Constraining the Mass of the Emerging Galaxy Cluster SpARCS1049+56 at $z = 1.71$ with Infrared Weak Lensing

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Abstract

In the hierarchical structure formation model of the universe, galaxy clusters are assembled through a series of mergers. Accordingly, it is expected that galaxy clusters in the early universe are actively forming and dynamically young. Located at a high redshift of $z = 1.71$, SpARCS1049+56 offers a unique look into the galaxy cluster formation process. This cluster has been shown to be rich in cluster galaxies and to have intense star formation. Its high redshift pushes a weak-lensing analysis beyond the regime of the optical spectrum into that of the infrared. Equipped with deep Hubble Space Telescope Wide Field Camera 3 UVIS and IR observations, we present a weak-lensing characterization of SpARCS1049+56. As few IR weak-lensing studies have been performed, we discuss the details of point-spread function modeling and galaxy shape measurement for an IR weak-lensing procedure and the systematics that come with the territory. It will be critical to understand these systematics in future weak-lensing studies in the IR with the next-generation space telescopes such as the James Webb Space Telescope, Euclid, and WFIRST. Through a careful analysis, the mass distribution of this young galaxy cluster is mapped and the convergence peak is detected at a $3.3\sigma$ level. The weak-lensing mass of the cluster is estimated to be $3.5 \pm 1.2 \times 10^{14} M_\odot$ and is consistent with the mass derived from a mass–richness scaling relation. This mass is extreme for a cluster at such a high redshift and suggests that SpARCS1049+56 is rare in the standard $\Lambda$CDM universe.

Unified Astronomy Thesaurus concepts: Weak gravitational lensing (1797); Observational cosmology (1146); Dark matter (353); Galaxy clusters (584); High-redshift galaxy clusters (2007)

1. Introduction

Galaxy clusters emerged from the largest overdensities in the primordial universe. Their evolution is sensitive to both the growth rate of structure and the expansion history of the universe. For this reason, they are a useful probe to test cosmological theories. The observed size, mass, and abundance of galaxy clusters are a valuable tool to constrain the parameters that formulate our cosmological models. In particular, the abundance of galaxy clusters is a sensitive probe of the matter density $\Omega_m$ and the normalization of the matter power spectrum $\sigma_8$ (e.g., Gladders et al. 2007). However, the strong degeneracy between $\sigma_8$ and $\Omega_m$ prevents the constraint of each parameter independently with the cluster mass function alone. This degeneracy can be alleviated by combining cluster mass functions over a wide range of redshift (e.g., Albrecht et al. 2006).

A dominant systematic uncertainty in using galaxy clusters as cosmological probes is their mass calibration. Many of the large studies of galaxy clusters estimate mass through scaling relations, such as velocity dispersion or X-ray temperature, that rely on equilibrium or quasi-equilibrium state assumptions. The systematic errors innate to the mass estimate are then inherited by the cosmological constraint. Weak lensing (WL) provides a mass estimation free of an assumption of the dynamical state of the cluster and has the ability to provide more robust mass estimates. This merit is particularly important for galaxy clusters at high redshift where they tend to be in an early stage of formation and thus subject to a large departure from dynamical equilibrium.

To date, very few high-redshift galaxy clusters have been measured with WL. The vast majority of WL surveys have been focused on redshift less than unity. In fact, of the large WL surveys, the Hubble Space Telescope (HST) studies of Jee et al. (2011) and Schrabback et al. (2018) are the only to include clusters at $z > 1$. Beyond a redshift of 1.5, only a single galaxy cluster has been studied with WL, IDCS J1426+3508 (Mo et al. 2016; Jee et al. 2017) at redshift 1.75. The lack of studies at high redshift can primarily be attributed to the difficulty of detecting the lensing signal. The lensing distortions are caused by a massive intervening object between source galaxies and the observer. When a high-$z$ cluster is the lens, more distant galaxies need to be probed to detect the lensing signal. This requires very deep imaging at infrared wavelengths to robustly detect galaxies in the 25th–28th mag range. Fortunately, some imaging programs with the HST are probing these depths of the universe.

One of the goals of the See Change program (PI: Perlmutter) is to probe the WL mass function of galaxy clusters at redshift greater than one. The See Change sample includes 11 galaxy clusters in the redshift range 1.10–1.75, with IDCS J1426+3508 being the highest. The second highest redshift cluster in the sample is SpARCS1049+56 (hereafter SpARCS1049 for brevity) and it is the focus of this study.
SpARCS1049 was discovered in the Spitzer Adaptation of the Red-sequence Cluster Survey (SpARCS; Muzzin et al. 2009; Wilson et al. 2009). This survey utilized a two IR filter system to detect galaxy overdensities by the 4000 Å break (Wilson et al. 2006). The survey footprint included 11 deg^2 of the Lockman Hole, a 59 deg^2 region that is relatively clear of galactic HI emission, and within this region lies SpARCS1049.

The first detailed study of SpARCS1049 was achieved by Webb et al. (2015). They used the archival Spitzer observations and supplemented them with their own observations of the cluster from the James Clark Maxwell Telescope, HST, and Keck. Their Keck-MOSFIRE spectroscopy determined the galaxy overdensity redshift to be centered at z = 1.709. Based on this redshift, they classified 27 cluster member galaxies as those within 1500 km s\(^{-1}\) and 1.8 Mpc projected distance of the brightest cluster galaxy (BCG). The velocity dispersion (\(\sigma = 430\pm100 \text{ km s}^{-1}\)) of these galaxies provides a mass \(M_{\text{vir}}\) of (8 \pm 3) \times 10^{13} M_\odot. The authors go into detail about the shortcomings of the velocity dispersion from this sample. Their classification of cluster member galaxies goes well beyond the expected virial radius of the cluster (~1 Mpc). Furthermore, the redshifts were detected by the Hα emission line, which only selects active galaxies. In addition to this mass estimate, they found the richness of the cluster to be \(N_{\text{gal}} = 30 \pm 8\) and used the mass–richness scaling relation from Andreon & Congdon (2014) to infer a mass \(M_{500,\text{kpc}}\) of 3.8 \pm 1.2 \times 10^{14} M_\odot.

We present a WL characterization of SpARCS1049 through the HST Wide Field Camera 3 (WFC3) IR filters. The mass estimate from WL is an independent test of the previous two versions of the Red-sequence Cluster Survey (Spitzer; Muzzin et al. 2014 February to 2015 May. In both programs the cluster was located in the center of the image with the “beads-on-a-string” interacting galaxy stretching from east of the BCG to ~50 kpc southwest. These features are more obvious in the zoomed inset. For more on the galaxies of SpARCS1049 see Webb et al. (2015, 2017) and Trudeau et al. (2019).

We chose to tune the input parameters of Multidrizzle to optimize the F160W image quality as it is used for our lensing analysis. The FWHM of the PSF in the IR detector is fractionally larger (FWHM \(\sim 0.16\)) than the native pixel scale \(0.13\). This causes undersampling of the PSF. The DrizzlePac handbook (Gonzaga et al. 2012) suggests that upsampling to a final pixel scale that samples the PSF by about 2.0–2.5 pixels is ideal. Following this advice, we chose a final pixel scale of \(0.05\) pix\(^{-1}\) to mitigate the effect of undersampling the PSF. Although this pixel scale is larger than the UVIS native pixel scale of \(0.04\), the downsampling F814W images are strictly used for color image generation and not in the scientific analysis. We set final_pixfrac to 0.7 and used a Gaussian kernel to drizzle the images. The color-composite image in Figure 1 was created by combining the F160W, F105W, and F814W filter images. The BCG is the deep orange galaxy located in the center of the image with the “beads-on-a-string” interacting galaxy stretching from east of the BCG to ~50 kpc southwest. These features are more obvious in the zoomed inset. For more on the galaxies of SpARCS1049 see Webb et al. (2015, 2017) and Trudeau et al. (2019).

We created a detection image by weight-averaging the F105W and F160W images with weights from Multidrizzle. Objects were detected with Source Extractor\(^8\) (Bertin & Arnouts 1996) in dual-image mode by selecting sources in the

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\(^8\) https://archive.stsci.edu/
\(^9\) https://www.astromatic.net/software/sExtractor
detection image and measuring them in each filter-specific image (F105W or F106W). Objects that subtend at least 5 pixels having signal at least 1.5σ above the background rms were measured. WL studies using the HST have shown that the background galaxy density is high (∼100 galaxies arcmin⁻²). An issue that arises with high source galaxy density is blending (overlapping) of galaxy images. Mandelbaum (2018) discusses the bias arising from deblending galaxies in detail. The primary concern for deblending in this study occurs when two images overlap from galaxies at large separation in redshift. To mitigate the effect, we deblended objects using Source Extractor with DEBLEND_NTHRESH = 8 and DEBLEND_D_MINCONT = 0.005. However, such rigorous deblending can cause a foreground galaxy to be deblended into multiple objects. These spurious detections, which contain no lensing signal, were removed after source selection (Section 3.3) through visual inspection. In total, ∼6900 objects were detected in the ∼3′ × 3′ WFC3-IR mosaic image and compiled into an object catalog.

2.1. PSF Model

Ground-based WL analyses rely heavily on the correction of the PSF as it causes a significant dilution of the observed lensing shear. In addition, the PSF tends to have a characteristic direction that mimics shearing. These two PSF effects are also present in the space-based HST imaging but to a lesser extent because of the lack of atmosphere.

This is the first study to use the WFC3-IR/F160W channel for a WL analysis. We modeled the PSF using a version of our PSF modeling pipeline based on principal component analysis and updated for the F160W channel. This pipeline has been described in detail in our previous papers (e.g., Jee et al. 2007a; Finner et al. 2017). Here we will briefly explain the PSF pipeline for the WFC3-IR/F160W channel and refer the reader to our previous work for an in-depth discussion.

A major hurdle for modeling the PSF of HST, which depends on time and position on the focal plane, is the lack of stars available in a single science frame. Fortunately, the HST PSF variation possesses a repeatable pattern (Jee et al. 2007a) that is dependent on the focus (breathing) of the telescope following the 1.5 hr orbit of the telescope around Earth. This allows a utilization of dense archival stellar images to model the PSF, which can then be applied to the science frames that are taken at a different epoch. Table 1 contains a list of the dense stellar fields that we tested for our PSF modeling pipeline. In the majority of these fields, the frames are overcrowded with stars and overlapping diffraction spikes significantly hamper our
ability to model the PSF. However, the exposures of NGC 104 (also known as 47 Tuc) and NGC 2808 in programs 11453, 11664, and 11665 contain the best spatial sampling to characterize the PSF and we relied on these frames for our PSF pipeline. These images were drizzled with the same settings as the single-drizzled science images (Section 2) and will be referred to as the stellar frames from here on. We ran our PSF modeling pipeline on the stellar frames and designed a position-dependent PSF for each frame.

Switching to the science frames, we selected several stars (5 ~ 10) from each single-drizzled science image and recorded their pixel coordinates and ellipticity. At the coordinates of these stars, we retrieved the modeled PSF for each stellar frame. This resulted in a catalog of PSFs for each stellar frame at the defined science frame’s star locations. To find the best-fit model stellar frame, we minimized the difference between the ellipticity of the modeled PSFs and science stars. The median reduced χ² value is 1.8 for the best-fit models. Furthermore the residual ellipticities when comparing our model to the measured stellar ellipticities are ∆e ~ 0.008, which is sufficient for cluster lensing. Finally, PSFs for all objects in the F160W mosaic image were built by retrieving the best-fit PSF model at each object location for each science frame and stacking them into a final PSF.

2.2. WFC3-IR Detector WL Systematics

Systematic effects inherent to the IR detector are a cause for concern for WL studies because they may falsely contribute to the WL signal. In the first WL analysis to use the WFC3-IR detector, Jee et al. (2017) reported four systematic effects that need to be considered: interpixel capacitance (IPC), persistence, detector nonlinearity, and undersampling. Readers are referred to Jee et al. (2017) for detailed discussions on these four topics from a WL perspective. Here, after briefly describing these aforementioned effects, we will provide a detailed discussion on the brighter–fatter effect.

IPC. The WFC3-IR detector is a 1024 × 1024 HgCdTe array with a plate scale of 0’013 per pixel. Brown et al. (2006) investigated the correlated noise in HgCdTe detectors and found charge sharing between neighboring pixels from capacitive coupling. This IPC is also present in the WFC3-IR detector (Hilbert & McCullough 2011). We follow the same method as Jee et al. (2017) and let our PSF model correct for IPC.

Persistence. IR detectors are also susceptible to a persistence of signal after a reset. The effect is described in detail in Smith et al. (2008). Their investigation showed that the persistence of charge is greater for pixels that have been exposed near saturation in previous imaging. The STScI provides a tool to search for persistence in archived observations. Our search shows that persistence levels are low in observations of SpARC1049, with a persistence of ≥0.01 e− s−1 in at most 0.1% of the pixels and ≥0.1 e− s−1 for 0.03% of the pixels.

Undersampling. The FWHM of the WFC3-IR F160W PSF is approximately the same size as the native plate scale (0’013 pixel−1), which causes signals to not be Nyquist sampled. As a first step to alleviate undersampling, a dithering of the individual exposures was done during observations. Combining the dithering technique with upsampling during drizzling allows us to catch some of the sampled details of the PSF. As done in our previous IR WL analysis (Jee et al. 2017), we let our calibration of galaxy shapes take care of the remaining undersampling bias.

Nonlinearity. The response of the WFC3-IR detector follows a nearly linear relation until close to saturation where it then becomes nonlinear. Nonlinearity in the detector was reported at the 5% level for saturated pixels (Dressel 2018). The calwf3 pipeline corrects the detector nonlinearity for pixels below the saturation level. As a precaution, we selected stars that are well below the saturation level when modeling the PSF.

Brighter–fatter. Analyzing the size–magnitude relation of the stellar frames that were used to model the PSF, we found a slope to the stellar locus with brighter objects tending to be larger. The brighter–fatter effect is well studied in CCDs and is thought to be caused by the electric field from the charges that have been accumulated in a pixel (Antilogus et al. 2014; Guyonnet et al. 2015). For CCDs, Antilogus et al. (2014) report that the size of the PSF increases by 2% over the full dynamic range. However, few studies (Plazas et al. 2017, 2018) have been carried out on the brighter–fatter effect in IR detectors.

The brighter–fatter effect requires attention for WL analyses because it will introduce a multiplicative bias to the measured shear. This is especially important for the faint galaxies that carry the WL distortion where forward-modeling an overly large PSF may lead to an overestimation of the shear. Mandelbaum (2015) showed that a 1% inflated PSF size introduces a systematic bias of m = 0.06 for a galaxy near the resolution limit. Our analysis of the stellar locus in the NGC 104 frames shows that the average size of stars varies by as much as 5% from the faintest detected objects to the saturation magnitude of the detector. In our PSF modeling, we intentionally avoid the stars near saturation. Thus, 5% should be taken as an upper limit. Nevertheless, we desire to understand the systematic bias that might be introduced when forward-modeling a PSF with a size up to 5% larger than the true PSF size. To do so, we simulated our forward-modeling shape measurement using GalSim (Rowe et al. 2015).

Simulated images of 10,000 Sérsic profile galaxies (100 × 100 equally spaced on a grid) were created with the Sérsic parameters sampled from the real galaxies of SpARC1049. A uniform shear typical of a galaxy cluster (∼0.05) was applied to the images. These simulated galaxy images were then convolved with a circular Gaussian PSF. Multiple passes of our shape-measurement pipeline were performed while forward-modeling PSFs of size ranging from −15% to +15% of the true PSF size. This experiment showed

| Object     | Program ID | Exposures | Obs. Year |
|------------|------------|-----------|-----------|
| 47 Tuc     | 11453      | 18        | 2009      |
| NGC 104    | 11664      | 6         | 2010      |
| NGC 2808   | 11665      | 6         | 2011      |
| NGC 6388   | 11739      | 10        | 2010      |
| NGC 6441   | 11739      | 20        | 2010      |
| OmegaCen   | 11928      | 27        | 2009      |
| OmegaCen   | 12353      | 15        | 2011      |
| OmegaCen   | 13691      | 6         | 2015      |

Note. Boldface programs were selected for PSF modeling.
that the multiplicative bias varies by $m = 0.02$ for a 5% change in PSF size. At this level, the brighter–fatter effect has a low impact on galaxy cluster studies where shape noise is still the dominant uncertainty. However, in cosmic shear studies the brighter–fatter effect will need to be addressed.

3. Weak-lensing Method

3.1. Theory

At the core of weak gravitational lensing studies is the measurement of the minute distortion of galaxies. In the context of SpARCS1049, these distortions are caused by the altered light path that a photon travels while crossing the gravitational potential of the galaxy cluster. The altered light path can be described by its deflection angle—the angle between its original path away from its galaxy to its new path toward our telescope. The deflection angle is the gradient of the deflection potential. The differential transformation from the photon’s emission position to the observed position is described by the Jacobian matrix:

$$A = \begin{pmatrix} 1 - \kappa - \gamma_1 & -\gamma_2 \\ -\gamma_2 & 1 - \kappa + \gamma_1 \end{pmatrix},$$

where the convergence $\kappa$ is an isotropic distortion defined as

$$\kappa = \frac{\Sigma}{\Sigma_c}.$$  

In Equation (2), $\Sigma$ is the projected mass density, while $\Sigma_c$ is the WL critical surface density:

$$\Sigma_c = \frac{c^2}{4\pi GD_l \beta^2},$$

where $c$ is the speed of light, $G$ is the gravitational constant, $D_l$ is the angular diameter distance of the lens, and $\beta = D_{ls}/D_s$ is the lensing efficiency, which is the lens to source over source angular diameter distances. In Equation (1), the shear $\gamma$ is an anisotropic distortion and its two components can be combined to formulate the complex shear, $\gamma = \gamma_1 + i\gamma_2$. Observationally, the two distortion effects cannot be separated and the observed effect is the reduced shear $g_i = \gamma_i/(1 - \kappa)$.

Without prior knowledge of the shape (ellipticity) of each galaxy image, measurement of $g$ directly based on a single galaxy image is not possible. Instead, the average complex ellipticity of an ensemble of galaxies is used to find $g$. This is done under the assumption that the average galaxy ellipticity is zero. We adopt the value of $\sigma_{\text{int}} = 0.25$ for the intrinsic ellipticity dispersion, a value recently confirmed with the CANDELS data in Schrabback et al. (2018). This value of the intrinsic ellipticity dispersion is used in inverse-variance weighting when fitting models for mass measurement (Section 4).

3.2. Shape Measurement

The WL observable, the reduced shear $g$, is ascertained by averaging the shapes of source galaxies. Our method of shape measurement is to fit a PSF-convolved elliptical Gaussian function to each object in the source catalog (source catalog defined in Section 3.3).

Postage-stamp images of each object are cut from our F160W mosaic image. The size of each postage-stamp image is chosen to be 12 times the semimajor axis of the object as determined by Source Extractor. This size reduces the effect of truncation bias that occurs when the light profile is prematurely truncated. However, a large postage-stamp image increases the number of neighboring objects whose signal may contaminate the fit. We mask out the signal of the neighboring objects using the segmentation map output from Source Extractor. The difference between the light profile of the postage-stamp image and the PSF-convolved elliptical Gaussian model is minimized with MPFIT (Markwardt 2009). We fix the centroid and background levels to the measurements from Source Extractor to reduce the free parameters of the fit. From the MPFIT output, we catalog the two complex ellipticity components

$$e_1 = \frac{a - b}{a + b} \cos 2\phi,$$

$$e_2 = \frac{a - b}{a + b} \sin 2\phi,$$

where $a$ and $b$ are the semimajor and minor axes of the ellipse, respectively, and $\phi$ is the angle measured counterclockwise from the positive $x$-axis. The ellipticity error $\sigma_e$ is also included in the catalog.

Measuring a galaxy’s shape by fitting the light profile with an analytic function that does not perfectly represent the light profile introduces model bias. Moreover, the nonlinear relation of the ellipticity measurement with the pixel noise causes noise bias. We correct for these biases by calibrating the ellipticities with a multiplicative factor of 1.25 that is derived through simulations. Our method has been shown as effective by the sFIT method in the GREAT3 challenge (Mandelbaum et al. 2015).

3.3. Source Selection

Selecting the source galaxies is an intricate step of a WL analysis. The lensing signal is observable only in the galaxies that are sufficiently behind the lens. Selection of source galaxies by spectroscopic or photometric redshift would be ideal but obtaining them is expensive and currently not possible with the limited $HST$ filter coverage. Instead, we select source galaxies based on their measured shape and photometric properties.

Galaxies residing in a cluster tend to be redder than field galaxies. The 4000 Å break, caused by the absorption of stellar light by ionized metals in stellar atmospheres, is a common feature in cluster galaxies and often gives rise to a red sequence in a color–magnitude diagram (CMD). For SpARCS1049, the 4000 Å break is redshifted to $\sim 10800$ Å. This wavelength is encapsulated in the F105W and F160W filters. Figure 2 shows the CMD for SpARCS1049 with black dots representing the full object catalog. Cluster member galaxies selected from the Keck spectroscopic observations within $1.67 < z < 1.75$ and within the $HST$ imaging footprint are shown as red circles. These spectroscopic redshifts are detected from the Hα emission line and give an active-galaxy selection bias to our cluster member sample. The BCG is shown as a red star and has a large magnitude separation from the other cluster members. The lack of a clear red sequence suggests that SpARCS1049 may be in an early state of formation.

A pure source catalog is one that only contains lensed galaxies. Cluster member galaxies and foreground objects in the source catalog will contaminate the sample and dilute the
lensing signal. Removal of these false sources is challenging without precise distances to each. Unfortunately, most removal techniques also filter out some true source galaxies. This is a problem because the lensing signal is proportional to the purity of the sample, whereas the noise is proportional to \(1/\sqrt{N}\). Furthermore, the uncertainty of the lensing efficiency \(\beta\) increases with a decreasing number of sources. Methods to maximize purity and source counts in the catalog vary. As a first step in defining a source catalog, we exclude foreground galaxies with an apparent magnitude cut that is fainter than the faintest spectroscopically confirmed cluster member. Our signal-to-noise ratio (S/N) tests show that retaining galaxies of F160W magnitude >25 provides the highest S/N. Including brighter galaxies decreases the detected WL signal and subsequently the S/N.

In WL, sampling the faintest galaxies is desired because the most distant source galaxies are subject to the greatest lensing distortions. However, fitting a model to a low S/N galaxy is difficult and is subject to noise bias. To decrease noise bias, we exclude galaxies with a measured ellipticity error greater than 0.25. This constraint causes the faint magnitude limit seen in the CMD. In addition, galaxies in the source catalog are constrained to have a semiminor axis greater than 0.3 pixels and ellipticity less than 0.9 to remove objects that are too small or too elongated to be galaxies. The total galaxy number density in our source catalog is \(\sim 105\) galaxies arcmin\(^{-2}\).

To test the source catalog for contamination by cluster galaxies, we analyze the radial variation of source density. In Figure 3, the radial source density is shown with radial bins centered on the BCG. Contamination by cluster galaxies could manifest as an overdensity near the cluster central region. The flat profile suggests that cluster member contamination is a minimum. The dashed vertical line represents the radius at which a circle is no longer complete in the HST image. Error bars are Poissonian errors. Blue circles are the radial number density weighted by shape-measurement ellipticity error \((1/\sigma_{\text{int}}^2 + \sigma_e^2)\).

Density slightly decreases. This decrease is likely due to the limited frame coverage near the edge of the mosaic image and from the bright foreground galaxy drowning out background galaxies in the northern region of the image.

### 3.4. Source Redshift Estimation

As shown in Equation (2), the WL signal is proportional to the lensing efficiency, \(\beta\). A proper characterization of \(\beta\) relies on accurate knowledge of the angular diameter distances to the galaxy cluster and the source galaxies. However, the limited filter coverage for SpARCS1049 prevents direct calculation of distances to the source galaxies. As an alternative, we use the UVUDF photometric redshift catalog (Rafelski et al. 2015) as a control field, model it to represent our source catalog, and infer a representative distance to the source galaxies.

We constrain the UVUDF catalog with the same magnitude constraint specified in Section 3.3. A comparison of the number density of galaxies in the source catalog and the UVUDF catalog is shown in Figure 4. The number density of galaxies in the two catalogs is consistent in the 25th–26th mag range. After the 26th mag the number density discrepancy can be attributed to the much deeper imaging of the UVUDF. To make the UVUDF catalog representative of our source catalog, we weight the UVUDF control catalog by the ratio of UVUDF to SpARCS1049 galaxy number density. The effective redshift and corresponding \(\beta\) is calculated from the weighted UVUDF catalog as

\[
\beta = \max \left(0, \frac{D_s}{D_l}\right),
\]

where all foreground galaxies are assigned zero before averaging because they contain no lensing signal. From the weighted UVUDF catalog, we infer an effective redshift of 2.08 and \(\beta = 0.107\) for our source catalog. Bias is introduced when representing all source galaxies by a single redshift. We reduce the bias as suggested in Seitz & Schneider (1997) by
The UVUDF redshift catalog is used as a control. The magnitude distribution of galaxies in the SpARCS1049 WL source catalog. In the 25th–26th mag range, the completeness in the SpARCS1049 and UVUDF catalogs are consistent. The discrepancy in the 25th mag arises from the vastly different exposure times between our cluster imaging and the UVUDF. We compute the error bars assuming Poisson distributions. Taking the width of the beta distribution, \( \langle \beta^2 \rangle = 0.03 \) into consideration. One may question whether the \( \beta \) derived from a small field such as the UVUDF is representative of the small field of SpARCS1049. Jee et al. (2014) compared the UVUDF to the UDF, GOODS-S, and GOODS-N redshift catalogs and found comparable \( \beta \) values for each catalog. They reported the uncertainty of \( \beta \) values between catalogs to affect mass estimates by at most \( \sim 4\% \). This small sample variance is attributed to the great depth of the HST image, which provides access to large distances along the line of sight. Adding this uncertainty to the statistical uncertainty (\( \sim 25\% \)) in quadrature shows that the statistical uncertainty on the mass will be dominant.

4. Results

4.1. Mass Reconstruction

A powerful aspect of WL is its ability to measure the projected mass distribution of the lens with minimal assumptions. There are multiple techniques that can be used to convert the observed shear \( g \) to the convergence \( \kappa \). We rely on the MAXENT method of Jee et al. (2007b), which converges to a solution that maximizes the entropy of a pixelized mass map while providing a reasonable goodness-of-the-fit for galaxy shapes.

Figure 5 is the convergence map for SpARCS1049. The convergence is smoothed with a \( \sigma = 10^\prime \) Gaussian kernel to remove a pixelization artifact. The convergence shows a slight elongation in the east–west direction but in general has a relaxed distribution for the applied smoothing scale. The mass peak lies \( \sim 10^\prime \) (\( \sim 90 \) kpc) to the southwest of the BCG. This offset, if significant, could be interpreted as an indication that the cluster mass is not centered at the BCG. To test the significance of the offset and the strength of our WL signal, we bootstrap the source catalog 1000 times. From the bootstrapped samples, we find that the cluster is detected at the 3.3\( \sigma \) significance. The resampled catalogs also reveal that the 1\( \sigma \) uncertainty of the convergence peak location is \( \sim 13^\prime \). Thus, we conclude that the mass map shows no statistically significant offset from the BCG.

4.2. Mass Estimation

Accurate estimation of the cluster mass is the primary goal of this work. There are numerous techniques that can be used to estimate the mass of a cluster from the observed galaxy ellipticity distribution. We choose to estimate the mass by fitting model profiles to the azimuthally averaged tangential shear.

The reduced tangential shear \( g_T \) at radius \( r \) is a measure of the surface density contrast between the mean value within \( r \) and the specific value at \( r \) divided by \( 1 - \kappa \). It is often written as the tangential components of the complex shear \( g \):

\[
g_T = -g_1 \cos 2\theta - g_2 \sin 2\theta
\]

where \( \theta \) is the angle measured from the center of the cluster to the source, counterclockwise. Rotating \( \theta \) by 45\( ^\circ \) gives the cross shear, which should be consistent with zero in the presence of no systematic effects and a circularly symmetric projected mass distribution. Figure 6 is the tangential shear measured in 10\( ^\prime \) bins centered on the BCG. The tangential shear profile is sensitive to the choice of center, particularly at small radii. We center our shear measurements on the BCG because it is consistent with the lensing peak and is an independent tracer of the cluster center. We also tested using the convergence peak as the center of the tangential shear fit and found that the derived mass is consistent with using the BCG as the center. The tangential shear profile clearly shows the detection of the lensing signal and the cross shear is consistent with zero. The outer limit of the tangential shear profile is set by the edge of the mosaic image and the data points beyond 60\( ^\prime \) are affected by the bright galaxy in the north.

To estimate the mass, we fit 1D density models to the tangential shear as shown in Figure 6. The first density profile that we fit is the singular isothermal sphere (SIS). The SIS profile returns a fitted velocity dispersion of \( \sigma_v = 833 \pm 84 \) km s\(^{-1}\). Many density profiles have been derived from cosmological simulations that would all be appropriate to fit to the tangential shear. We fit some of the popular NFW-based models (Navarro et al. 1997) to our tangential shear so that direct comparison can be made with published galaxy cluster lensing masses. These fits are done by assuming the tangential shear profile follows a fixed concentration–mass (c–M) relation and fit only mass \( M_{200}\). We utilize the Colossus code (Diemer 2018) when performing the fits. The masses are summarized in Table 2. All three models return consistent masses. However, not all c–M relations should be considered equal. As explained in detail in Diemer & Joyce (2019), the c–M relation strongly depends on redshift and cosmology. Models that fit average concentration, such as Duffy et al. (2008) and Dutton & Macciò (2014), are only valid under the assumed cosmology and redshift range of the simulations they are extracted from. Furthermore, power-law fits do not capture the upturn at high redshift and high mass as is shown in Diemer & Kravtsov (2015). Ludlow et al. (2013) attribute the upturn to unrelaxed clusters. As SpARCS1049 is at a redshift of 1.71 and is likely in an early stage of formation, we suggest that the Diemer & Joyce (2019) model is a good choice for a reasonably high mass cluster. Furthermore, of the three c–M models...
models that we fit, the Diemer & Joyce (2019) model provides the best fit with reduced $\chi^2 = 1.03$. Throughout the discussion, we will use $3.5 \pm 1.2 \times 10^{14} M_\odot$ for our WL mass estimation.

5. Discussion

5.1. Mass Comparison with Previous Studies

A single previous study on the mass of SpARCS1049 exists. Webb et al. (2015) estimated the mass of SpARCS1049 through a mass–richness scaling relation and by velocity dispersion of cluster member galaxies. They determined the abundance of cluster galaxies from Spitzer 3.6 $\mu$m observations to be $30 \pm 8$. This returned a mass of $M_{500\text{kpc}} = 3.8 \pm 1.2 \times 10^{14} M_\odot$ from application of the mass–richness scaling relation of Andreon & Congdon (2014). The authors note that the $\sim 30\%$ uncertainty on this mass does not take into consideration any redshift evolution of the scaling relation.

To find the velocity dispersion of cluster galaxies, Webb et al. (2015) obtained Keck spectroscopic measurements. The classification of cluster galaxies by these observations relied on the detection of the H$\alpha$ line. As the authors noted, this biases the sample to emission galaxies. Nevertheless, they classified 27 cluster member galaxies within 1500 km s$^{-1}$ of the mean cluster redshift and within a 1.8 Mpc cluster-centric radius. The authors also mentioned that this included galaxies beyond the virial radius of the cluster. From the classified cluster members, the resulting velocity dispersion is $\sigma = 430^{+80}_{-60}$ km s$^{-1}$ and the inferred mass is $M_{\text{virial}} = (8 \pm 3) \times 10^{13} M_\odot$, after applying the velocity dispersion to virial mass relation of Evrard et al. (2008). The $\sim 40\%$ uncertainty reflects the unreliability of using strictly emission galaxies to derive the velocity dispersion. It is peculiar that the mass from velocity dispersion is much lower than from the mass–richness relation. It goes against the notion that emission galaxies should be infalling and have inflated velocity dispersion.

Using an updated spectroscopic redshift catalog, we selected member galaxies in the same manner as Webb et al. (2015). Applying the bi-weight velocity dispersion to the 27 detected members gives $\sigma_v = 446 \pm 80$ km s$^{-1}$ and $M_{\text{virial}} = 1.0 \pm 0.4 \times 10^{14} M_\odot$ with the conversion from Evrard et al. (2008). We attach the same 40% uncertainty on mass as Webb et al. (2015). This mass is consistent with the findings of Webb et al. (2015).

Figure 7 shows the histogram of cluster galaxies. Performing an
Blue circles are the tangential shear and black crosses are the cross shear. Error bars are the Poisson error. The cross shear has been shifted by 1 \(^{\circ}\) for display purposes. Three density profiles are shown: the SIS sphere, the \(c-M\) relation of Dutton & Macciò (2014), and the \(c-M\) relation of Diemer & Joyce (2019).

![Figure 6](image_url)

**Figure 6.** Density model fits to the tangential shear profile centered at the BCG. Blue circles are the tangential shear and black crosses are the cross shear. Error bars are the Poisson error. The cross shear has been shifted by 1 \(^{\circ}\) for display purposes. Three density profiles are shown: the SIS sphere, the \(c-M\) relation of Dutton & Macciò (2014), and the \(c-M\) relation of Diemer & Joyce (2019).

Table 2

| Model            | \(c_{200}\) | \(M_{200c} \times 10^{14} \, M_\odot\) | \(\chi^2_f\) |
|------------------|-------------|----------------------------------------|--------------|
| Duffy et al. (2008) | 2.2 ± 0.2   | 5.9 ± 3.5                              | 1.27         |
| Dutton & Macciò (2014) | 3.1 ± 0.1   | 4.5 ± 2.3                              | 1.16         |
| Diemer & Joyce (2019) | 4.5 ± 0.3   | 3.5 ± 1.2                              | 1.03         |

Anderson–Darling test on the cluster galaxies fails to reject the null hypothesis that they follow a normal distribution. Our WL mass estimate \(M_{200c} = 3.5 \pm 1.2 \times 10^{14} \, M_\odot\) provides the first mass estimate free of a dynamical equilibrium assumption. This mass estimate is consistent with the mass–richness estimation. However, there is a discrepancy with the mass from the velocity dispersion.

5.2. Rarity

Massive galaxy clusters at high redshift are expected to be rare according to the hierarchical structure formation model. SpARCS1049 was selected for this study because of its known large mass and should be tested for its rarity. Future work will fully analyze the rarity of the See Change sample of massive galaxy clusters between redshifts 1.10 and 1.75.

We determine the rarity of this cluster by integrating the number of clusters above a minimum mass and redshift as

\[
N(M, z) = \int_{z_{\text{min}}}^{\infty} \int_{M_{\text{min}}}^{M_{200c}} \frac{dV(z) \, dn}{dM} \, dM \, dz,
\]

where \(dV/dz\) is the volume element and \(dn/dM\) is the mass function. We set the lower limits of the integrals to \(z_{\text{min}} = 1.71\) and \(M_{\text{min}} = 3.5 \times 10^{14} \, M_\odot\), the central mass estimate. The exact upper limits of the integral are insignificant because the rarity of the cluster (steepness of the mass function in this regime) causes the integral to converge quickly. Using HMFCalc (Murray et al. 2013a), we adopt the mass function of Tinker et al. (2008) that has been updated by Behroozi et al. (2013). The estimated abundance of a cluster with mass and redshift of SpARCS1049 is \(\sim 12\) over the full sky or \(\sim 0.01\) clusters within the \(\sim 41.9 \, \text{deg}^2\) footprint of SpARCS. Alternatively, taking the \(1\sigma\) lower limit of our mass estimation result \(M_{\text{min}} = 2.3 \times 10^{14} \, M_\odot\) gives a rarity of \(\sim 185\) clusters in the entire sky or \(\sim 0.2\) in the SpARCS field. For comparison, the rarity of two additional See Change clusters, ICDS J1426+3508 (\(z = 1.75\)) and SPT-CL J2040-4451 (\(z = 1.48\)), are \(\sim 1200\) and \(\sim 1\) clusters in the full sky, respectively, using their WL measured central mass values (Jee et al. 2017). Thus, SpARCS1049 is similar in rarity to other See Change clusters.

This type of rarity calculation has well-documented limitations (Hotchkiss 2011; Hoyle et al. 2012; Harrison & Hotchkiss 2013). As pointed out by Hotchkiss (2011), the rarity integral only considers clusters that have mass and redshift greater than or equal to the selected lower limits. The rarity calculation neglects equally rare clusters that exist at higher mass but lower redshift and vice versa, which results in a bias that causes low rarities. Furthermore, the rarity calculation relies on integration of a mass function that is derived from cosmological simulations that often poorly reproduce the high-mass high-redshift end of the mass function. Murray et al. (2013b) report that the halo mass function has \(\sim 20\%\) uncertainty at the high-mass end. An additional limitation comes from Eddington bias (Eddington 1913; Mortonson et al. 2011). Eddington bias occurs because the mass function of the universe is steeply declining with increasing mass at the mass and redshift of SpARCS1049. Therefore, it is more likely to overestimate a cluster mass than to underestimate a cluster mass for such an extreme object.

6. Conclusions

An HST-IR WL analysis of the massive galaxy cluster SpARCS1049 is presented. HST-IR detector systematics have been quantified with a specific focus on the brighter–fatter effect. Our simulations show that the brighter–fatter effect...
gives at most a 2% shape bias in our shear measurements. The systematics discussed will be important for future WL studies with next-generation telescopes, such as the James Webb Space Telescope, Euclid, and WFIRST.

The projected mass distribution has been reconstructed from the averaged background galaxy ellipticities. The mass distribution is seemingly relaxed for the applied smoothing scale with the centroid consistent with the BCG. We have found the mass of the cluster to be $(3.5 \pm 1.2) \times 10^{14} M_\odot$ for our best-fit NFW model. This mass is consistent with the mass estimated from a mass–richness scaling relation. However, it is inconsistent with the mass from velocity dispersion of spectroscopically confirmed cluster galaxies. Finally, we have tested the mass of the cluster for its rarity. We have found the expected abundance of similarly massive clusters to be $<1$ within the parent survey, thus suggesting that SpARCS1049 is a uniquely massive cluster.

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