The SAMI Galaxy Survey: Mass as the Driver of the Kinematic Morphology–Density Relation in Clusters

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Abstract

We examine the kinematic morphology of early-type galaxies (ETGs) in eight galaxy clusters in the Sydney-AAO Multi-object Integral-field spectrograph Galaxy Survey. The clusters cover a mass range of 14.2 < log(M_200/M_☉) < 15.2 and we measure spatially resolved stellar kinematics for 315 member galaxies with stellar masses 10.0 < log(M_☉/M_☉) < 11.7 within 1 R_200 of the cluster centers. We calculate the spin parameter, λSR, and use this to classify the kinematic morphology of the galaxies as fast or slow rotators (SRs). The total fraction of SRs in the ETG population is F_SR = 0.14 ± 0.02 and does not depend on host cluster mass. Across the eight clusters, the fraction of SRs increases with increasing local overdensity. We also find that the slow-rotator fraction increases at small clustercentric radii (R < 0.3 R_200), and note that there is also an increase in the slow-rotator fraction at R < 0.6 R_200. The SRs at these larger radii reside in the cluster substructure. We find that the strongest increase in the slow-rotator fraction occurs with increasing stellar mass. After accounting for the strong correlation with stellar mass, we find no significant relationship between spin parameter and local overdensity in the cluster environment. We conclude that the primary driver for the kinematic morphology–density relationship in galaxy clusters is the changing distribution of galaxy stellar mass with the local environment. The presence of SRs in the substructure suggests that the cluster kinematic morphology–density relationship is a result of mass segregation of slow-rotating galaxies forming in groups that later merge with clusters and sink to the cluster center via dynamical friction.

Key words: galaxies: clusters: general – galaxies: elliptical and lenticular, cD – galaxies: evolution – galaxies: groups: general – galaxies: kinematics and dynamics

1. Introduction

The relative fraction of different galaxy morphological types has been shown to vary with environment such that galaxies visually classified as early-type galaxies (ETGs) are more prevalent in the high-density environment of galaxy clusters than in the low-density environment. This is the morphology–density relationship (Oemler 1974; Davis & Geller 1976; Dressler 1980). ETGs can also be classified based on a kinematic morphology using their spin parameter, probed through their stellar kinematics. In this classification system, ETGs with high spin parameter are classified as fast rotators (FRs) and those with low spin parameter are classified as slow rotators (SRs; Cappellari et al. 2007; Emsellem et al. 2007; see Cappellari 2016 for a review).

The ATLAS3D team examined the kinematic morphology–density relationship for the first time (Cappellari et al. 2011). They observed that there are fewer slow-rotating ETGs, relative to the total number of ETGs, in the lowest density local environments such that the fraction of slow-rotating ETGs is F_SR = N_SR/N_ETG ~ 0.13. However, the fraction of SRs more than doubles in the densest region of the Virgo cluster where F_SR ~ 0.28. Virgo is not a very massive cluster and provides only a single example of a dense environment. The kinematic morphology–density relationship has since been studied in seven additional clusters of differing cluster masses (Abell 1689, D’Eugenio et al. 2013; Coma, Houghton et al. 2013; Fornax, Scott et al. 2014 and Abell 85, 168 and 2399, Fogarty et al. 2014). These authors all find a total SR fraction, F_SR ~ 0.15 with no dependence on the global environment studied between the field/group sample of ATLAS3D and the most massive cluster studied to date, Abell 1689. However, these studies also find that F_SR generally rises with increasing local environmental density within these global environments. The Abell 168 and Abell 2399 clusters studied by Fogarty et al. (2014)
form an exception. In these two clusters the SR fraction peaks at intermediate densities and then falls. These two clusters are known to be undergoing mergers (Hallman & Markevitch 2004; Fogarty et al. 2014), and the SRs in these systems are associated with cluster substructure. Houghton et al. (2013) argued that the relationship observed between kinematic morphology and density is a result of mass segregation by dynamical friction because the total SR fraction is consistent across a range of global environments, while the SRs are segregated into the densest local environments. Cappellari (2016) came to the same conclusion owing to the presence of SRs near the centers of the Fornax, Virgo, and Coma clusters or subgroups within those clusters.

There is also a known relationship between galaxy mass and spin parameter such that more luminous galaxies with the highest mass show the lowest spin parameters (e.g., Emsellem et al. 2007; Cappellari 2013; Jimmy et al. 2013; Veale et al. 2016; Oliva-Altamirano et al. 2017). While low-mass dwarf galaxies have a strong relationship between spin parameter and environment (Guérrou et al. 2015; Toloba et al. 2015), it is not yet clear whether the driving force in the kinematic morphology–density relationship for the general elliptical galaxy population is environmental density or galaxy mass. Scott et al. (2014) examined this question for a sample of 30 ETGs in the Fornax cluster and found that even in mass-matched samples of slow and fast rotators, the SRs were found at preferentially higher projected environmental density than the fast rotators. They argued that dynamical friction alone could therefore not be responsible for the differing distributions of slow and fast rotators.

The Sydney-AAO Multi-object Integral-field spectrograph (SAMI; Croom et al. 2012) now makes it possible to obtain spatially resolved optical spectra for large numbers of galaxies covering a broad range in mass and environment. The SAMI Galaxy Survey (SGS) (Bryant et al. 2015) will observe ~3600 galaxies with stellar masses $7 < \log(M_*/M_\odot) < 12$ in a range of environments including eight galaxy clusters (Owers et al. 2017). We present here the kinematic morphology–density relationship for the ETGs observed in the eight SAMI clusters. This is the largest sample of cluster galaxies available to date and allows a robust analysis of the dependence of kinematic morphology on stellar mass as well as global and local environment.

In Section 2 we describe our observations and data reduction. Derived parameters are described in Section 3, and our kinematic classification is defined in Section 4. Results are presented in Section 5 and are discussed in Section 6, before conclusions are drawn in Section 7. Throughout this paper we assume a Hubble constant of $H_0 = 70$ km s$^{-1}$ Mpc$^{-1}$, $\Omega_M = 0.3$, $\Omega_\Lambda = 0.7$ and a Chabrier (2003) initial mass function.

## 2. Observations and Sample

The observations used in this analysis are selected from the SGS. The survey is ongoing and will observe ~3600 galaxies at redshifts $0.04 < z < 0.095$ with regular public data releases that will be accessible via the survey website, https://sami-survey.org (A. Green et al. 2017, in preparation). The survey target selection is described in Bryant et al. (2015). In brief, the main SAMI sample is selected from the Galaxy And Mass Assembly survey (GAMA; Driver et al. 2011; Hopkins et al. 2013). The GAMA sample covers broad ranges in stellar mass and environment, but does not include massive clusters. The SGS also targets eight additional clusters (Owers et al. 2017) to probe higher density environments.

Following previous analyses of the kinematic morphology–density relation, we focus here on the SAMI cluster sample; a separate paper will analyze the kinematic morphology–density relation in the main SAMI sample (J. van de Sande et al. 2017, in preparation). The selection of the eight clusters and their constituent galaxies is described in detail in Owers et al. (2017). In brief, cluster members were selected following a dedicated redshift program using the AAOmega spectrograph (Sharp et al. 2006) on the 3.9 m Anglo-Australian Telescope (AAT) fed by its multi-object fiber-feed: 2df. The redshift survey has high spectroscopic completeness, and 94% of potential cluster members have redshift measurements. Cluster membership was defined using a caustic analysis (Owers et al. 2017). The cluster members were used to estimate $R_{200}$, as well as the mass within $R_{200}$, $M_{200}$. The SAMI clusters range from $14.2 < \log(M_{200}/M_\odot) < 15.2$ in mass and $0.02 < z < 0.06$ in redshift (Table 1).

SAMI has a stepped stellar mass selection function, so that if the cluster redshift is less than $z = 0.045$, the stellar mass limit is $\log M_*/M_\odot = 9.5$ and clusters with redshifts above that have a stellar mass limit of $\log M_*/M_\odot = 10.0$. SAMI targets cluster members within $1 R_{200}$ and $\pm 3.5 V_{rad}/c_5$ (the clustercentric recession velocity with respect to the cluster velocity dispersion), and there are 848 cluster members meeting these stellar mass, radius, and recession velocity criteria. The stacked color–stellar mass distribution of the 848 cluster members is illustrated in Figure 1.

Previous analyses of the kinematic morphology–density relation focussed on the ETGs. We do not have visual morphologies for all of the galaxies in our sample. The majority of visually classified ETGs are red (although not all, e.g., Bassett et al. 2017), particularly in the cluster environment. We therefore make a color cut to select only the red galaxies (within $\pm 1\sigma$ of the fitted red sequence, equivalent to $+0.11$ mag) in the $g-i$ color–mass relationship for the stacked clusters (cf. Houghton et al. 2013). This selection is illustrated in Figure 1. The precise choice of color cut does not change our conclusions.

As lower mass galaxies were not observed by SAMI above $z = 0.045$, we only select galaxies with stellar masses $\log M_*/M_\odot > 10.0$ for the analysis presented here to maximize sample completeness over the whole redshift range. The mass and color cuts remove 181 and 108 galaxies, respectively, leaving a sample of 559 early-type cluster members with $\log M_*/M_\odot > 10.0$ (Table 1). Henceforth, we do not include cluster members with $\log M_*/M_\odot < 10.0$ in our analysis.

### 2.1. SAMI Observations and Data Reduction

The SAMI instrument (Croom et al. 2012) deploys 13 imaging fiber bundles, “hexabundles” (Bland-Hawthorn et al. 2011; Bryant et al. 2014), over a 1° field at the Prime Focus of the AAT. Each hexabundle consists of 61 circularly packed optical fibers. The core size of each fiber is 1.6 arcsec, giving each hexabundle a field of view of 15 arcsec diameter. All 819 fibers (793 object fibers and 26 sky fibers) feed into the AAOmega spectrograph. For SAMI observing, AAOmega is configured to a wavelength coverage of 370–570 nm with $15$ The radius at which the mean interior density is 200 times the critical density of the universe.
The Properties of the Eight Galaxy Clusters Observed by SAMI

| Cluster | R.A. J2000 | Decl. J2000 | zcl | M_200 | R_200 Mpc | σcl km s^{-1} | N_mem | N_ETG | N_obs | N_s | CompETG,ξ | N_SPS |
|---------|------------|-------------|-----|-------|------------|---------------|--------|-------|-------|-----|------------|-------|
| EDCC0442 | 6.38068 | −33.04657 | 0.0498 | 14.45 | 1.41 | 583 | 50 | 42 | 33 | 31 | 0.74 | 0 |
| Abell0085 | 10.460211 | −9.303184 | 0.0549 | 15.19 | 2.42 | 1002 | 167 | 138 | 65 | 58 | 0.42 | 12 |
| Abell0119 | 14.06715 | −1.25537 | 0.0442 | 14.92 | 2.02 | 840 | 253 | 138 | 64 | 55 | 0.40 | 0 |
| Abell0168 | 18.815777 | 0.213486 | 0.044 | 14.28 | 1.33 | 546 | 112 | 53 | 21 | 21 | 0.40 | 8 |
| Abell2399 | 329.389487 | −7.794236 | 0.0579 | 14.66 | 1.63 | 690 | 92 | 73 | 54 | 49 | 0.67 | 3 |
| Abell3880 | 336.97705 | −30.575371 | 0.0578 | 14.64 | 1.62 | 660 | 56 | 48 | 28 | 28 | 0.58 | 0 |
| APMCC0917 | 355.39788 | −29.236351 | 0.0509 | 14.26 | 1.19 | 492 | 29 | 23 | 18 | 15 | 0.65 | 0 |
| Abell4038 | 356.93781 | −28.140661 | 0.0452 | 14.36 | 1.46 | 597 | 89 | 44 | 37 | 35 | 0.80 | 0 |

Total: | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |

Note. R.A., decl., zcl, M_200, R_200, and σcl are all from Owers et al. (2017). N_mem is the number of members within 1 R_200 and ±3.5 V_σ/σcl (for clusters with zcl < 0.045, this includes galaxies with 9.5 < log(M_*/M_☉) < 10.0). N_ETG is the number of early-type members with log(M_*/M_☉) ≥ 10.0 (ETGs). N_obs is the number of observed ETGs. N_s is the number of observed ETGs for which the spin parameter could be measured. CompETG,ξ gives the fraction of ETGs with spin parameter measurements. N_SPS gives the number of SAMI pilot survey galaxies added to the analysis (Section 4.2.2).

Figure 1. g − i color as a function of stellar mass, M_*, for the 848 SAMI cluster members (within 1 R_200 and ±3.5 V_σ/σcl) and stellar mass log(M_*/M_☉) > 9.5 for zcl < 0.045 and log(M_*/M_☉) > 10.0 for zcl > 0.045: black open points and the galaxies selected for this analysis (red galaxies with log(M_*/M_☉) ≥ 10.0; red filled points). The solid line indicates a straight-line fit to the red sequence and the dashed line indicates the ±1σ scatter around this. The dotted line indicates the stellar mass limit log(M_*/M_☉) = 10.0.

R = 1812 in the blue arm, and 630–740 nm with R = 4263 in the red arm (van de Sande et al. 2017). A seven-point dither pattern achieves near-uniform spatial coverage (Sharp et al. 2015), with 1800 s exposure time for each frame, totalling 3.5 hr per field.

As described in Allen et al. (2015), 12 galaxies and a secondary standard star are observed in every field. The secondary standard star is used to probe the conditions as observed by the entire instrument. The flux zero-point is obtained from SDSS, while the shape of the flux correction is derived from primary standard stars observed in a single hexabundle during the same night for any given field of observation. The raw data from SAMI were reduced using the AAOmega data reduction pipeline, 2dfDRv5.62 (Croom et al. 2004; Sharp & Birchall 2010) followed by full alignment and flux calibration through the SAMI Data Reduction pipeline.

3. Derived Parameters

3.1. Photometry

The photometry for the SGS clusters is described in detail in Owers et al. (2017). We provide a brief overview here.

Four of the clusters (Abell 85, 119, 168, and 2399) lie within the Sloan Digital Sky Survey (SDSS; York et al. 2000). For these clusters the ugriz SDSS DR10 photometry has been remeasured using the IOTA software, which is used to measure aperture-matched photometry for the GAMA survey (Hill et al. 2011; Driver et al. 2016). Each frame was convolved to a common point-spread function (PSF) full-width at half-maximum (FWHM) = 2″ before using the SExtractor software (Bertin & Arnouts 1996) in dual-image mode to extract the aperture- and seeing-matched photometry. The r-band image was used for detection.

The four clusters not in the SDSS regions (EDCC 0442, APMCC0917, Abell 3880, and 4038) are covered by the Very large telescope Survey Telescope (VST) ATLAS (Shanks et al. 2015) survey. Raw VST/ATLAS data in the gri-bands were retrieved from the archive and reduced using the Astro-WISE optical image raw reduction pipeline (McFarland et al. 2013). The aperture- and PSF-matched photometry is measured as for the SDSS data, with one exception: the image quality of the VST/ATLAS imaging is higher than that of SDSS (Shanks et al. 2015), so each 1° × 1° gri-band tile was convolved to a...
common 1.5 FWHM. Owers et al. (2017) use the duplicate measurements of galaxies in Abell 85, which has full SDSS and partial VST/ATLAS coverage, to show that any systematic differences in the photometric and stellar mass measurements between the two surveys are less than 0.05 dex.

To calculate absolute magnitudes, we K-correct the apparent magnitudes to \( z = 0 \) using the IDL calc_kor.pro code (Chilingarian & Zolotukhin 2012).

### 3.2. Stellar Masses

Stellar masses are estimated using the empirical proxy between \( i \)-band absolute magnitude and \( g - i \) color given in Taylor et al. (2011) and also used by Bryant et al. (2015) for the main SGS. We use the aperture- and PSF-matched photometry described above, corrected for Galactic extinction using Schlegel et al. (1998) dust maps.

### 3.3. Photometric Fits

Galaxy effective radii and ellipticities are measured using the Multi Gaussian Expansion (Emsellem et al. 1994) technique implemented in the code from Cappellari (2002). These are presented in F. d’Eugenio et al. (2017, in preparation). The code measures circularized effective radii that we convert into the semimajor axis effective radius used throughout this paper as

\[
R_e = R_{e,circ}/(\sqrt{1 - \epsilon}),
\]

where the ellipticity, \( \epsilon \), is the luminosity-weighted ellipticity of the galaxy.

### 3.4. Galaxy Environment

The local environment of galaxies can be measured using a nearest-neighbor surface density to probe the underlying density field. The principle behind the nearest-neighbor measurement is that galaxies with closer neighbors are in denser environments (Muldrew et al. 2012). The nearest-neighbor measurement can be refined to an overdensity that parametrizes whether galaxies are in an environment that is more or less dense than the average in a given sample. We calculate the nearest-neighbor surface density, \( \Sigma_{N,\text{lim},\text{Mlim}} \), for all galaxies with reliable redshifts in the parent cluster redshift sample. The surface density is defined using the projected comoving distance to the \( N \)th nearest neighbor \( (\delta_3) \) with a velocity limit \( \pm V_{\text{lim}} \) km s\(^{-1} \):

\[
\Sigma_{N,\text{lim},\text{Mlim}} = N/\pi d_N.
\]

The neighbors are all within a volume-limited density-defining population that has absolute magnitudes \( M_r < M_{\text{lim}} = Q \Sigma \). \( Q \) defines the expected evolution of \( M_r \) as a function of redshift, \( z \), (\( Q = 1.03 \); Loveday et al. 2015). The large spatial extent of the SAMI Cluster Redshift Survey (extending to \( 2R_{200} \); Owers et al. 2017) means that when measuring nearest-neighbor surface densities within \( 1R_{200} \), no \( N \)th nearest neighbors are separated from the galaxy in question by more than the distance to the \( 2R_{200} \) cluster “edge.” Such large separations would rapidly increase the uncertainty of such measurements. We also measured the overdensity, \( \delta_{5000,18.3} = \Sigma_{5000}/\Sigma \), dividing the density by the mean density of the early-type members with log \( M_r/M_\odot > 10 \). In the Appendix we analyze the effect the choice of limits has on the nearest-neighbor surface density and overdensity. We find that the nearest-neighbor surface density is sensitive to the choice of limits applied, while the overdensity is not. We therefore choose an overdensity of \( \delta_{5000,18.3} \) in this analysis.

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16 http://kcor.sai.msu.ru

### 4. Kinematic Classification

#### 4.1. Stellar Kinematics

The stellar kinematic measurements made for the SAMI survey are described in detail in van de Sande et al. (2017). We summarize the salient points here.

The mean line-of-sight stellar velocity, \( V \), and velocity dispersion, \( \sigma \), are measured using the penalized pixel-fitting code (pPXF; Cappellari & Emsellem 2004). The SGS runs pPXF in two different modes. All results presented here consist of fits using a Gaussian line-of-sight velocity dispersion (LOSVD). To measure the stellar kinematics, the spectra from the blue and red arms of the spectrograph are combined. Before this can happen, the red spectra (FWHM\(_{\text{red}} = 1.61 \AA \)) are convolved to the instrumental resolution of the blue spectra (FWHM\(_{\text{blue}} = 2.65 \AA \)).

Optimal templates are constructed for 1–5 annular bins per galaxy (depending on the signal-to-noise ratio; \( S/N \)) by running pPXF over the combined spectra using the full MILES stellar library (Sánchez-Blázquez et al. 2006). After this, pPXF is run three times with only the optimal templates for each galaxy spaxel. The first run is used to estimate the real noise from the residuals. The second run uses the new noise spectrum for masking emission lines and bad pixels. The third run derives the LOSVD parameters. For each spaxel, pPXF is allowed to use the optimal templates from the annular bin in which the spaxel lives as well as from neighboring annuli. The uncertainties on the LOSVD are standard deviations after fitting pPXF to 150 simulated spectra. To construct the simulated spectra, the best-fit template is first subtracted from the spectrum. The residuals, together with the noise spectrum, are then randomly rearranged in wavelength space within eight wavelength sectors. The residuals are added to the best-fit template to construct the simulated spectra, which are then refitted with pPXF. We compared the measurement uncertainties we obtain from these 150 simulations with the pPXF uncertainty estimates and find that they agree well (van de Sande et al. 2017). However, looking at the 2D uncertainty maps, the simulated spectra provide uncertainty maps that are less stochastic and are more consistent with the \( S/N \) of the galaxy spectra, so we use these in our analysis.

We apply a quality cut, which ensures that we keep a large fraction of the low-velocity dispersion spaxels, while keeping a strict quality cut for the higher velocity dispersions (van de Sande et al. 2017): only spaxels with velocity dispersions \( \sigma > \text{FWHM}_{\text{inst}}/2 \sim 35 \text{ km s}^{-1} \) and a velocity and velocity dispersion uncertainties \( \Delta V < 30 \text{ km s}^{-1} \) and \( \Delta \sigma < (\sigma + 0.1) + 25 \text{ km s}^{-1} \) are retained in the final analysis. These quality cuts result in a sample whose median velocity dispersion uncertainty at \( S/N < 20 \text{ Å}^{-1} \) is 12.6% and 2.6% for \( S/N > 20 \text{ Å}^{-1} \) (Figure 2, van de Sande et al. 2017).

#### 4.2. Spin Parameter

Emsellem et al. (2007) defined the luminosity-weighted spin parameter (\( \lambda_R \)):

\[
\lambda_R = \frac{\sum_{i=0}^{i=N} F_i R_i V_i}{\sum_{i=0}^{i=N} F_i R_i V_i^2 + \sigma_i^2},
\]

in this analysis \( R_i \) is the semimajor radius of the ellipse in which spaxel \( i \) is located and \( F_i \) is the flux of the \( i \)th spaxel. \( \lambda_R \) is summed over all spaxels, \( N \), that meet the quality cut.
described above within the ellipse of the semimajor axis \( R \). The \( \lambda_R \) profiles are illustrated in Figure 2.

The spin parameter is summed within a fiducial radius, \( R_{\text{fid}} \), that can be either 0.5\( R_e \), 1\( R_e \), or 2\( R_e \) (following previous analyses). A particular fiducial radius is only used when \( R_{\text{fid}} > R_{\text{PSF}} \sim 1'.5 \), and the percentage of spaxels within that radius that meet the quality cut is >75%. Our first choice is to measure \( \lambda_{R_{\text{fid}}} \) within 1\( R_e \). However, if the effective radius is smaller than \( R_{\text{PSF}} \), we use \( R_{\text{fid}} = 2 \; R_e \), and if the galaxy has \( R_e > 15'' \), then we use \( R_{\text{fid}} = 0.5 \; R_e \). Two hundred and twenty-four galaxies have \( R_{\text{fid}} = 1 \; R_e \), 46 have \( R_{\text{fid}} = 2 \; R_e \), and 19 have \( R_{\text{fid}} = 0.5 \; R_e \). Three of the exceptionally large brightest cluster galaxies in these clusters have semimajor effective radii >15'' and so the fiducial radius for their \( \lambda \) measurements are \( \sim 0.3 \; R_e \). We cannot measure \( \lambda \) at all for 28 galaxies because nearby galaxies affect their observation (\( N = 22 \)) and the observations have a too low \( S/N \) (\( N = 6 \)). This leaves a sample of 292 galaxies for which we can measure \( \lambda_{R_{\text{fid}}} \).

The spin parameter as a function of ellipticity is shown in Figure 3. This plot has been constructed in the same way as the equivalent in van de Sande et al. (2017). The figures are not identical as here we focus on the cluster galaxies and do not study the higher-order stellar kinematics, so we have a lower \( S/N \) cut and include galaxies with \( \lambda \) measured within fiducial radii other than 1\( R_e \). For each galaxy in Figure 3, we show the velocity map to highlight the stellar velocities. To avoid overlap between the galaxy velocity maps, the data are first placed on a regular grid with a spacing of 0.02 in \( \lambda_{R_{\text{fid}}} \) and \( \epsilon \). We position each galaxy on its closest grid point, or its neighbor if its closest grid point is already filled by another galaxy. The size of the grid and velocity maps are chosen such that no galaxy is offset by more than one grid point from its original position. The stellar mass—Tully–Fisher (Dutton et al. 2011) relation is used for the velocity map color scale: for a galaxy with stellar mass \( \log M_*/M_\odot > 10 \), the scale of the velocity map ranges from \(-95 < V (\text{km s}^{-1}) < 95 \), whereas a galaxy with stellar mass \( \log M_*/M_\odot > 11 \) is assigned a velocity range from \(-169 < V (\text{km s}^{-1}) < 169 \). The kinematic position angle is used to align the major axis of all galaxies to 45°. The velocity maps are truncated where the \( S/N \) is too low (<3) and the errors do not meet the quality criteria. This truncation is different for every galaxy.

4.2.1. Choice of Fiducial Radius

Measuring \( \lambda_R \) within a fiducial radius of 0.5\( R_e \), or smaller could potentially affect our findings. It might introduce a bias as we are generally unable to reach a fiducial radius for the more massive galaxies of 1\( R_e \). We test the effect of this by taking the 224 galaxies with \( \lambda \) measured at 1\( R_e \) and find the value of \( \lambda \) at 0.5\( R_e \) for these galaxies. Because \( \lambda \) increases with increasing radius (Figure 2), measuring \( \lambda \) at radii smaller than 1\( R_e \) is likely to bias the measured \( \lambda \) lower, which artificially inflates the number of slow-rotating galaxies. Therefore, we focus on the 26 of the 224 test galaxies that have \( \lambda_{0.5\,R_e} \leq 0.1 \). The mean change in \( \lambda_R \) measured at 0.5\( R_e \) compared to 1\( R_e \), \( \lambda_{0.5\,R_e}/\lambda_{R_e} = 0.58 \pm 0.03 \), with no correlation with stellar mass. We test the effect of this offset by applying it to the 19 galaxies for which we can only measure \( \lambda_R \) within 0.5\( R_e \) or even smaller radii. Applying the offset does decrease the fraction of SRs, but it does not change any of the conclusions we draw in the remainder of this paper. We present the following results using the uncorrected \( \lambda_{R_{\text{fid}}} \) values.

4.2.2. Including SAMI Pilot Survey Observations

The SAMI Pilot Survey (Fogarty et al. 2014, 2015) includes stellar kinematics measured for 106 galaxies in 3 of the clusters (Abell 85, 168, and 2399) presented here. The Pilot Survey stellar kinematics were calculated in the same way as described here (Section 4.1), and their stellar masses and colors have been measured here. Of the 106 pilot survey galaxies, 78 are within 1\( R_{200} \) of their cluster center, and 69 of those 78 meet the color and stellar mass selection criteria we apply here. Of the 69 that meet our criteria, 46 have been reobserved to date as part of the SGS. We use slightly different quality criteria here, which mean that only 37/46 of these galaxies use the same fiducial radius to measure the spin parameter. In Figure 4 we compare the Pilot survey spin parameters with those measured here and find a mean difference \( \lambda_{R_{\text{fid}}} \) (SGS – Pilot) = 0.025 ± 0.007 for these 37 galaxies. There are some outliers in this distribution. Our methods for determining the stellar kinematics, sizes, and ellipticities have been improved, and when we compare the kinematic maps for the outliers, we find holes that are due to the exclusion of low \( S/N \) data in the maps of Fogarty et al. (2015) that are not present in our data. We note that the FR and SR classifications do not change between the two surveys. We therefore include the 23 Pilot Survey galaxies that meet our selection criteria but have not yet been reobserved as part of the SGS in our analysis from this point. The inclusion of these galaxies does not affect the conclusions we draw.

Adding the 23 SAMI Pilot Survey galaxies to the 292 SGS galaxies gives us a final sample of 315 galaxies.

4.3. Sample Completeness

We are interested in fractional quantities, so it is important to understand the observed completeness of our sample.

The completeness of galaxies with spin parameter measurements as a function of stellar mass is shown in Figure 5. The observed completeness rises as a function of increasing stellar mass as a result of early targetting decisions (Owers et al. 2017). We analyze the effect of this further in Section 5.1.
The completeness of galaxies with spin parameter measurements as a function of overdensity is shown in Figure 6. We find that the observed completeness is flat as a function of environmental overdensity.

4.4. Slow/Fast Rotator Separation

Using a quantitative analysis of stellar velocity maps, the ATLAS3D team (Krajnović et al. 2008, 2011) found that their sample of 260 ETGs broke into broad groupings of fast- and slow-rotating galaxies that could be separated using the following definitions from Emsellem et al. (2011; Equations (2) and (3)) and Fogarty et al. (2014; Equation (4)):

$$\lambda_R < 0.08 + \epsilon_e / 4 \text{ with } \epsilon_e < 0.4,$$

as well as the classification from Emsellem et al. (2007),

$$\lambda_R < 0.1,$$

shown as the dashed line in Figure 3. Similar to the ATLAS3D team, we see broad groups of fast- and slow-rotating galaxies in
Figure 3. When we use the Emsellem et al. (2011) definition, there are 30 SRs in our sample, in comparison to 42 using the Cappellari (2016) definition and 38 using the Emsellem et al. (2007) definition. Figure 3 shows that the Emsellem et al. (2011) SR definition appears not to select some slow-rotating galaxies, while both the Emsellem et al. (2007) and Cappellari (2016) definitions capture all the slow-rotating galaxies. We therefore use the Cappellari (2016) definition in the remainder of this paper, but we note that the choice of slow/fast rotator separation does not affect the conclusions we draw.

Figure 5. Completeness of galaxies with spin parameter measurements as a function of stellar mass. The lower panel shows the stellar mass distribution of early-type members with log(M*/M⊙) > 10 (Member ETGs) and those with spin parameter measurements (Observed), while the upper panel shows the fraction of them (Completeness) as a function of stellar mass. The error bars show the 1σ binomial confidence limits on these measurements. The observed completeness rises as a function of increasing stellar mass.

Figure 6. Completeness of galaxies with spin parameter measurements as a function of overdensity. The lower panel shows the overdensity distribution of early-type members with log(M*/M⊙) > 10 (Member ETGs) and those with spin parameter measurements (Observed), while the upper panel shows the fraction of them (Completeness) as a function of overdensity. The error bars show the 1σ binomial confidence limits on these measurements. The observed completeness is flat as a function of overdensity.

Figure 7. Total fraction of slow rotators, FSR, as a function of host cluster mass, M200. The dashed line shows the total fraction of slow rotators across our whole sample. The uncertainties are the fractional uncertainties. We find no relationship of total FSR with host cluster mass over this mass range.

5. Results

5.1. Fraction of SRs

We investigate here the fraction of slow-rotating galaxies as a function of the number of ETGs, FSR. Throughout this section, we plot FSR in bins of equal numbers of galaxies with fractional uncertainties calculated using binomial confidence intervals that have been shown to be accurate for small to intermediate sample sizes (Cameron 2011).

The total fraction of SRs in our sample is FSR = 0.14 ± 0.02. We examine the total fraction of SRs per cluster as a function of host cluster mass in Figure 7 and find no significant relationship over the mass range examined here.

In the left-hand panel of Figure 8 we examine the fraction of slow-rotating galaxies, FSR, as a function of local environmental overdensity. We calculate the significance of our results by comparing the fraction of SRs in the four lowest overdensity bins (which are statistically equal) to the fraction in the highest overdensity bin, taking into account the uncertainties in this measurement: significance = (FSR(high) − FSR(low))/\sqrt{σ2_{FSR,high} + σ2_{FSR,low}}. We find an increasing fraction of SRs with increasing overdensity with a significance of 3.4σ.

We note that there is a higher SAMI observing completeness at higher stellar masses as a result of the early targetting decisions (Figure 5). This bias could affect the FSR–δ relationship. We test this by Monte Carlo resampling the observed data. We determine the lowest completeness in stellar mass (40% at log(M*/M⊙) ~ 10), and random resampling galaxies with stellar masses above log(M*/M⊙) ~ 10.5 down to that lowest observed completeness. We recalculate the SR fraction as a function of overdensity for each of 100 random resamplings. The mean FSR(corrected) is shown by the dashed line in the left-hand panel of Figure 8 and is indistinguishable from the observed relationship within the uncertainties.

In Figure 9 we examine FSR as a function of the environment overdensity for each of the eight clusters in the SAMI survey. These fractions are noisier than the stacked relationship shown in Figure 8 and do not all rise with increasing overdensity. Like
Fogarty et al. (2014), we find that the SR fraction within Abell 2399 peaks at intermediate overdensities and then drops, although we note that the uncertainties are large, so that the radial change is not statistically significant.

Galaxy density increases at the centers of galaxy clusters. We test how the fraction of slow-rotating galaxies changes as a function of stacked clustercentric radius ($R_{cl}/R_{200}$) in the central panel of Figure 8. We see that $F_{SR}$ increases as a function of decreasing clustercentric radius with a significance of 2.9σ. Projection effects are significant in clusters, and these effects act to dilute correlations with clustercentric radius. There is also an increase in $F_{SR}$ at $\sim 0.6 R_{200}$. Although this bump is not statistically significant, we also examine the spatial distribution of the SRs in each cluster in Figure 10. While this plot suffers from the effects of sample incompleteness as well as uncertainties in our spin parameter measurements, these effects are mitigated by the higher completeness for higher stellar mass galaxies (which are more likely to be slow-rotating galaxies from the right-hand panel of Figure 8) and by showing the galaxies’ $\lambda_{R_{cl}}$ values rather than simply whether they are FRS or SRs. The slow-rotating (redder) galaxies are generally located in the cluster centers ($R_{cl} < 0.3 R_{200}$; as indicated by the central panel of Figure 8), and those few that are located outside the cluster centers are generally associated with substructure in the galaxy distribution (Abell 85, 119, and 2399). Abell 168 has a massive SR at $R_{cl} > 0.3 R_{200}$ but does not show deviations in the galaxy distribution in Figure 10, however, it is a well-known merging cluster (e.g., Ulmer et al. 1992) with substructure visible in the X-rays at the position of the SR (Fogarty et al. 2014). The EDCC 0442 cluster is an outlier to this picture, with three slow-rotating galaxies located away from the cluster center ($R_{cl} \sim 0.5 R_{200}$). While this cluster does not have substructure visible in the smoothed galaxy distribution, in the X-ray it is a warm-core cluster and is likely to have relaxed only recently (Burns et al. 2008), which could be responsible for the broader distribution of slow-rotating galaxies.

After examining the relationship between SR fraction and different measures of environment, we now examine the relationship between SR fraction and stellar mass. The right-hand panel of Figure 8 shows that the fraction of slow-rotating galaxies increases with increasing stellar mass with a significance of 5.0σ.

We test again whether the higher SAMI observing completeness at higher stellar masses affects the $F_{SR} – M_*$ relationship. The mean $F_{SR, corrected}$ is shown by the dashed line in the right-hand panel of Figure 8 and is indistinguishable from the observed relationship.

The relationship of $F_{SR}$ with mass is a more significant relationship than that seen with overdensity, suggesting that...
higher stellar masses could be the dominant cause of the increase in slow-rotating galaxies observed with increasing environmental density and decreasing clustercentric radius. We explore this idea further in the next section.

5.2. Distribution of Spin Parameter

To explore the relationship between spin parameter, environmental density, and stellar mass further, we now examine the distribution of these parameters, rather than simply separating the sample into slow and fast rotators. In order to do this, we need to take into account the fact that $\lambda_{\text{fid}}$ is a projected quantity.

We examine the distribution of spin parameter, $\lambda_{\text{fid}}$, applying an approximate correction for the effects of projection by dividing by ellipticity, $\sqrt{\epsilon}$ (Emsellem et al. 2011). The upper panel of Figure 11 shows $\lambda_{\text{fid}}/\sqrt{\epsilon}$ as a function of stellar mass. The points are colored by their environmental densities. We also show the mean $\lambda_{\text{fid}}/\sqrt{\epsilon}$ as a function of stellar mass for two overdensity bins (the lowest and highest quartiles; mean overdensities $\log \delta_{\text{low}} = -0.80, \log \delta_{\text{high}} = 0.35$). Both density bins show a relationship of decreasing $\lambda_{\text{fid}}/\sqrt{\epsilon}$ with increasing stellar mass, but no significant difference in that relationship as overdensity increases. We note that the most massive galaxies, with $\log M_*/M_\odot > 11.3$, have the lowest $\lambda_{\text{fid}}/\sqrt{\epsilon}$ and are generally in the most overdense regions ($\delta_{5500,-18.3} > 0.5$).

The lower panel of Figure 11 shows $\lambda_{\text{fid}}/\sqrt{\epsilon}$ as a function of overdensity, $\delta_{5500,-18.3}$ with points colored by stellar mass. We also show the mean $\lambda_{\text{fid}}/\sqrt{\epsilon}$ as a function of overdensity for two stellar mass bins (the lowest and highest quartiles; mean mass $\log M_\odot/10^{14} M_\odot = 10.14, \log M_\odot/10^{11} M_\odot = 11.08$). Neither mass bin shows a strong relationship of $\lambda_{\text{fid}}/\sqrt{\epsilon}$ with overdensity. However, there is a systematic offset to lower $\lambda_{\text{fid}}/\sqrt{\epsilon}$ for the higher stellar mass sample. We also note that the most overdense regions ($\delta_{5500,-18.3} > 0.5$) are dominated by the group of massive, $\log M_*/M_\odot > 11.3$, low $\lambda_{\text{fid}}/\sqrt{\epsilon}$ galaxies that are also visible in the upper panel.

A partial correlation analysis shows that the strongest relationship is between stellar mass and $\lambda_{\text{fid}}/\sqrt{\epsilon}$ ($R = -0.30, p = 4 \times 10^{-3}$), with a correlation between stellar mass and environment ($R = 0.18, p = 0.001$), while the relationship between $\lambda_{\text{fid}}/\sqrt{\epsilon}$ and surface density is not significant ($R = -0.11, p = 0.04$). We conclude that the kinematic morphology–density relationship is due to the changing distribution of stellar mass with environment.

6. Discussion

We find a total SR fraction $F_{\text{SR}} = 0.14 \pm 0.02$ and that this fraction does not depend significantly on host cluster mass. The lack of dependence of SR fraction on global environment is consistent with previous measurements from the ATLAS$^{3D}$ field/group sample (Cappellari et al. 2011) to the massive dense cluster Abell 1689 (D’Eugenio et al. 2013), who find a total $F_{\text{SR}} \sim 0.15$.

We find that $F_{\text{SR}}$ does depend on local environment, measured by overdensity, such that the fraction of slow-rotating galaxies increases as the local environment overdensity increases. The dependence of SR fraction on local environment is also consistent with previous analyses (Cappellari et al. 2011; D’Eugenio et al. 2013; Houghton et al. 2013; Fogarty et al. 2014; Scott et al. 2014).

We also find a strong relationship of $F_{\text{SR}}$ with galaxy stellar mass. This relationship has been observed before (e.g., Emsellem et al. 2007; Cappellari 2013; Jimmy et al. 2013; Veale et al. 2016; Oliva-Albamirano et al. 2017) and is not surprising given the analytic relationship between spin parameter $\lambda$, angular momentum $J$ and total mass $M$: $\lambda = (J/E^{1/2})/(GM^{5/2})$ (where $E$ is the total energy of the system and $G$ is the gravitational constant; e.g., Fall &
Efstathiou 1980; Romanowsky & Fall 2012). A strong relationship between specific angular momentum, $j_e$, and stellar mass has also been observed in the main SGS by Cortese et al. (2016).

Simulations are also observing relationships between specific angular momentum, spin parameter, and mass. The analysis of specific angular momentum in the EAGLE simulation (Schaye et al. 2015) by Lagos et al. (2017) also finds that it depends on stellar mass and concludes that galaxies with low $j_e$ at $z \sim 0$ are a product of two pathways: galaxy mergers, and early star formation quenching. Similarly, the analysis of the Illustris simulation (Genel et al. 2014) galaxies

by Penoyre et al. (2017) finds that the slow-rotating elliptical galaxies are more massive than the fast-rotating galaxies. They also find that the slow-rotating galaxies have evolved from fast rotators since $z = 1$ as a result of mergers that caused them to spin down. However, neither of these simulations include the massive cluster environments studied here. Choi & Yi (2017) examine the evolution of the spin parameter of galaxies in cluster environments in a cosmological hydrodynamic simulation. They find that the spin evolution is mass dependent, with more massive galaxies ($\log(M_*/M_\odot) > 10.5$) experiencing more spin-down, mainly as a result of major and minor mergers. In contrast, while the spin parameter of the lowest mass galaxies ($\log(M_*/M_\odot) < 10.5$) also falls with time, this decrease is more driven by environment than by mergers. Because this mass range is at the very lowest end of our sample, we cannot rule this prediction out. We also note that observations of low-mass dwarf galaxies see a strong relationship between spin parameter and environment (Toloba et al. 2015).

In this analysis, we have a large enough sample to separate the effects of local environment and stellar mass on spin parameter. When the distribution of $\lambda_{R fid}$ with stellar mass is analyzed together with the galaxies’ local environment, we find no significant residual dependence on environment. The lack of dependence of spin parameter on environment, once the effects of mass are removed, is in contrast to the analysis of the Fornax and Virgo clusters by Scott et al. (2014). They found that even in mass-matched samples of slow and fast rotators, the SRs were found at higher projected environmental densities than the FRs. However, we note that that study was of $N \sim 70$ galaxies in two low-mass clusters, and our analysis of the kinematic morphology—density relationship shows that the picture in individual clusters may differ from the distribution as a whole. Our observations are consistent with more recent analyses of the classical morphology—density relationship, which show that at fixed stellar mass, morphology is only weakly dependent on environment (Bamford et al. 2009).

Figure 10 shows the spatial distribution of the slower and faster rotators in each of the clusters. The slow-rotating galaxies that are not within the cluster cores are generally observed to reside within substructure in those clusters. These substructures are likely to be made up of groups that have fallen into these clusters (Yi et al. 2013). We postulate that this is evidence that the kinematic morphology—density relationship is a result of mass segregation due to dynamical friction. This evidence would suggest that slow-rotating ETGs form in a group environment that either accretes other groups over time to become a cluster, or is itself consumed to become substructure in a larger system. This hypothesis is consistent with the conclusions from Cappellari (2016). It will be possible to test this hypothesis with the main SGS sample (J. van de Sande et al. 2017, in preparation), which is based on the GAMA survey of galaxies and includes a robustly selected sample of galaxy groups (Robotham et al. 2011). Examining the kinematic morphology—density relationship in the GAMA group sample will verify whether slow-rotating galaxies form in the group environment.

7. Conclusions

We have presented here the kinematic morphology—density relationship for a sample of 315 ETGs in eight galaxy clusters...
from the SGS. The eight clusters span a halo mass range of $14.2 < \log(M_{200}/M_\odot) < 15.2$. Cluster members were observed within $1 R_{200}$ and $\pm 3.5 V_{gal}/c$, which covers local galaxy environments (measured by overdensity) between $-1.5 < \log(\delta) < 1.0$. The stellar masses observed range from $10.0 < \log(M_*/M_\odot) < 11.7$. We classify these galaxies as fast or SRs depending on their spin parameter, $\lambda_{rot}$, measured from spatially resolved stellar kinematics. We analyze the fraction of SRs, $F_{SR}$, as a function of local galaxy environment and stellar mass. We also examine the distribution of $\lambda_{rot}$ as a function of both environment and stellar mass. We draw the following conclusions that are not qualitatively dependent on fiducial radius or choice of fast/slow galaxy classification:

1. We find a total SR fraction of $F_{SR} = 0.14 \pm 0.02$.
2. The SR fraction per cluster shows no dependence on host cluster mass in the range studied.
3. We find $F_{SR}$ to depend on local cluster environment such that it increases with increasing environmental overdensity, from $F_{SR} = 0.14^{+0.05}_{-0.03} \log(\delta) - 0.9$ to $F_{SR} = 0.20^{0.06}_{-0.05} \log(\delta) - 0.4$, a significance of $3.4\sigma$.
4. $F_{SR}$ depends more strongly on stellar mass than on local cluster environment. The fraction of SRs increases with increasing stellar mass from $F_{SR} = 0.13^{0.06}_{-0.03} \log(M_*/M_\odot) = 10.1$ to $F_{SR} = 0.41^{0.07}_{-0.06} \log(M_*/M_\odot) = 11.2$, which is a significance of $5.0\sigma$.
5. Once any dependence on stellar mass is removed from the distribution of spin parameter, $\lambda_{rot}/c$, no significant relationship with local cluster environment remains.

We conclude that the cluster kinematic morphology–density relationship is a result of mass segregation. We will test this hypothesis further with the broader SGS sample (J. van de Sande et al. 2017, in preparation).

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GAMA is a joint European-Australasian project based around a spectroscopic campaign using the Anglo-Australian Telescope. The GAMA input catalog is based on data taken from the Sloan Digital Sky Survey and the UKIRT Infrared Deep Sky Survey. Complementary imaging of the GAMA regions is being obtained by a number of independent survey programs including GALEX MIS, VST KiDS, VISTA VIKING, WISE, Herschel-ATLAS, GMRT and ASKAP providing UV to radio coverage. GAMA is funded by the STFC (UK), the ARC (Australia), the AAO, and the participating institutions. The GAMA website is: http://www.gamasurvey.org/.

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**Appendix**

**Environmental Densities**

We investigated $N^\text{th}$ nearest-neighbor surface density measurements, testing the effect the choice of limits has on the environmental density we measure. We measured a suite of environmental densities, varying the $N^\text{th}$ nearest neighbor ($N = 3, 5, 10$), velocity ($V_{lim} = 300, 500, 1000 \text{ km s}^{-1}$), and absolute magnitude ($M_{lim} = -18.3, -19 \text{ mag}$) limits. We also measured the overdensity, $\delta_N, V_{lim}, M_{lim} = \Sigma/\Sigma_\odot$, dividing the density by the mean density of the ETGs with $M_* > 10^9 M_\odot$ within $1 R_{200}$ of their cluster centroid.

The choice of limits affects the specific value of the environment density calculated (Figure 12). This emphasizes the need for caution when directly comparing the densities measured from non-homogeneous data sets that have different velocity or magnitude limits or background corrections. However, we find the overdensities, the density divided by the mean density, to be independent of the applied limits (Figure 13). We therefore use the overdensity $\delta_{5500}^{18.3}$ in this work.
density, they no longer depend on the limits applied. Once the nearest-neighbor measurements are corrected to an over-

\[ V_{\text{lim}} = \text{range of velocities} \]

\[ M_{\text{lim}} = \text{range of magnitudes} \]

\[ N_{\text{th neigbor}} = \text{number of neighbors} \]

\[ F_{\text{SR}}(N_{\text{th neigbor}}) \]

\[ \log(\sigma_{V_{\text{lim}}, M_{\text{lim}}}/\text{Mpc}^{-1}) \]

\[ V_{\text{lim}} = \text{range of velocities} \]

\[ M_{\text{lim}} = \text{range of magnitudes} \]

\[ N_{\text{th neigbor}} = \text{number of neighbors} \]

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\[ V_{\text{lim}} = \text{range of velocities} \]

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\[ N_{\text{th neigbor}} = \text{number of neighbors} \]

\[ F_{\text{SR}}(N_{\text{th neigbor}}) \]

\[ \log(\sigma_{V_{\text{lim}}, M_{\text{lim}}}/\text{Mpc}^{-1}) \]