GALAXY INFALL BY INTERACTING WITH ITS ENVIRONMENT: A COMPREHENSIVE STUDY OF 340 GALAXY CLUSTERS

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ABSTRACT

To study systematically the evolution of the angular extents of the galaxy, intracluster medium (ICM), and dark matter components in galaxy clusters, we compiled the optical and X-ray properties of a sample of 340 clusters with redshifts <0.5, based on all the available data from the Sloan Digital Sky Survey and Chandra/XMM-Newton. For each cluster, the member galaxies were determined primarily with photometric redshift measurements. The radial ICM mass distribution, as well as the total gravitational mass distribution, was derived from a spatially resolved spectral analysis of the X-ray data. When normalizing the radial profile of galaxy number to that of the ICM mass, the relative curve was found to depend significantly on the cluster redshift; it drops more steeply toward the outside in lower-redshift subsamples. The same evolution is found in the galaxy-to-total mass profile, while the ICM-to-total mass profile varies in an opposite way. The behavior of the galaxy-to-ICM distribution does not depend on the cluster mass, suggesting that the detected redshift dependence is not due to mass-related effects, such as sample selection bias. Also, it cannot be ascribed to various redshift-dependent systematic errors. We interpret that the galaxies, the ICM, and the dark matter components had similar angular distributions when a cluster was formed, while the galaxies traveling in the interior of the cluster have continuously fallen toward the center relative to the other components, and the ICM has slightly expanded relative to the dark matter although it suffers strong radiative loss. This cosmological galaxy infall, accompanied by an ICM expansion, can be explained by considering that the galaxies interact strongly with the ICM while they are moving through it. The interaction is considered to create a large energy flow of 10^{44}−45 erg s^{-1} per cluster from the member galaxies to their environment, which is expected to continue over cosmological timescales.

Key words: galaxies: clusters: general – galaxies: evolution – intergalactic medium – X-rays: galaxies: clusters

1. INTRODUCTION

In the standard cosmological model, the formation of large-scale structure is dominated by gravitational dynamics, while the gas physics plays a minor role. The gravitational collapse of cosmic matter over several megaparsecs creates galaxy clusters, the largest virialized system in the universe. On small-scale domains (e.g., galaxies), instead, nongravitational processes, related to the dynamics and evolution of baryons, become more and more important. The physical states of galaxies and the intracluster medium (ICM) are shaped by complex processes such as radiative cooling, feedback from supernovae and active galactic nuclei, star formation, and interactions between galaxies and the ICM. Because galaxy clusters stand at the transition between the two scale domains, they are often studied for both cosmological and astrophysical aims. It is hence crucial to quantify how the astrophysical processes have affected the cosmological properties of the cluster baryons.

So far, the cluster member galaxies and the ICM have been assumed in many cases to be subject to different sets of astrophysical processes and evolve separately over cosmological timescales. However, based on X-ray observations of the ICM with ASCA, Makishima et al. (2001, hereafter M01) proposed a novel picture: physical interactions occur universally between member galaxies and the ICM, which transfer free energy from galaxies to the ICM and drag the galaxies so that they ultimately fall to the cluster center over the Hubble time. Indeed, possible remnants of the interaction have been observed around many galaxies in H I (e.g., Oosterloo & van Gorkom 2005, Hα (e.g., Yoshida et al. 2008), and X-ray bands (e.g., Sun et al. 2006; Gu et al. 2013b).

This simple idea can potentially shed light on several unsolved issues. First, it immediately explains why the ICM in nearby clusters has a much more extended angular distribution than the member galaxies. The implied member galaxy infall may also be intimately linked to the formation of central brightest cluster galaxies (or the cD galaxies), as well as that of diffuse intracluster light. At the same time, it can potentially answer a long-standing question: why do galaxies evolve differently inside and outside clusters? The fraction of blue
galaxies, which are considered to be gas-rich and star-forming objects, increases with redshift up to $z \sim 0.4$ in the cores of rich clusters (known as the Butcher–Oemler effect; Butcher & Oemler 1984); such a feature is not found in low-density environments. This kind of “environmental effect,” the ultimate driving force of which remained unknown, may be understood as a consequence of the proposed interactions of the moving galaxies with the ICM. As recently reported in Stroe et al. (2015), star formation in member galaxies appears to be excited by cluster-scale merging shocks in the ICM. The massive star formation would rapidly consume the molecular content in the galaxies and eventually transforms them to the red population. The M01 scenario can also explain, in a natural way, how the large amount of metals, which must have originally been synthesized in galaxies, are currently distributed to much larger radii than the galaxies (e.g., Kawaharada et al. 2009; Matsushita et al. 2013), and the ICM in nearby clusters is metal enriched uniformly up to the periphery (e.g., Werner et al. 2013); the galaxies used to be distributed to larger radii and enriched that portion of the ICM while they gradually fell to the center.

Another important consequence expected from the M01 scenario is the energy transfer toward the ICM. The high ICM density in the cluster center would result in a runaway cooling, which leads to the formation of enormous cooling flow of gas and massive star formation in the cD galaxy. However, broadband X-ray spectroscopy, starting with ASCA, found that the effect of cooling is much weaker than previously predicted (e.g., M01; Peterson et al. 2001), suggesting that some heating mechanisms are in operation. In the galaxy infall scenario, the energy flow from galaxies provides an important and inherent heating source for the ICM, operating essentially in all clusters. To verify the M01 scenario, the key is to compare the spatial extents of member galaxies and the ICM in clusters at different ages. As reported in Gu et al. (2013a, hereafter Paper I), we studied the expected galaxy infall phenomenon using a statistical sample of 34 massive clusters with a redshift range of 0.1–0.9. We detected, for the first time, a significant evolution spanning a time interval of $\sim$6 Gyr in the relative spatial distributions of the cluster galaxies and the ICM; while the galaxy component was as spatially extended as the ICM at $z > 0.5$, toward the lower redshifts, it has indeed become more centrally concentrated relative to the ICM sphere. Since the concentration was found to be rather independent of the galaxy mass, it cannot be explained by pure gravitational drag. This result provides an important support to the galaxy–ICM interaction scenario proposed by M01.

Although the results in Paper I are quite firm, the implied view is so novel that further efforts are still needed to make them more convincing and detailed. To minimize the statistical uncertainty (currently $\sim$20%; Figure 8 of Paper I), it is necessary to increase significantly the sample size. To strengthen the detection of the evolutionary effects, our new study should also include the redshift-dependent angular distributions of the baryon versus dark matter components, which were only briefly studied in Paper I. In addition, we would like to examine in detail how the relative distributions of member galaxy, ICM, and dark matter components evolve as a function of such parameters as the cluster mass, dynamical state, ICM density, galaxy mass, and galaxy color. Such a comprehensive study will enable us to establish conclusively the view of galaxy infall proposed in M01.

We present the new study in the present paper, which has the following layout. Section 2 gives a brief description of the sample selection and data reduction procedure. The data analysis and results are described in Section 3. We discuss the physical implications of our results in Section 4, and we summarize our work in Section 5. Throughout the paper, we assume a Hubble constant of $H_0 = 70 \, h_\text{71} \, \text{km} \, \text{s}^{-1} \, \text{Mpc}^{-1}$, a flat universe with the cosmological parameters of $\Omega_M = 0.27$ and $\Omega_{\Lambda} = 0.73$, and quote errors by the 68% confidence level unless stated otherwise. The optical magnitudes used in this paper are all given in the AB system.

## 2. OBSERVATION AND DATA PREPARATION

### 2.1. Sample Selection

To correlate the ICM and galaxy evolution, it is important to construct a large statistical sample of X-ray-bright clusters, which is based on high-quality X-ray and optical data and has a large coverage in the cluster mass and redshift. Usually the sample size is limited by the availability of X-ray data, which is often much less complete than the optical ones. Therefore, we started with selecting all clusters available in the Chandra and XMM-Newton archives until 2015, 509 and 442, respectively. The two archives have an overlap of 262, so that the total number available in X-ray is 689. For the 262 clusters in both X-ray archives, we selected the data with better signal-to-background ratio (see Section 3.5.2 for definition). Then, as shown in Table 1, about two-thirds of the X-ray clusters, 468, were found to be covered by the Sloan Digital Sky Survey (SDSS) up to the data release 12. The list, consisting of 468 clusters, forms our “Preliminary Sample.”

We have applied two basic filters to our Preliminary Sample. First, the nearby Virgo, Coma, and Perseus Clusters were discarded for their oversized angular extents. Second, it was further screened to remove those with poor X-ray data quality (i.e., net count $<5000$). The remaining number is 381, to be called the “Intermediate Sample.” They were further categorized into three subsamples by their redshift, i.e., a low-redshift sample with 138 objects ($z \sim 0.0–0.08$; hereafter subsample L), an intermediate-redshift subsample with 130 objects ($z \sim 0.08–0.22$; hereafter subsample M), and a high-redshift subsample consisting of 113 clusters ($z \sim 0.22–0.45$; hereafter subsample H). The superposed optical and X-ray images of six example clusters, two for each subsample, are shown in Figure 1.

### 2.2. Completeness Check

As described above, the sample selection is primarily based on observational limitations, e.g., archival state and X-ray data quality, rather than on some objective criteria (e.g., flux or redshift limited). Therefore, our Intermediate Sample may be

| Data Source | Raw | Selected | SDSS | X-ray Quality* | $M_{500}$ Filter |
|-------------|-----|----------|------|----------------|------------------|
| Chandra      | 509 | 332      | 223  | 161            | 141              |
| XMM-Newton   | 442 | 357      | 245  | 220            | 199              |
| Total        | 951 | 689      | 468  | 381            | 340              |

Note. $^*$ The neighboring Virgo, Perseus, and Coma Clusters are also removed.
subject to some selection biases. To examine this issue, we investigate in detail the sample completeness. In essence, minor incompleteness is acceptable for our purpose, since we already found in Paper I that the major aim of our study, the systematic correlation between X-ray and optical structures of the clusters, does not depend strongly on the cluster richness.

To test X-ray completeness of the Intermediate Sample, we compared it with existing flux-limited catalogs of the NORAS

Figure 1. RGB-colored composite images (red for SDSS-i, green for SDSS-g, and blue for Chandra/XMM-Newton 0.3–0.8 keV) of six example clusters in our sample. Detected member galaxies (Section 2.4) are marked by small green boxes. The X-ray-determined $R_{500}$ of each cluster is shown with a white circle.
and REFLEX surveys, which are known to be fairly complete (∼80% and >90%, respectively; Böhringer et al. 2000, 2004). In Figure 2(a), the fraction of NORAS/REFLEX clusters covered by our sample is plotted as a function of the 0.1–2.4 keV X-ray flux. Our sample recovers ∼90%/73% of the sky density of clusters compared to the NORAS/REFLEX survey at $5 \times 10^{12}$ erg s$^{-1}$. Even at $1 \times 10^{12}$ erg s$^{-1}$, the relative completeness reaches 60%. This means that most of the X-ray-detected clusters in the SDSS sky coverage are included in our sample.

To characterize the sample in another way, we calculated the X-ray luminosity functions of sample clusters, normalized to the effective survey volume, i.e., a sphere centered on Earth and extending up to the most distant object. Here the X-ray fluxes were taken from the NORAS/REFLEX catalogs. As shown in Figure 2(b), the obtained luminosity function is consistent with that of the REFLEX II survey (Böhringer et al. 2014) for bright objects, i.e., $L_X > 10^{44}$ erg s$^{-1}$, while at the fainter end, our Intermediate Sample is less complete, by ∼30%–40%, than the REFLEX II catalog.

A more vital test for this study is to investigate the sample coverage in the cluster mass versus redshift space. In Figure 2(c), we plot $M_{500}$ vs. redshift of the sample. Subsamples L, M, and H are shown with blue, green, and red points, respectively. To quantify the effect, we projected the $M_{500}$ vs. redshift plot onto the $M_{500}$ axis to derive a mass function for each subsample. As shown in Figure 2(d), the three subsamples define similar mass functions at $>10^{14} M_\odot$, while the subsample L contains twice as many objects with $M_{500} \lesssim 10^{14} M_\odot$, as the other two, thus reconfirming Figure 2(c). To avoid this selection bias, we limit the subsequent study to the clusters with $M_{500}$ in the range of $(0.1–2) \times 10^{15} M_\odot$. This defines our “Final Sample,” containing 340 clusters, which are subdivided into 119, 117, and 104 objects in the subsamples of L, M, and H, respectively.

2.3. X-Ray Data

Our Final Sample includes 141 and 199 clusters with X-ray data from Chandra and XMM-Newton archives, respectively.

2.3.1. Chandra

The Chandra data obtained with its advanced CCD imaging spectrometer (ACIS) were screened with the CIAO v4.6 software and CALDB v4.6.3. Following the standard procedure described in, e.g., Gu et al. (2009), we discarded bad pixels and columns, as well as events with ASCA grades 1, 5, and 7, and
corrected the event files for the gain, charge transfer inefficiency, astrometry, and cosmic-ray afterglow. By examining the light curves extracted in 0.3–10.0 keV from source-free regions, we identified and excluded time intervals contaminated by a flare-like particle background with a count rate more than 20% higher than the mean quiescent value. The ACIS-S1 data were also used to cross-check the flare detection. This usually reduced the exposure time by 1–3 ks; in a few cases the flares occupy ∼10 ks. All point sources detected above the 3σ threshold with the CIAO tools celldetect and wavdetect were masked out. The spectral ancillary response files and redistribution matrix files were created with mkwarf and mkacisrmf, respectively.

To determine the background for each observation, we extracted a spectrum from a region $r > 2$ Mpc away from the cluster X-ray peak and fit it with a model consisting of an absorbed thermal component for the ICM, an absorbed power-law component (photon index set to 1.4) for the cosmic X-ray background (hereafter CXB), and another absorbed low-temperature thermal component (temperature = 0.2 keV and abundance = 1 $Z_\odot$) for the Galactic foreground. The quiescent particle background was calculated based on the stowed ACIS observations (e.g., Markovitch et al. 2003). The normalizations of the ICM, CXB, and Galactic components, as well as the ICM temperature and abundance, were set free in the fitting. The best-fit sample-average CXB flux is $6.8 \times 10^{-8} \text{erg cm}^{-2} \text{s}^{-1} \text{sr}^{-1}$ in 2.0–10.0 keV, which agrees well with the ASCA result reported in Kushino et al. (2002). The combined uncertainties of the CXB and Galactic components were measured to be ∼15% and ∼30% for typical high-count and low-count data, respectively. These values will be used in Section 3.2 to determine the uncertainties of the ICM mass profiles.

2.3.2. XMM-Newton

The SAS software v13.5.0 was used to screen and calibrate the data obtained with the XMM-Newton European Photon Imaging Camera (EPIC). Following the procedure in, e.g., Gu et al. (2012), we selected FLAG = 0 events and set the PATTERN ranges to be 0–12 and 0–4 for the MOS and pn data, respectively. Then we defined a source-free region for each observation and extracted light curves in both 10.0–14.0 keV and 1.0–5.0 keV bands. By investigating the light curves, we discarded time intervals contaminated by flare-like background, in which the count rate exceeds 2σ above the quiescent mean value in either of the two bands (e.g., Katayama et al. 2004; Nevalainen et al. 2005). We detected and removed point sources with the SAS tool edetect_cha-in and created the ancillary responses and redistribution matrices with xissimarfgen and xisrmfgen, respectively. The background model for the XMM-Newton data was determined in a similar way to that for the Chandra data. The EPIC spectra extracted from $r > 2$ Mpc away from the cluster center were fit with a model combining the ICM, CXB, and the Galactic foreground components. The same region in a wheel-closed data set was used to calculate quiescent particle background. The MOS1, MOS2, and pn spectra were fitted simultaneously with the same model, but leaving free the cross normalizations among the three. The best-fit sample-average CXB flux is $7.2 \times 10^{-8} \text{erg cm}^{-2} \text{s}^{-1} \text{sr}^{-1}$ in 2.0–10.0 keV, which agrees well with the Chandra result. The typical uncertainty of the X-ray background (CBX + Galactiс), ∼20%, will be later included in the error of the ICM mass profiles (Section 3.2).

2.4. Optical Data

Our prime purpose for using the SDSS-III data is to select a relatively complete set of member galaxies of each cluster and derive their radial distribution around the cluster center. For this purpose, we used all the available photometric data. For each target cluster in the sample, a sky region of projected radius $R < 2.5$ Mpc was defined, and all galaxies within this region were initially selected. For each of them, we used photometric redshift (hereafter $z_{\text{phot}}$). K-correction, and absolute magnitude $M_r$ from a table Photoz, which is based on the method of Csabai et al. (2007). Saturated star-like objects and those with blending problems were removed using object flags. To exclude $z_{\text{phot}}$ determinations with apparently bad photometry, we discarded objects when the $z_{\text{phot}}$ errors exceed 0.08(1 $+$ $z_{\text{phot}}$). This removed about 20% of the galaxies. Then, following Wen et al. (2009), the member galaxies of each sample cluster were selected with a redshift filter of $z_{\text{cl}} - 0.04(1 + z_{\text{cl}}) < z_{\text{phot}} < z_{\text{cl}} + 0.04(1 + z_{\text{cl}})$, where $z_{\text{cl}}$ is

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14 (flags and $0 \times 20$) = 0 and (flags and $0 \times 80,000$) = 0 and ((flags and $0 \times 400,000,000,000$) = 0 or psfimage, < =0.20) and (flags and $0 \times 40,000$) = 0.
cluster redshift measured with spectroscopy. As shown in Wen et al. (2012), as well as in Figure 3, the $z_{\text{spec}}$ values of the member galaxies in the SDSS sample determined in this way are consistent with their spectroscopic redshifts (hereafter $z_{\text{spec}}$), when available, within an average scatter of $\approx 0.02$–0.03.

Since the SDSS data become slightly less complete in the faint end at high redshift, it is necessary to define a redshift-dependent limiting magnitude to keep a constant completeness in the whole redshift range. Following Wen & Han (2015b), the limiting magnitude was set as $M_{r}^c \leq -20.5$, where $M_{r}^c$ is the r-band absolute magnitude corrected for passive evolution, $M_{r}(z) = M_{r}(z_e) + Qz$, and $Q = 1.16$ is an evolution coefficient. To estimate the possible light lost by the magnitude cut, we examined the composite luminosity function of nearby clusters based on the WINGS survey (Fasano et al. 2006), which is complete down to $M_r = -15.15$. Our limiting magnitude would filter out $\approx 37\%$ of the total light; when focusing on the non-dwarf members with $M_{r} < -19$ (Moretti et al. 2015), $\approx 17\%$ of the light would be lost owing to the magnitude cut. This suggests that most of the bright galaxies are retained after the filtering.

To further enhance the completeness of member galaxy selection, we incorporated the $z_{\text{spec}}$ values of the candidate galaxies measured with the SDSS DR12 (Alam et al. 2015). As shown in Figure 3(a), measurements of galaxy $z_{\text{spec}}$ are available for $\approx 80\%$ of the sample clusters, although the completeness relative to the photometric measurements decreases significantly toward higher redshifts. For each cluster, we selected objects for which the $z_{\text{spec}}$-specified recession velocity is within $\pm 2500$ km s$^{-1}$ of the cluster system velocity in the cluster frame. By further applying the limiting magnitude $M_{r}^c \leq -20.5$, a new member galaxy catalog was then created. The results of spectroscopic and photometric selections were combined to form the final catalog: all galaxies selected by $z_{\text{spec}}$ were automatically considered as members, while the $z_{\text{phot}}$ selection was used in case $z_{\text{spec}}$ was not available. The sample is roughly $z_{\text{phot}}$ based, since the $z_{\text{spec}}$ selection contributes only about 20% of the total members.

For later use in Section 3 when cross-checking results to be obtained with the $z_{\text{phot}}$-based sample, we have also constructed a pure $z_{\text{spec}}$-based sample by selecting objects with recession velocities within $\pm 2500$ km s$^{-1}$ of the cluster velocities. The $z_{\text{spec}}$ sample thus has a typical galaxy number of $\approx 100$ per cluster at $z < 0.1$, while at higher redshifts, the number becomes $\approx 50$ per cluster for most objects. In this work we focus on clusters with relatively good completeness, i.e., detected number $\geq 50$. This gives a $z_{\text{spec}}$ subsample with 80 clusters. Details of the $z_{\text{spec}}$ subsample are discussed in Section 3.1.3.

We measure the galaxy luminosity function of each cluster by binning the selected members into an interval of 0.4 in absolute magnitude and counting the number in each bin. To take into account the detection limit of the SDSS data set, we corrected the derived luminosity functions by the completeness in the observed r band. The completeness was determined by comparing the number of objects in each magnitude bin found by the SDSS with that from the WINGS database, in a sky region that has been covered by both surveys. The SDSS sample is reasonably complete, at levels of $\geq 90\%$ and $\approx 85\%$, for $M_r < 21.5$ and $M_r \sim -20.5$, respectively. After applying the correction for the detection limit, we have also corrected the luminosity function for a photometric redshift background, which is described in detail in Section 3.1, as well as for the effect of passive evolution of red-sequence galaxies. As shown in Figure 4, the obtained luminosity functions are nearly the same among the three redshift subsamples and agree well with the luminosity function of a more extensive SDSS sample reported in Wen & Han (2015b).

2.5. Determination of $R_{500}$ and the Cluster Center

To compensate for the difference of the sample clusters in their physical scales, we normalize the galaxy and ICM radial profiles to a characteristic radius $R_{500}$ (see Section 2.2 for definition). For this purpose, the $R_{500}$ values of the clusters in our final sample were determined through two independent approaches. First, we calculated a total mass distribution for each cluster based on the X-ray data and hydrostatic method and determined directly its $R_{500,X}$. Details of the mass calculation are in Section 3.3. The second method is based on an empirical scaling relation between the optical luminosity integrated over member galaxies and the cluster radius. As reported in, e.g., Popesso et al. (2004), the total optical luminosity $L_{\text{op}}$ can be used to estimate $R_{500,op}$ by

$$ \log R_{500,op} = 0.44 \log(L_{\text{op}}/10^{42}L_{\odot}) - 0.15. $$

As shown in Figure 5, the two determinations of $R_{500}$ are consistent with each other within a typical scatter of 30%. Similar scatter is seen in other works (e.g., Wen et al. 2012). We employ $R_{500,X}$ in the subsequent study, since it is closer to the original definition.

Next, the central point for each cluster was determined. Following Paper I, we set the cluster center as the centroid of the X-ray brightness, around which the X-ray emission becomes most circularly symmetric, because the X-ray deprojection analyses (Sections 3.2 and 3.3) must be precisely centralized on the X-ray centroid. The position of the brightest cluster galaxy is not employed for this purpose, because it can sometimes differ from the X-ray centroid by a few tens to hundreds of kiloparsecs (Shan et al. 2010) in cases of merging clusters. As shown in Paper I, the X-ray centroid was calculated in an iterative way for each cluster. By using a CIAO task dmstat on the central $r_1 = 5'$ region, the first centroid was
derived. Then we carried out iterations by reducing gradually the radius \( r_{i+1} = r_i - 1' \) and calculated the new centroid, as well as the shift from the previous one. The iteration was considered as converged when the shift became less than 10\(^{\prime}\). The X-ray centroid is compared in detail to the optical center in Section 3.5.1.

3. DATA ANALYSIS AND RESULTS

3.1. Galaxy Number Density Profiles

For each cluster, the galaxies selected with the \( z_{\text{phot}} \)-based method (Section 2.4) were combined to calculate surface number density profiles. These profiles might still contain a residual background component due to possible false member selection with \( z_{\text{phot}} \). To remove such a component, we calculated the mean galaxy density in a surrounding region, i.e., between 4 and 8 Mpc from the cluster center. To exclude possible large-scale structures in the background region, we further divided it into 48 sectors with equal area. Those sectors with galaxy densities larger than 2\( \sigma \) of the mean value were discarded, and a new mean background density was then obtained from the remaining sectors. The same method was used in Popesso et al. (2004). After subtracting the background for each cluster, we normalized the density profile by dividing the radius \( r \) by \( R_{500} \). This compensates the difference of cluster scale along the sky plane; to further correct it in the line-of-sight direction, we further divided the surface galaxy number density by \( R_{500} \).

In Figure 6, we show the sample-averaged radial number density profile of galaxies in comparison with the one reported in Budzynski et al. (2012, hereafter B12). The B12 profile was obtained by stacking over 50,000 clusters and groups in 0.15 < \( z \) < 0.4, and the number density of each cluster was calculated by a direct background subtraction approach instead of member selection. Though derived with quite different methods, the two profiles nicely agree with each other up to 2\( R_{500} \).

Figure 5. Comparison between the values of \( R_{500} \) of the Final Sample clusters derived with the X-ray hydrostatic estimate (ordinate) and those with a scaling of total optical luminosity via Equation (1) (abscissa). The solid line means an exact agreement, while the two dashed lines indicate 30% discrepancies.

Figure 6. Galaxy number density profile averaged over the entire Final Sample, with 68% error bar. The dashed line is the mean galaxy density profile from Budzynski et al. (2012). Both profiles are scaled to \( R_{500} \).

3.1.1. Error Modeling

In order to estimate uncertainties in the galaxy number density profiles, we considered two error sources; the Poisson error on the number of selected member galaxies in each radial bin, and the systematic uncertainty on the \( z_{\text{phot}} \) measurements. The former was calculated using the formulae of Gehrels (1986), while the latter was determined using a Monte Carlo simulation in the redshift space. For each cluster field, \( z_{\text{phot}} \) of all galaxies (both members and nonmembers) were shifted randomly within the measurement error, i.e., \( \sigma_z = 0.03 \) and 0.07 for objects with \( z_{\text{phot}} \) < 0.5 and \( z_{\text{phot}} \) > 0.5, respectively. Then, we reselected member galaxies by using the method in Section 2.4, and we recalculated the number density profiles. By 1000 Monte Carlo realizations for each cluster, we determined the systematic uncertainty to be associated with the \( z_{\text{phot}} \) measurement. Typically the systematic error is larger than the Poisson one by a factor of ~5. The two errors were combined in quadrature to make up the total error. As shown in Figure 6, the combined uncertainty is ~8% in the central bin and ~10% at outer radii.

3.1.2. Dependence on Cluster Mass and Redshift

It has been expected that the dark matter and member galaxy distributions should in general show self-similarity, i.e., the density profiles of different systems become nearly identical after some scaling (e.g., Navarro et al. 1997). To explore any additional processes beyond this simple self-similarity, we examine the possible mass and redshift dependence of the galaxy density profiles. First, we split the sample into three mass groups, i.e., \( 1 \times 10^{14} \text{–} 2 \times 10^{15} M_\odot \), \( 2 \times 10^{14} \text{–} 5 \times 10^{14} M_\odot \), and \( 5 \times 10^{14} \text{–} 2 \times 10^{15} M_\odot \). Details of the cluster mass calculation are given in Section 3.3. The choice of the three mass groups is to have a similar number of clusters in each group. The average number density profiles of the three mass groups are shown in Figure 7(b). The scaling by \( R_{500} \) has removed most of the mass dependence in the galaxy distribution. The only small difference is seen in the second radius bin, where the high-mass profile is ~20% lower than the other two profiles.

The essence of Paper I was a discovery of significant evolution of the member galaxy distribution, relative to that of the ICM. As a major step to confirm this discovery, we
calculated number density profiles individually for the entire sample of clusters and then took their ensemble averages over the redshift-sorted subsamples L, M, and H (Section 2.1). Unlike the case of mass sorting, the three galaxy distributions reveal, as shown in Figure 7(d), a clear evolution; toward lower redshifts, the profile becomes systematically steeper, with a stronger central increment within $0.2R_{500}$ and a quicker outward drop at $>0.5R_{500}$. The average ratio between subsamples L and H is $\sim 1.6$ at the central bin and $\sim 0.8$ at the outermost. A similar result was reported in Ellingson et al. (2001). As shown in their Figure 9, the populations of red-sequence and field-like + post-star-formation galaxies both become more concentrated toward lower redshifts.

3.1.3. Zspec Sample

To cross-check the above results, we repeated the same analysis on the $z_{\text{spec}}$-based sample constructed in Section 2.4. Compared to the $z_{\text{phot}}$-based sample, the spectroscopic one gives more reliable determination of cluster members, but it is more biased to high-flux galaxies ($r < 19.2$; e.g., Eisenstein et al. 2001). As described in Section 2.4, the $z_{\text{spec}}$ sample was constructed to include 80 clusters. Then, for each cluster, we calculated in the same way the radial profile of surface galaxy number density. As shown in Figure 8, the galaxy profiles are averaged over two redshift ranges, $z < 0.075$ and $z > 0.075$. Error bars show only the Poisson error on the number of galaxies. The two averaged profiles clearly exhibit different gradients: the nearby one is more centrally peaked than the distant counterpart. The central bin of nearby clusters is on average $\sim 40\%$ higher, while the outer bins are $\sim 20\%$ lower, than those of distant objects. Compared to the $z_{\text{phot}}$-measured profiles (Figure 7(d)), the $z_{\text{spec}}$ ones are systematically lower by $\sim 20\%$–$50\%$ owing to the relatively low completeness. The redshift-resolved gradients of the galaxy profiles are found consistent between the two $z_{\text{phot}}$ and $z_{\text{spec}}$ samples.

Figure 7. (a) Galaxy density profiles of all sample clusters, divided into three groups by the cluster mass. The low-mass, intermediate-mass, and high-mass systems are presented in black, red, and green, respectively. (b) The mean profiles of the three mass bins. (c) Same as panel (a), but the color specifies the redshift subsamples. The subsamples L, M, and H are shown with blue, green, and red, respectively. (d) Mean profile of the three subsamples.

Figure 8. Same as Figure 7(d), but using only the spectroscopically measured galaxies and dividing the clusters into two redshift-dependent subsamples instead of three.
3.1.4. Concentration

To quantify the difference in the member galaxy distributions, we measured the concentrations of the density profiles as described by a standard NFW model (Navarro et al. 1997), which can be written as

$$
\rho(r) = \frac{\rho_0}{r/R_s} \left(1 + \frac{r}{R_s}\right)^{-2}.
$$

(2)

Here $\rho(r)$ is the galaxy density, $\rho_0$ is the normalization, and $R_s$ is the scale radius. Since the galaxy profiles typically extend out of $R_{500}$, the ratio $c_{500} = R_{500}/R_s$ is used here as the concentration parameter. The same or a similar definition is often used to describe the structure of dark matter halos (e.g., Dolag et al. 2004; Vikhlinin et al. 2006). Then we projected the NFW model along the line of sight and fit the galaxy profiles in Figure 7(d) over the range of $0.05R_{500}$–$2.4R_{500}$. The best-fit concentrations are $2.5 \pm 0.2$, $2.1 \pm 0.1$, and $1.8 \pm 0.1$ for subsamples L, M, and H, respectively, reconfirming the evolution found in Section 3.1.2. The range of the concentration parameter agrees with those reported in previous works (e.g., Lin et al. 2004; Muzzin et al. 2007; B12).

3.2. ICM Density Profiles

Next, we address whether or not the same evolution is present in the ICM component. To determine the ICM mass distribution, the 3D ICM density profile was calculated for each cluster in a standard way based on deprojection spectral analysis with the XMM-Newton and Chandra data (e.g., Paper I). After removing point sources, we extracted spectra from several concentric annulus regions for each cluster. The radial boundaries of each annulus were determined to include sufficient net counts, i.e., 1000 and 3000 for low-flux and high-flux clusters, respectively. The extracted spectra were fit in 0.7–8.0 keV with an absorbed, single-temperature APEC model in XSPEC. All annuli were linked by a PROJCT model, which assumes the plasma parameters (e.g., emission measure and temperature) of the corresponding 3D shells individually, calculates via projection a set of X-ray spectra to be observed from the specified 2D annuli, and compares them with the actual data. For each 3D shell, the gas temperature and metal abundance were set free. When the model parameters cannot be well constrained in some annuli owing to relatively low statistics, we tied them to the values of their adjacent regions. The column density of the neutral absorber was fixed to the Galactic value given in Kalberla et al. (2005). We have also calculated spectral parameters with a more direct deprojection method presented in Sanders & Fabian (2007), which gave consistent results with the PROJCT method. All the fits were acceptable, with reduced chi-squares of $\sim0.8$–1.2 for a typical number of degrees of freedom of $\sim300$–1000.

The 3D ICM density profile was then calculated from the best-fit model normalization of each annulus. Errors were estimated by taking into account both statistical and systematic uncertainties. The former was estimated by scanning over the parameter space with an XSPEC tool steppar iteratively, while the latter was assessed by renormalizing the level of the CXB component for each region by 20% (Section 2.3). The two kinds of errors are comparable for most annuli. The sample-average gas density profile is then obtained and plotted in

Figure 9. Mean ICM density profile of our sample clusters (black data points). The 68% error is also plotted, although it is too small to be visually apparent. The reference ICM density profiles obtained by Croston et al. (2008) are shown by gray thin lines.

As shown in Figures 10(a) and (b), the $R_{500}$-scaled ICM density profiles were first grouped by the cluster mass as introduced in Section 3.1.2. At small to intermediate radii ($<0.5R_{500}$), vertical separation can be clearly seen among the average profiles of the three mass groups: the value of the high-mass group at $0.3R_{500}$ is larger than those of the medium-mass and low-mass ones by a factor of $\sim1.4$ and 1.6, respectively. At larger radii, however, the dependence becomes weaker, and the three profiles are consistent at $\gtrsim R_{500}$. Such a feature agrees well with those reported in previous observations (e.g., Croston et al. 2008) and simulation works (e.g., Borgani et al. 2004).

Then we investigate how the spatial distribution of hot ICM evolves with time. As shown in Figures 10(c) and (d), the scaled ICM density profiles were binned, as before, by their redshifts. In the inner region, the average profiles of the three subsamples show weak dependence on redshift, in contrast to the cluster mass sorting. The core density at $0.1R_{500}$ decreases by $\sim10\%$ from subsample H to subsample M, and by $\sim15\%$ from subsample M to subsample L. Similar to the case of the mass sorting, this correlation becomes further weaker and vanishes at $>0.5R_{500}$. The evolutionary trend of ICM density profiles is opposite to that found in Figure 7(d) in the galaxy density profiles.

3.3. Total Gravitating Mass

To study the most dominant mass component, the dark matter, we calculated the total gravitating mass profile for each cluster. Here we employed the standard hydrostatic mass estimates using the X-ray data (e.g., Sarazin 1988; M01). Based on the best-fit 3D gas temperature profiles $T_X(R)$ and density profiles $n_g(R)$ obtained with the deprojected analysis (Section 3.2), and assuming spherical symmetry and a hydrostatic equilibrium, the total gravitating mass within a
3D radius $R$ can be calculated generally as

$$M(R) = \frac{-kT_x(R)R}{G\mu m_p} \left( \frac{d\ln n_e(R)}{d\ln R} + \frac{d\ln T_x(R)}{d\ln R} \right),$$

where $G$ is the gravitational constant, $\mu = 0.609$ is the assumed average molecular weight, and $m_p$ is the proton mass. The associated errors were calculated combining those on the temperature and the density. By stacking the mass profiles obtained from individual clusters, we show the sample-average mass profile in Figure 11. The radius $R$ and mass $M(R)$ are normalized to $R_{500}$ and $M_{500}$, respectively.

To verify our mass measurements, we compare the results derived in this way with those reported in previous X-ray and weak-lensing studies. As shown in Figure 11, our sample-average mass profile agrees nicely with those presented in Zhang et al. (2008), which were calculated with the same X-ray technique for 37 clusters at $z = 0.14$–0.3. Most objects in their sample are also present in this work. The profiles of the two samples, this work and Zhang et al. (2008), also give similar scatter, $\approx 0.25$ and $\approx 0.2$ of the mean value at $r = 0.3R_{500}$. For a further test, we compared our mass profiles with those obtained in previous weak-lensing studies (e.g., Dahle et al. 2002). Since the lensing gives aperture mass projected along the line of sight, we have projected the X-ray mass profiles that were originally obtained in 3D. The X-ray and weak-lensing results are generally consistent with each other in most radii, albeit with some differences seen at the central $\sim 0.1R_{500}$ in several objects. These discrepancies cannot affect the key profiles shown in Section 3.4, which have innermost bins at $\sim 0.25R_{500}$.

Having derived the X-ray mass profiles, let us investigate how the matter distribution varies with the cluster mass $M_{500}$. The mass profiles for clusters in the three mass groups...
are plotted in Figures 12(a) and (b) in different colors. The total gravitating masses of the low-mass clusters thus exhibit higher concentration than those of the high-mass ones. To quantify this effect, we again calculated the concentration parameter as
\[ c_{500} = \frac{R_{500}}{R_s} \]
(Section 3.1.4), where \( R_s \) is the scale radius determined by fitting the mass density profiles with the NFW model (Equation 2). The central region \( r < 0.05R_{500} \) was excluded in the fitting, because the mass densities at these radii are often found to be biased from the universal model (e.g., Gu et al. 2012). The obtained \( c_{500} \) is plotted as a function of \( M_{500} \) in Figure 13. Thus, \( c_{500} \) decreases significantly as a function of mass; the average value of high-mass clusters is about two-thirds of that of low-mass ones. Our result is generally consistent with the expected \( c-M \) relation by cluster simulations (Dolag et al. 2004) and those from previous X-ray observations (e.g., Vikhlinin et al. 2006).

Following this, the total gravitating mass profiles were regrouped in redshift. As shown in Figure 12(d), the average profiles of the three subsamples in general trace each other, and the shape of the profiles does not vary strongly across the redshift range. The mean value of subsample L is shifted to be slightly lower, by 20%–30%, than those of the distant subsamples, probably owing to the sample selection effect as noticed in Figure 2(d). Combining Figures 12(b) and (d), it is suggested that the concentration of the dark matter halo depends primarily on the cluster mass rather than the redshift.

Figure 12. (a) Gravitating mass profiles of the Final Sample clusters. The black, red, and green points specify the low-mass, intermediate-mass, and high-mass ranges, respectively. The radii are scaled to the characteristic value \( R_{500} \). (b) Mean mass profiles for the three mass ranges. (c) Same as (a), but the subgrouping is based on the redshift; subsamples L, M, and H are represented in blue, green, and red, respectively. (d) Mean mass profiles for the three subsamples.

Figure 13. Concentration parameters of the NFW model, \( c_{500} = R_{500}/R_s \), as a function of the cluster mass \( M_{500} \). The black, red, and green points are the mean values for the low-mass, intermediate-mass, and high-mass clusters, respectively. The gray points are those measured in X-rays by Vikhlinin et al. (2006). The solid line is the \( c_{500}-M_{500} \) relation from a numerical simulation by Dolag et al. (2004), and the dotted lines specify the 2\( \sigma \) scatter around the simulated relation.

3.4. Comparison of Galaxy/ICM/Dark Matter Distributions

To compare directly the spatial distributions of the three mass components, here we calculated three types of radial profiles, i.e., galaxy number versus ICM mass ratio (hereafter GNIMR), galaxy number versus total mass ratio (hereafter
GNTMR), and ICM mass versus total mass ratio (hereafter IMTMR). As a first step, the obtained radial profiles of the mass components were transformed to have the same form and dimension. For each cluster, the galaxy number densities enclosed in each 2D radius \( r \) (Section 3.1) were integrated to obtain a quasi-continuous integral profile. To match with the optical profile, the ICM mass profile was calculated by projecting numerically the ICM density distribution (Section 3.2) along the line of sight and integrating over the same set of radius bins. As for the total mass, we converted the mass profile (Section 3.3) into the 3D mass density profile, \( \rho(R) = (4\pi R^2)^{-1}dM(R)/dR \), projected \( \rho(R) \) onto the sky plane, and integrated it over each radius \( r \) to obtain a 2D mass profile. The radially integrated profiles of the three components, now in the same form and dimension, were further normalized to their central value at \( r = 0.25 R_{500} \) (~250 kpc), because we are interested in their relative shape differences. This radius was chosen to be in line with Paper I. Finally, by dividing one component by another, we obtained the GNIMR, GNTMR, and IMTMR profiles for each cluster. The ratio profiles were again averaged over subsamples as a function of either the cluster redshifts or the masses, and they are shown in Figure 14. Uncertainties of the averaged profiles were calculated by combining in quadrature the errors of all clusters in the subsample.

### 3.4.1. Dependence on Cluster Mass and Redshift

As shown in Figure 14(b), the GNIMR profiles exhibit apparent evolution with the redshift: the average profile of subsample H is significantly flatter than that of subsample L, and that of subsample M shows an intermediate gradient. This result is a direct consequence of the opposite evolutionary trends revealed in Figures 7(d) and 10(d). Furthermore, it is consistent with the one reported in Paper I (in its Figure 13(b)), which is obtained with a very different galaxy selection method. As expected, the high-redshift \((z = 0.4 - 0.9)\) profile of Paper I appears to be clearly flatter than that of subsample H of this work. This indicates that the observed evolution might extend continuously to high redshifts. Furthermore, the shape of the GNIMR profile does not depend strongly on the cluster mass (Figure 14(a)), which proves that the redshift dependence cannot be attributed to the obvious selection bias that we tend to select more massive objects at higher redshifts (Section 2.2).

Similar to the GNIMR, the shape of the GNTMR profile also depends clearly on redshift, as shown in Figure 14(d). While the galaxies used to be less concentrated than the underlying dark matter in subsample H, they have evolved to become slightly more concentrated than the latter in subsample L. As can be inferred by comparing Figures 7(d), 10(d), and 12(d), the redshift dependences of the GNIMR and GNTMR profiles are mainly contributed by the \( z \)-dependent changes in the galaxy number distributions. On the other hand, as shown in Figure 14(c), the GNTMR profiles do not depend significantly on the cluster mass in the central \( 0.7 R_{500} \). The only difference is seen in the outer radii, where the low-mass clusters show relatively flatter GNTMR than the high-mass ones. This is mainly due to the mass-concentration effect of the dark matter halos (Figure 13).

The ICM exhibits the most extended distribution among the three components. Furthermore, as shown in Figures 14(e) and (f), the IMTMR profile depends on both redshift and the cluster mass: the X-ray size relative to that of dark matter increases with decreasing redshift and cluster mass. While the two dependences are both caused mainly by the behavior of the ICM density profiles shown in Figures 10(b) and (d), the mass dependence is also contributed by the trend of mass concentration (Figure 13).

Figure 15 provides another presentation of the above results, where the subsamples (either redshift or mass sorted) are plotted on the GNTMR versus IMTMR plane. Thus, relative to the dark matter, the stellar component kept shrinking from \( z = 0.5 \) to \( z = 0 \), while the ICM component evolved to be more and more extended even though it suffers strong radiation loss. Any mass-related effect cannot be a major driving force of this evolution, since the mass-sorted subsamples are seen to behave in quite different ways on the same plot.

#### 3.4.2. Galaxy Light versus ICM Mass Ratio

In order to examine consistency with Paper I, we may need to slightly modify the GNIMR calculation by replacing the galaxy number profile with a galaxy light profile. We calculated the rest-frame \( r \)-band luminosity of all galaxies selected by \( z_{\text{photo}} \), corrected it for passive evolution (Section 2.4), subtracted a residual background to eliminate possible false selection (Section 3.1), and integrated the galaxy light in each radius. The galaxy light versus ICM mass ratio (hereafter GLIMR) profile was then obtained for each cluster. As shown in Figure 16, the subsample-averaged GLIMR profiles exhibit strong evolution in the gradient as a function of redshift; the distribution of stellar-to-ICM ratio is more concentrated in lower redshifts. This feature is in good agreement with those obtained in Paper I. When compared to Figure 14(b), the GLIMR profiles appear to be steeper than the GNIMR ones in subsamples L and M, indicating that the brighter galaxies, hence more massive objects, preferably reside in the central regions for the nearby clusters. This feature is less significant in subsample H.

#### 3.5. Systematic Errors and Biases

Here we examine in detail the possible systematic errors and biases that might be involved in the galaxy-to-ICM comparison.

##### 3.5.1. Selection of Cluster Center

In the above studies, the center of each cluster was defined as the X-ray centroid (Section 2.5). It is, however, known that the X-ray center could be offset from the true bottom of the potential, if, e.g., the ICM is not in hydrostatic equilibrium. An obvious alternative is to identify the cluster center with the position of the brightest cluster galaxy (although this could also be biased). To assess the possible bias due to the center selection, we have calculated another set of galaxy number profiles, and hence the GNIMR profiles, by resetting the center to the brightest cluster galaxy of each cluster. For the reason described in Section 2.5, the ICM mass profiles, centralized on the X-ray centroid, were kept unchanged. As shown in Figure 17(a), the new GNIMR profiles appear to be systematically steeper than the original ones. This is because the new galaxy number profiles, with optically defined center points, often give a higher value in the inner regions. Nevertheless, the main result obtained in Section 3.4.1, i.e., the redshift dependence on the gradients of the GNIMR profiles, remains intact after the center shifting.
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**3.5.2. X-Ray Background**

The largest source of systematic error on the X-ray profiles is uncertainties in the background subtraction, and the effect must be more severe in X-ray-faint clusters. If the X-ray background was systematically oversubtracted in low-flux clusters, the ICM density at large radii would be suppressed, and the GNIMR profiles would thus be flattened. To examine this possible effect, we calculated the X-ray source-to-background ratio at $0.6R_{500}$ ($S/B_{0.6}$ hereafter) for each cluster. The mean ratios are 2.2, 2.2, and 2.1 for subsamples L, M, and H, respectively, which show no redshift dependence. These values are larger than the limiting $S/B$ in previous studies (e.g., $S/B^* = 0.6$; Leccardi & Molendi 2008). To further clarify the background effect, we divided each subsample into two groups by the mean source-to-background ratio, $S/B_{0.6} = 2.2$. As shown in Figure 17(b), such effects of the X-ray background are actually visible to some extent, but the GNIMR profiles of the low-$S/B$ and high-$S/B$ groups are still consistent with each other in each subsample, and the evolution discovered in Section 3.4.1 is seen in both groups. This indicates that the observed evolution cannot be explained by background uncertainties of the X-ray data.

**Figure 14.** (a) Mean galaxy number vs. ICM mass ratio (GNIMR) profiles of the low-mass clusters (black), intermediate-mass clusters (red), and high-mass clusters (green). (b) Mean GNIMR profiles of clusters in subsamples L (blue), M (green), and H (red). Gray points and error bars show the GNIMR profiles measured in Paper I. (c) Mean galaxy number vs. total mass ratio (GNTMR) profiles of the three mass groups. (d) Mean GNTMR profiles of the three redshift subsamples. (e) Mean ICM mass vs. total mass ratio (IMTMR) profiles of the three mass bins. (f) Mean IMTMR profiles of the three subsamples.
respective redshift-sorted subsamples (subsamples L, M, and H in blue, green, and red, respectively). The mean values of the three mass-sorted subsamples are shown with diamonds, and black, red, and green stand for low-mass, intermediate-mass, and high-mass bins, respectively.

Figure 15. Characterization of the clusters on the GNTMR vs. IMTMR plane. The data points shown with plus signs are the mean values of the three redshift-sorted subsamples (subsamples L, M, and H in blue, green, and red, respectively). The mean values of the three mass-sorted subsamples are shown with diamonds, and black, red, and green stand for low-mass, intermediate-mass, and high-mass bins, respectively.

Figure 16. Galaxy light vs. ICM mass ratio (GLIMR) profiles averaged over subsamples L, M, and H, shown in blue, green, and red, respectively. As a reference, the GLIMR results from Paper I are given in open gray symbols.

3.5.3. Cosmological Growth

As noticed in Paper I, there is one additional effect involved in the cluster evolutionary study. Since clusters are considered to grow via matter accretion onto their outskirts, the cluster scale (i.e., the mass and radius) should increase continuously over cosmological timescales. This makes \( R_{500} \) of each cluster a time-dependent quantity. To compensate for such underlying differences in the cluster evolution stage, we have attempted to recalculate the cluster profiles by defining a new scale, \( R_{500}^{\text{new}} \), the expected \( R_{500} \) that a cluster will achieve after evolving to \( z = 0 \). Based on the empirical cosmic growth function derived from N-body numerical simulation (e.g., Wechsler et al. 2002), the cluster mass is expected to increase by a factor of \( \sim e^{1.33\alpha} \) from \( z = z_0 \) to \( z = 0 \), so that \( R_{500}^{\text{new}} \) can be estimated by

\[
R_{500}^{\text{new}} \approx R_{500}^{z_0} \times e^{0.56\alpha} E(z_0)^{0.58},
\]

where \( E(z_0) \) is the cosmological evolution factor. The new radial scale is utilized to calculate a new set of GNIMR profiles shown in Figure 17(c). Although the profiles, especially those of the high-redshift subsamples, are modified to be slightly steeper, the new result is still consistent with the original one within error bars, and the GNIMR evolution still remains significant. The systematic uncertainties caused by such a “cosmological growth” effect in cluster outer regions should be \( \leq 10\% \).

3.5.4. Dynamical States

Since the sample clusters have a wide scatter in morphology, it is important to examine whether or not the detected evolution on galaxy-to-ICM profiles is created by the evolving dynamical state of clusters. Based on the spatial distributions of member galaxies, we can divide the sample into merging and relaxed clusters, while the X-ray data allow us to define cool-core and non-cool-core objects. Both of these classifications can be considered to represent the dynamical state of clusters (e.g., West et al. 1988; Allen et al. 2008). First, we utilized the two-dimensional galaxy distributions. As shown in Wen & Han (2013), unrelaxed clusters tend to exhibit asymmetrical galaxy distribution and bumpy brightness profiles in outer regions, while relaxed ones are on the opposite. Such features are quantified by three parameters, i.e., asymmetry factor \( \alpha \), ridge flatness \( \beta \), and normalized deviation \( \delta \) defined in Equations (7), (9), and (12) of Wen & Han (2013), respectively. As described in Equation (14) of their paper, the three factors can be further combined into one characteristic parameter \( \Gamma \), which is most sensitive to the dynamical state. By applying the same analysis, we define a relaxation type for each cluster. As seen in Figure 18, this provides a clear separation of the sample into 46% relaxed and 54% unrelaxed clusters. The GNIMR profiles were rebinned by the dynamical type for each redshift-dependent subsample. As shown in Figure 17(d), the differences between the two types are \( \leq 10\% \), \( \leq 20\% \), and \( \leq 8\% \) for subsamples L, M, and H, respectively. Although subsample M is somewhat affected, the evolution suggested in Section 3.4.1 is still clearly seen in both types.

Then we cross-check the above result with the X-ray data. It is well known that the relaxed clusters frequently possess cool ICM cores, while merging clusters have relatively non-cool-core objects. Both of these classifications can be further separated into cool-core and non-cool-core subsamples. Following Sanderson et al. (2006), we measured the core temperatures using spectra of the \(<0.1R_{500}\) regions and the cluster mean temperatures in the \(0.1 - 0.2R_{500}\) regions. We defined cool-core objects as those systems in which the mean temperature exceeds the core temperature at \( >3\sigma \) significance. The sample was thus divided into 44% cool-core and 56% non-cool-core objects. As shown in Figure 18, this X-ray classification is in fact close to the optical method. This is probably because the optical \( \Gamma \) factor has been somewhat calibrated to the X-ray sample (Wen & Han 2013). Then, we rebinned the GNIMR profiles by the ICM core state (Figure 17(e)). Again, the GNIMR profiles are not strongly affected by this operation, and the resulting variations are \( \leq 6\% \), \( \leq 12\% \), and \( \leq 5\% \) for subsamples L, M, and H, respectively. The detected evolution remains intact for both cool-core and non-cool-core objects.

3.5.5. Galaxy Color

Another major concern with the GNIMR evolution is possible redshift dependence of member galaxy type (e.g.,
color), coupled with the decreasing detecting efficiency of blue galaxies toward higher redshift (e.g., Csabai et al. 2003). To assess the effects of galaxy evolution, we here divided our sample, by a color–magnitude diagram employing the $g-r$ color, into red-sequence galaxies and blue galaxies. To determine the color–magnitude relation for each cluster, first we calculated the zero point of the $g-r$ versus $r$ diagram by fitting it with a straight line, assuming a uniform slope of $-0.02$ for all redshifts. This is valid because the slope is primarily a result of metallicity and has little evolution with the stellar age (e.g., Kodama et al. 1998). We utilized a robust biweight linear least-squares method (Beers et al. 1990), which is insensitive to data points that are much deviated from the relation. The fitting was repeated for a few times, as the data points outside the 3σ range of the fitted relation were rejected in the next iteration. In performing the fitting, we included those cluster members that have been confirmed by spectroscopic measurements. The best-fit zero point increases from 1.1 to 2.5 toward high redshift. The zero point versus redshift relation is in good agreement

\[ z_{\text{fit}} = 0.22 - 0.5 \]
\[ z_{\text{fit}} = 0.08 - 0.22 \]
\[ z_{\text{fit}} = 0.0 - 0.08 \]

Figure 17. Effects of various biases on the subsample-averaged GNIMR profiles. Subsamples L, M, and H are plotted in blue, green, and red, respectively. (a) Subsample-averaged GNIMR profiles calculated with an optically defined center (solid lines) and those with an X-ray-defined center (dashed fainter lines). The difference of the centers applies only to the galaxy profiles, while the ICM profiles are the same. (b) Those of clusters with high (solid lines) and low (dashed lines) X-ray signal-to-background ratio. (c) Those calculated based on the predicted evolutionary scale $R^{*}_{500}$. (d) Comparison of the clusters with high signal-to-background ratio (solid lines) and those with unrelaxed morphology ($\Gamma > 0$; dashed fainter lines). (e) Comparison between the clusters with a likely cool core (solid lines) and those without a cool core (dashed fainter lines). (f) Results using only selected red galaxies. (g) Those obtained with only blue galaxies.

Figure 18. Histograms of the morphologically relaxed clusters (green) and unrelaxed ones (blue), as a function of the ratio between the core ICM temperature and the cluster average temperature. The black and red histograms show the cool-core and non-cool-core systems, respectively.
with the metallicity sequence model reported in Kodama & Arimoto (1997).

Then, we created color–magnitude diagrams of all member galaxies of the $\zeta_{\text{phot}}$ sample and selected red-sequence members as those within $\pm 0.15$ mag of the best-fit color–magnitude relation. This simultaneously defines the blue members as those in the blue cloud. Figure 19 shows six examples of the color–magnitude selection. Since the color cut extends to the blue side below the color–magnitude relation, most red-sequence galaxies have been selected, although the red sample is inevitably contaminated by blue galaxies in the faint end. To examine the effect of blue contamination, we carried out two independent approaches. First, we selected galaxies only on the redder side of the best-fit color–magnitude relation and mirrored the distribution about the relation to determine the blue side. A similar selection was used in Gilbank et al. (2008). This affected the average galaxy surface number density profiles only by $\sim 5\%$. Second, we excluded the faint end, $M_r > -20$, from the red galaxy sample. Although the resulting number density profiles are systematically lower in the second approach, the slopes of the profiles are consistent within $1\sigma$ between the two methods. Hence, we applied the first approach to separate the red and blue galaxies.

Next, we examine the possible evolution in the spatial distributions of red and blue galaxies. In Figure 20 we show the surface number density profiles of red and blue galaxies as a

![Figure 19. Galaxy color–magnitude diagrams ($m_g - m_r$ vs. $m_r$) of six cluster examples. The selected red-sequence and blue galaxies are shown in red and blue, respectively.](image-url)
function of redshift. The red-sequence members have a higher density, together with a more centrally peaked distribution, than the blue galaxies detected within $r_{500}$. This feature has been reported in many previous works (e.g., Kodama et al. 2005; Koyama et al. 2011). When comparing the three redshift bins, the red members exhibit clear evidence of evolution that is analogous to that of the total profiles (Figure 7(d)); the nearby clusters show more peaked galaxy distributions than the distant ones. As for the blue component, a similar feature is again seen in the cluster outer regions ($r > 0.2R_{500}$). Toward the center, the blue profiles of the three redshift subsamples do not show marked differences, within rather large errors that are probably due to the relatively poor capacity of the blue galaxy selection with $z_{\text{phot}}$. As shown in Figures 17(f) and (g), similar properties are present on the GNIMR profiles of the red and blue galaxies; both exhibit signs of evolution, while those of the blue galaxies are overpowered by error bars. Hence, we prove that the observed redshift dependence on the galaxy-to-ICM profiles cannot be fully ascribed to the time evolution of the galaxy colors.

4. DISCUSSION

4.1. Evolution of the GNIMR/GNTMR/IMTMR Profiles

By analyzing the optical and X-ray data of a large, X-ray-bright cluster sample with $z < 0.5$, we studied radial distributions of the three major mass components, i.e., the member galaxies, the hot ICM, and dark matter. The galaxy membership was determined primarily with the photometric redshift. All 340 clusters were grouped into three subsamples (L, M, and H) by their redshifts. The GNIMR and GNTMR profiles were found to drop toward outer regions, with a slope that clearly steepens from subsamples H through M to L. The IMTMR profiles exhibited opposite (positive) slopes and an opposite evolution compared to GNIMR. The behavior of GNIMR was explained as a composite of the above two effects. The combined evolution on GNIMR, GNTMR, and IMTMR profiles cannot be mimicked by mass-related effects (Figures 14 and 15). As shown in Figure 16, these results quantitatively confirm and substantially expand the discoveries of Paper I, which is based on 34 galaxy clusters with $z = 0.1$–0.9.

In Section 3.5, we examined all conceivable sources of systematic errors and biases that could potentially produce, as artifacts, the apparent GNIMR/GNTMR/IMTMR evolution. However, none of them were found to affect the observed GNIMR profiles by $>20%$; the detected evolution remains intact. All the pieces of evidence consistently indicate that the radial distribution of member galaxies has indeed been evolving to be more centrally concentrated relative to the ICM and dark matter components, while the ICM has slightly expanded relative to dark matter in spite of its strong radiation loss.

The behavior of GNIMR/GNTMR could be grossly explained in at least three different ways. First, the higher galaxy density in the centers of nearby clusters could be directly created by enhanced star/galaxy formation therein toward lower redshifts. However, this idea is opposite to the currently understood evolution in the star-forming rate, which is thought to have decreased from $z \sim 2$ (e.g., Tresse et al. 2007). Furthermore, this scenario cannot explain the central decrease of the iron-mass-to-light ratio (IMLR; the mass of iron in the ICM to the galaxy light), observed in some nearby galaxy groups (Kawaharada et al. 2009). Second, the GNIMR evolution could be explained by successive addition of primordial gas onto the outermost periphery of individual clusters as they grow up. However, this disagrees with the recent X-ray results that the ICM of nearby clusters is metal enriched uniformly out to their virial radii (Fujita et al. 2008; Werner et al. 2013) with a rather constant IMLR (Sato et al. 2012). A third idea is hierarchical cluster growth combined with static galaxy evolution. As clusters grow up from inside out, their outer regions would contain more newly formed galaxies, which should be optically dimmer owing to the decreasing star formation rate toward lower redshifts. Then, nearby clusters would have lower galaxy densities in the outer regions (relative to dark matter) than distant objects. However, this view can explain neither the observed gradual increase with time in the galaxy density at the cluster central regions (Figure 7(d)) nor the evolution of the number density profiles of old population galaxies (i.e., red sequence, Figure 20).

We are hence left with the fourth view, first proposed by Makishima et al. (2001) and reinforced in Paper I, that the member galaxies have actually been falling, relative to the ICM and dark matter, toward the cluster center on a cosmological timescale. At the same time, the ICM is considered to have somewhat expanded relative to the dark matter distribution. This simple yet so-far-unexplored dynamical scenario can explain all essential results presented in Section 3. Furthermore, it can explain some important features of the metal distribution in the ICM, namely, the central decrease and outward flatness of IMLR. However, the postulated infall of galaxies clearly requires dissipation of their dynamical energies. In the next two subsections, we consider two possible origins of such energy dissipation.

4.2. Dynamical Friction

As a mechanism that causes galaxies to fall to the cluster center, we first consider dynamical friction, which occurs as gravitationally induced energy exchange between a moving galaxy and surrounding cluster media including dark matter, ICM, and other galaxies/stars (Dokuchaev 1964; Rephaeli & Salpeter 1980; Miller 1986). As shown in, e.g., El-Zant et al. (2004), the dynamical friction can create a strong energy flow.
of $10^{44}$ erg s$^{-1}$ per cluster out of the member galaxies and effectively drag the galaxies inward. To be quantitative, consider a model galaxy with a mass $m$ on a circular orbit with a radius of $R$ and an orbit velocity of $v = \sqrt{GM(R)/R}$, where $M(R)$ is the gravitating mass of the cluster inside $R$. Owing to the gravitational interaction, the galaxy receives drag force as

$$F_{DF}(R) \approx \frac{4\pi \rho(R)(Gm)^2}{v^2}$$

(e.g., Ostriker 1999; Nath 2008), where $\rho(R)$ is the cluster mass density. The angular momentum of the galaxy, $L \sim mvR$, decreases with time by $dL/dt \sim F_{DF}(R) \times R$, so that the galaxy moves in a spiral trajectory with the radius changing by

$$\frac{dR}{dt} \approx \frac{4\pi \rho(R)G^2mR}{(k+1)v^3},$$

where $k \equiv d\ln v/d\ln R$ is the logarithmic velocity gradient. Employing typical parameters, the infall distance can be written as

$$\Delta R \approx 10 \left(\frac{\rho}{10^{-4} M_\odot \text{ pc}^{-3}}\right)\left(\frac{m}{10^{12} M_\odot}\right)\left(\frac{R}{500 \text{ kpc}}\right) \times \left(\frac{\Delta t}{10^9 \text{ years}}\right)\left(\frac{v}{1000 \text{ km s}^{-1}}\right)^{-3} \text{kpc}.$$ (6)

As the most prominent characteristic of this mechanism, more massive galaxies are thus predicted to fall faster than less massive ones.

To investigate the effect of dynamical friction, it is hence important to examine whether or not the GNIMR profiles depend on the galaxy mass. By adopting the observed total mass-to-light versus luminosity relation given in Cappellari et al. (2006, their Equation (9)), we calculated the mass $m$ for each galaxy. Then, all galaxies were divided into two groups by a limiting mass $m^* = 1 \times 10^{11} M_\odot$, below which the effect of dynamical friction becomes negligible (with the infall rate $\sim 2 \text{kpc per Gyr}$; Equation (6)). Figure 21 shows the GNIMR profiles as a function of redshift for the less massive member galaxies. Although the significance is slightly lower than the original result shown in Figure 14(b), the average GNIMR profiles of the three subsamples are still clearly separated from each other. Therefore, the dynamical friction between individual galaxies and the host cluster cannot be the sole mechanism to explain the observed galaxy-to-ICM distribution. This reconfirms a conclusion already derived in Paper I.

4.3 ICM Effects

Besides the gravitational effects considered in Section 4.2, galaxies are also affected by their direct interaction with the ambient ICM. One well-known ICM effect is ram pressure caused by motion of galaxies through the ICM (Gunn & Gott 1972). The ram pressure force on a single galaxy is written as

$$F_{KP}(R) \approx \pi R_{int}^2 \rho_{ICM}(R)v^2$$

(Sarazin 1988), where $R_{int}$ is the effective interaction radius of the moving galaxy and $\rho_{ICM}(R)$ is the ICM density distribution. Since the galaxies and ICM have similar specific energy per unit mass, while galaxies have much lower specific entropy, the free energy would flow from galaxies to the ICM. As a result of the continuous ram pressure, the galaxy orbit will decay by

$$\Delta R \approx 10 \left(\frac{R_{int}}{5 \text{ kpc}}\right)^2\left(\frac{n_{ICM}}{10^{-3} \text{ cm}^{-3}}\right)\left(\frac{v}{1000 \text{ km s}^{-1}}\right) \times \left(\frac{R}{500 \text{ kpc}}\right)^{-3}\left(\frac{\Delta t}{10^9 \text{ years}}\right)^{-3} \text{kpc},$$

where $n_{ICM}$ is the ICM number density. Contrary to the dynamical friction (Equation (6)), which is more effective on more massive galaxies, the ram pressure can thus drag less massive ones more effectively. In addition, the ICM can also affect the galaxy motion via viscosity. As shown in, e.g., Nulsen (1982), the viscous force by the laminar ICM flow through a galaxy can be written as

$$F_{VIS}(R) \approx \pi R_{int}^2 \rho_{ICM}(R)v^212/R_e,$$

where $R_e$ is the Reynolds number of the ICM. This has the same form as Equation (7), and its effect would be comparable to the ram pressure when the Reynolds number of the ICM is low.

The influence of the ICM ram pressure on the stellar component of galaxies has been extensively investigated in the past decade (e.g., Kenney et al. 2004; Yoshida et al. 2004). By analyzing the combined UV to radio data of the Virgo galaxies, Boselli et al. (2009) discovered that the ram pressure would be responsible for morphological disturbances of both the gas and the young stellar population in the disk. The underlying physics may be revealed in the simulation work of Vollmer (2003) and Steinhauser et al. (2012), that the stars formed in the ram pressure tails can interact gravitationally with the parent galaxy and eventually make its disk thicker. They also discovered that, when exposed to a medium-level ram pressure, the interstellar medium would not be immediately stripped, but remains displaced to the downstream direction (i.e., backward) on a timescale of several hundred megayears, to gravitationally pull the entire galaxy backward. In such a way, the ICM ram pressure can induce drag not only on the gaseous component - but also on the other mass components (stars and dark matter, in particular) of a galaxy.
As shown in Equation (7), the strength of the galaxy–ICM interaction is proportional to the galaxy velocity squared and the ambient gas density. Since the galaxy velocity measurements are less complete with the current data (Section 3.1.3), it is natural to examine whether the GNIMR evolution depends on the ICM density. Here we calculated the ICM density at $r = 0.5R_{500}$ for each cluster using the X-ray spectroscopy results (Section 3.2) and divided each subsample into two parts by a dividing value $\rho_{\text{ICM}}^0 = 10^{-3} \text{ cm}^{-3}$. As shown in Figures 22(a) and (b), in a high-$\rho_{\text{ICM}}$ environment, galaxies have in fact been concentrated to within $\lesssim 0.1R_{500}$, whereas such an evolution in galaxy distribution is weaker in low-$\rho_{\text{ICM}}$ clusters. As a result, the GNIMR profiles of the high-$\rho_{\text{ICM}}$ objects (Figure 22(d)) exhibit nearly the same pattern of evolution as the sample-average one shown in Figure 14(b), while the low-$\rho_{\text{ICM}}$ profiles (Figure 22(c)) suggests a less significant evolution. Therefore, in $z < 0.5$, the clusters with relatively lower ICM densities have a slightly smaller number of galaxies infalling toward the center than those with higher ICM densities. These results support the prediction in Paper I, that the galaxy–ICM interaction contributes significantly to the observed GNIMR evolution. On the other hand, for subsample H, the average GNIMR profile of the relatively low $\rho_{\text{ICM}}$ clusters appears to be as steep as the one of high-$\rho_{\text{ICM}}$ objects, indicating that the galaxy-to-ICM concentration was already in place at $z \sim 0.5$ for the entire $\rho_{\text{ICM}}$ range considered in this work.

The above scenario describes the interaction between individual galaxies and their environment. In reality, galaxies are not always traveling alone; a fraction of infalling galaxies are bound to subcluster-scale groups before fully merging with the cluster. In such a group-in-cluster configuration, both the galaxy–cluster and group–cluster interactions are naturally expected. Based on a weak-lensing study in the central 1.6 Mpc region of the Coma Cluster, Okabe et al. (2014) discovered 32 subcluster-scale halos, each with a mass of several times $10^{12}$–$10^{13} \, M_{\odot}$. According to Equation (6), the dynamical friction can affect the clustocentric distances of these halos by 50–150 kpc Gyr$^{-1}$. Meanwhile, Sanders et al. (2013) observed the same cluster in X-rays and discovered two coherent linear ICM structures up to 650 kpc. These structures provide smoking-gun evidence for strong direct interactions, e.g., ram pressure, between infalling subclusters and cluster ICM on a timescale of several megayears. The interaction between the ICM and individual galaxies in the group may also be enhanced owing to a large relative velocity $v$. As reported in Yagi et al. (2015), most of the galaxies showing features of strong interaction, e.g., extended H$\alpha$ emission, in two $z = 0.4$ clusters are found to belong to infalling subclusters. In addition, in such subclusters, pairs of galaxies orbiting each other will lose their angular momentum, via interaction with the cluster environment, and will merge together on the infalling timescales. Thus, the interaction is considered to enhance morphological evolution of member galaxies.
4.4. An Energy Flow on Cosmological Timescales

The observed persistent and large-scale matter infall, first predicted by M01 and confirmed through Paper I and the present study, is expected to generate a large energy flow from galaxies to the ICM. The energy loss rate of a single galaxy in the interactions can be represented by

\[ L = (F_{\text{DF}} + F_{\text{RP}}) \times v. \]

Employing typical values, the energy flow per galaxy can be written as

\[
L \approx 1 \times 10^{41} \left( \frac{\rho}{10^{-4} M_\odot \text{ pc}^{-2}} \right) \left( \frac{m}{10^{11} M_\odot} \right)^2 \left( \frac{v}{1000 \text{ km s}^{-1}} \right)^{-2} \\
+ 2 \times 10^{41} \left( \frac{R_{\text{int}}}{5 \text{ kpc}} \right)^2 \left( \frac{n_{\text{ICM}}}{10^{-3} \text{ cm}^{-3}} \right) \\
\times \left( \frac{v}{1000 \text{ km s}^{-1}} \right)^3 \text{ erg s}^{-1}. \quad (10)
\]

The energy thus transfers silently from thousands of galaxies to the environment, creating a flow of \(10^{44} - 10^{45}\) erg s\(^{-1}\) per cluster on a timescale of several gigayears. This makes it one of the largest energy events in the universe, comparable to outbursts of most luminous AGNs and to the X-ray luminosity from each cluster. Furthermore, the spatial distribution of \(L\) is highly concentrated in the cluster center, where the galaxy–ICM term \(F_{\text{RP}}\) has a larger contribution than the gravitational term \(F_{\text{DF}}\) owing to larger galaxy velocities therein. As a result, more than half of the energy is expected to transfer directly to the ICM component, which significantly contributes to the ICM heating to suppress cooling flows (M01). Furthermore, this can explain how the ICM component has been evolving to achieve larger angular extents than the galaxy and dark matter components in nearby clusters (Figure 14(f)). Such a gas heating by galaxy infall has been successfully reproduced with recent numerical simulations (e.g., Asai et al. 2007; Ruszkowski & Oh 2011; Parrish et al. 2012).

Besides the possible ICM heating, the current scenario has several other important implications. First, it immediately explains how the centrally concentrated galaxy distributions in nearby clusters were formed. Second, it provides an important clue to the origin of the environmental effects and evolution seen in member galaxies (Butcher & Oemler 1984; Wen & Han 2015a). Third, it may potentially explain the observed central drop in IMLR and the large-scale \(> R_{\text{vir}}\) abundance uniformity in the ICM (Werner et al. 2013).

5. CONCLUSION

In Paper I, we measured radial profiles of stellar light and ICM mass for 34 galaxy clusters with redshifts of 0.1–0.9 and detected, for the first time, a significant evolution that the member galaxies get more centrally concentrated with time relative to the ICM and dark matter. By using the SDSS photometric and Chandra/XMM-Newton data, we now have greatly enhanced our study by constructing an unprecedented cluster catalog with 340 X-ray-bright clusters, although the redshift range \((z < 0.5)\) became somewhat narrower. Using this new sample, we have quantitatively confirmed and reinforced the results reported in Paper I. While the member galaxies are continuously falling to the center relative to the ICM and dark matter, the ICM has been slightly expanded relative to the dark matter even though it keeps radiating. The observed evolution cannot be explained by various systematic errors or z-dependent selection biases. The galaxy infall is seen in both more massive and smaller member galaxies and is enhanced in clusters with higher ICM densities. Therefore, the observed effects cannot be explained by dynamical friction alone, but require more direct galaxy–ICM interaction. These interactions are considered to create a large energy flow of \(10^{44} - 10^{45}\) erg s\(^{-1}\) per cluster from the member galaxies to their environment, which is estimated to continue over cosmological timescales. The present results have several important implications for the evolution of clusters and galaxies within them.

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