshift errors for the source population may contribute significantly to the errors, once again particularly for faint galaxies at high redshift. These statistical errors are included, at least approximately, in our estimates of \( \chi^2 \). Our likelihood estimator also assumes a particular form for the average radial density profile of the groups \( \Sigma_0 \). This profile may not correspond exactly to the true mean surface mass density contrast, and individual groups may deviate from the average in ways that depend systematically on redshift. The tests described in Section 3.2 suggest that these effects are at or below the level of our statistical errors.

The relative importance of the source or lens sample selection is less clear a priori. To explore the dependence of our results on source galaxy properties, we have rerun parameter constraints with samples cut in magnitude range, S/N, size, and redshift error. In each case reducing the sample size increases the final error in \( \Omega_\Lambda \), but there is no obvious systematic trend for cuts in magnitude, S/N, or size beyond the dependence on source redshift discussed in Section 4.3.2 below. The precision of the final results is quite dependent on the redshift cuts we make in the source sample and on cuts to the group sample; including all the groups in the full (rather than the restricted) sample increases the uncertainties by \( \sim 25\% \), including source galaxies with \( z_{\text{sec}} > 0 \) increases them by \( \sim 30\% \), and including sources with \( \Delta z/(1+z) > 0.05 \) doubles them. Examining the catalog, it seems that the particular sensitivity to redshift errors is related to sources at \( z = 2–3 \), which contribute strongly to the constraints due to the sensitivity function (cf. Figure 4). Cutting these objects out, or diluting their signal with many poorly determined redshifts, both reduce out final sensitivity quite strongly. We have also experimented with using more restrictive cuts on redshift errors or using a larger 10' aperture around each group; these do not improve our results significantly.

Finally, the relative importance of redshift errors is also slightly unclear in our method. We have attempted to estimate the contribution of photometric redshift errors to our total
error budget by adding scatter to the photometric redshifts in the source catalog. Adding scatter equal to 5× or 10× Δz (increasing (Δz/(1 + z)) to 0.09 or 0.18) increases our 68% CL bounds on ΩΛ by 12% and 30%, respectively. (We note that this is after making the usual initial cut on relative redshift error; including galaxies with large initial redshift errors increases the scatter much faster, as noted above.) This suggests that our current redshift errors contribute only 5%–10% of the total error budget. If so, this is a promising result for other surveys that are likely to have redshift errors larger than COSMOS.

4.3.2. Cosmic Sample Variance—Empirical Estimates

A more complicated source of systematic uncertainty in the measurement comes from structure along the line of sight. The COSMOS survey probes a relatively small field and several large structures are clearly visible in the galaxy redshift distribution below redshift z = 1. We can look for evidence of cosmic sample variance effects in various ways. One simple estimate of the importance of cosmic variance can be obtained by bootstrapping, splitting the sample into two or more disjoint sets. We can test for systematics by measuring how ΔPoissonian, the width of the 68% confidence region for ΩΛ, changes relative to the Poissonian expectation ΔPoissonian ∝ √n0 when we cut the sample from n0 sources down to n0. Splitting the field into four quadrants with roughly equal numbers of source galaxies, we find ΔPoissonian increases by a factor of 1.9–2.5, so this seems consistent with the Poissonian expectation. We note, however, that one quadrant gives a best-fit value of ΩΛ = 0 (albeit with a 68% uncertainty of +0.6) whereas the others give values of ∼0.9. Examining the lenses and source distribution in this quadrant, it is not immediately obvious whether specific structures produce this shift. This is a sobering lesson about the possible effects of cosmic sample variance.

Splitting our lenses into two groups by redshift, each with nL ∼ n0/2, we find that ΔPoissonian increases by a factor of 1.33 for zL > 0.4 and 2.4 for zL < 0.4. Thus the noise increases at a roughly Poissonian rate in the high-redshift sample, while in the low-redshift sample it increases much more quickly. Similarly, in a sample cut at zS > 1, ΔPoissonian increases by 1.33, while for zS < 1, it increases by 1.73. These results suggest the signal at low redshift is more prone to systematics. We have tested alternative weighting schemes that attempt to correct for the trend with source redshift. Down-weighting sources with zS < 1.5 by a factor of 0.5 or 0.33 reduces ΔPoissonian by 30% or 40%, respectively, but moves the peak of the likelihood 1.3σ or 1.7σ away from the WMAP7 value. Thus, it seems there is some trade-off between precision and accuracy in the redshift weighting. Clearly this subject requires further theoretical work, using realistic simulations of large-scale structure. For the moment, in the absence of an optimal weighting scheme motivated by theory, we choose not to apply either weighting to our final results.

One final concern is that the average line of sight to all of our shear measurements gives essentially identical constraints on ΩΛ (the peak value shifts by less than 1σ), so the method appears to be robust to any small offset of this kind.

4.3.3. Cosmic Sample Variance—Theoretical Prediction

We can also use the error description of Taylor et al. (2007) to estimate the effect on our measurements of structure along the line of sight. They calculate that the tangential shear induced by large-scale structure between the observer and two background galaxies at redshifts zL and zS introduces a covariance in shear measurements given by

\[ \text{var}[γ_{ij}] = \int_0^\infty \frac{ℓdℓ}{2π} C_{ij}^{γγ}(ℓ) \left( \frac{2[1 - J_0(ℓθ)]}{(ℓθ)^2} - \frac{J_1(ℓθ)}{(ℓθ)} \right)^2, \]

where we have integrated over a circular aperture of radius θ by multiplying by the Fourier transform of the aperture (the term in brackets), and C_{ij}^{γγ}(ℓ) is the tomographic cosmic shear power spectrum (Hu 1999).

In our case, we are only concerned with the auto-correlations between redshift bins with i = j. These will give an estimate of the excess variance added to our shear measurements by cosmic structure, as a function of source redshift. Figure 7 shows this error term calculated for an aperture of 6′ in a WMAP7 cosmology and assuming the COSMOS redshift distribution. We have calculated the shear error using 20 discrete bins in redshift between 0 < z < 2. Ideally we would use a continuous cosmic shear in this measurement, as described in Kitching et al. (2011), but since the error contribution we find here is small and smoothly varying, this approximation seems adequate.

We see that the extra error term is always less than our empirical shear dispersion σγ ≥ 0.25, and that it reaches a maximum of ∼6% of the empirical dispersion. This suggests that the contribution from cosmic shear is much smaller than the excess variance seen in the previous section, which may then be due to individual halos or to other systematics. Clearly more detailed simulations are needed to determine realistic cosmic sample variance errors for our particular technique, but these estimates reassure us that systematics do not completely dominate our current results.
Gravitational lensing is a versatile source of cosmological tests; it can be used to measure both the spectrum of matter fluctuations in the universe and also its geometry. One particularly simple geometric test measures the rate at which the signal grows behind a lens and uses this to determine empirically the relationship between angular diameter distance and redshift. Here we introduce a new form of this test, the “stacked shear ratio test,” which can be applied to large sets of individually weak lenses. By factoring out the source and redshift dependence from the expected amplitude of the lensing signal, we define a universal geometric dependence \( \Gamma(x) \equiv (1 - 1/x) \), where \( x \equiv \omega_S/\omega_L \) and \( \omega_S, \omega_L \) are the comoving distances to the source and lens, respectively, which can be estimated from individual measurements of tangential shear. As we vary cosmological parameters, the goodness of fit of our data with respect to the universal form provides a constraint on the geometry or equation of state of the universe. Physically, the test measures the relative distance scale behind the lenses; a higher value of \( \Omega_X \) will increase this scale, so we can get particularly sensitive upper bounds on the density of dark energy.

The new method requires a significant lensing signal, although the signal can be spread over many individual lenses, and a large set of sources with well determined redshifts. The X-ray-selected group sample from the COSMOS survey provides a good test of the method, although cosmic sample variance is a concern over a field this small. Stacking groups with well-determined centers, we find a radial surface mass density contrast in good agreement with theoretical predictions. Examining the mean tangential shear behind the stacked group sample as a function of \( x \), we see a clear manifestation of the geometric signal.

Calculating goodness of fit for flat cosmological models with matter and a cosmological constant, with optimal weighting for signal.

5. CONCLUSIONS

The new method is a concern over a field this small. Stacking groups with well-determined centers, we find a radial surface mass density contrast in good agreement with theoretical predictions. Examining the mean tangential shear behind the stacked group sample as a function of \( x \), we see a clear manifestation of the geometric signal.

Calculating goodness of fit for flat cosmological models with matter and a cosmological constant, with optimal weighting for sensitivity to cosmology over a broad range of \( \Omega_X \), we obtain the constraints \( \Omega_A \equiv \Omega_X (w = -1) = 0.848^{+0.0435}_{-0.187} \) (68.2% CL). Extended to dark energy models with constant equation-of-state parameters \( w \neq -1 \), our results do not constrain \( w \) from below, but give the upper bound \( w < -0.4 \) (68.2% CL). The shape of the confidence regions differs from those derived from observations of the CMB, supernovae, or BAOs, providing interesting complementarity with these other methods. Systematic uncertainties and the effect of cosmic sample variance are hard to estimate accurately without more detailed simulations, but the simple tests presented in Section 4.3 suggest they do not dominate our statistical errors. Overall, the consistency of our results with more accurate determinations of \( \Omega_A \) (Larson et al. 2011) provides a good demonstration of the potential of the stacked shear ratio test in larger-area surveys.

Our method does have several limitations. The simple scaling only applies for flat cosmologies; for \( k \neq 0 \) the geometric term in the lensing equation depends separately on \( z_S \) and \( z_L \), so a more complicated two-dimensional fit is required. We have also worked exclusively in the weak shear limit where \( k \sim \gamma \ll 1 \). This does not restrict us significantly for COSMOS groups, since they are in the weak regime well into their central regions where other effects (e.g., baryonic mass, centering errors) become important anyway. Applying our method to more massive clusters would require modeling the full shear, and thus introduce separate dependence on \( z_S \) and \( z_L \). We also assume a specific functional form for the surface mass density contrast, namely the projected NFW profile. In practice this form seems to be an excellent fit to the data, however, and therefore we would not expect our results to change significantly if we used instead of this an empirical profile based on the data alone. Finally, our lower bound on \( \Omega_X \) is sensitive to the average physical scale radius of the lenses; sensible priors and/or reduced weighting at small radii are necessary here to avoid unphysical solutions which degrade the lower bound on \( \Omega_X \).

For the large samples of relatively low-mass lenses such as those expected from current (e.g., CFHliens\(^{16}\)) and forthcoming lensing surveys (e.g., DES;\(^{17}\) PanStarrs;\(^{18}\) LSST;\(^{19}\) HSC;\(^{20}\) Euclid;\(^{21}\) see Massey et al. 2010 for a full listing), our method provides a simple, intuitive measure of geometry. The statistical uncertainties on \( \Omega_A \) obtained for the 1.64 deg\(^2\) COSMOS field are only 4 × larger than those obtained from current CMB results (Larson et al. 2011) and should scale as (survey area)\(^{1/2}\). Thus if larger shear surveys can overcome systematics, there are good prospects for significantly reducing the uncertainties on abundance of dark energy. Given its sensitivity to sources at high redshift, our method might profitably be adapted to lensing measurements based on magnification (Van Waerbeke et al. 2010; Hildebrandt et al. 2011; Heavens & Joachimi 2011). Since these require only measurements of magnitudes, not shapes, they may provide larger samples of galaxies at \( z = 2–3 \). It could also be combined with CMB lensing (Hu et al. 2007; Das & Spergel 2009), which would provide an anchor point at the largest observable values of \( x \). In the longer term, as larger data sets become available and the sensitivity of lensing tests increases, their most interesting cosmological application might be to search for changes in the equation of state at particular redshifts. From this point of view, the method proposed here is flexible, as the redshift sensitivity depends simply on the distribution of source and lens redshifts used, and thus it can provide constraints over a range of redshifts, depending on the samples used.

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17 http://www.darkenergysurvey.org
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Berriman, and John Good) in providing online archive and server capabilities for the COSMOS data sets.

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