The dynamical state of Abell 2399: a bullet-like cluster

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ABSTRACT

While there are many ways to identify substructures in galaxy clusters using different wavelengths, each technique has its own caveat. In this paper, we conduct a detailed substructure search and dynamical state characterization of Abell 2399, a galaxy cluster in the local Universe ($z \sim 0.0579$), by performing a multiwavelength analysis and testing the results through hydrodynamical simulations. In particular, we apply a Gaussian mixture model to the spectroscopic data from SDSS, WINGS, and OmegaWINGS Surveys to identify substructures. We further use public *XMM–Newton* data to investigate the intracluster medium (ICM) thermal properties, creating temperature, metallicity, entropy, and pressure maps. Finally, we run hydrodynamical simulations to constrain the merger stage of this system. The ICM is very asymmetrical and has regions of temperature and pressure enhancement that evidence a recent merging process. The optical substructure analysis retrieves the two main X-ray concentrations. The temperature, entropy, and pressure are smaller in the secondary clump than in the main clump. On the other hand, its metallicity is considerably higher. This result can be explained by the scenario found by the hydrodynamical simulations where the secondary clump passed very near to the centre of the main cluster possibly causing the galaxies of that region to release more metals through the increase of ram-pressure stripping.

Key words: galaxies: clusters: general – galaxies: clusters: intracluster medium – galaxies: evolution.

1 INTRODUCTION

The standard hierarchical model implies that galaxy clusters are formed by the accretion of galaxies and groups of galaxies (Press & Schechter 1974). Observational evidence at different wavelengths shows that this process continues to occur in the local universe (e.g. Hurier, Singh & Hernández-Monteagudo 2017; Hurier 2019) and galaxy clusters are still accreting groups and even suffering major mergers. Cosmological simulations show that a massive ($\sim 10^{15} \, h^{-1} \, \mathrm{M_\odot}$) cluster at $z = 0$ has typically accreted $\sim$40 per cent of its galaxies ($> 10^9 \, h^{-1} \, \mathrm{M_\odot}$) from groups (more massive than $10^{13} \, h^{-1} \, \mathrm{M_\odot}$) (McGee et al. 2009). Observations further suggest that 10–20 per cent of low-z clusters are undergoing cluster–cluster mergers (Katayama et al. 2003; Sanderson, Edge & Smith 2009; Hudson et al. 2010) and, therefore, have not yet reached the relaxed dynamical state.

The identification of unrelaxed clusters is essential to investigate the possible influence they exert on the scattering of scaling relations (Ventimiglia et al. 2008) and the uncertainties of cosmological parameters obtained through galaxy cluster studies (Dooley et al. 2014). Also, that is fundamental to understand how interactions impact the evolution of galaxies (Stroe et al. 2014; Kelkar et al.

2020). However, conventional observational methods to study cluster dynamics are limited.

An important issue in the assessment of a cluster dynamical state from optical data is that we can normally measure only projected quantities. Ideally, one would use 3D information, instead of only RA and Dec. Redshift surveys help to mitigate this problem, but those should have high completeness so that the identification of substructures and interactions is reliable. Even so, radial velocity has been used as a proxy for a third spatial coordinate in several substructure studies (e.g. Einasto et al. 2010; Ribeiro et al. 2013; Cohen et al. 2014; Jaffé et al. 2016).

In the case of the X-ray substructure studies, important information about the dynamical state of the gas could be derived from the analysis of spectral maps, instead of using solely the surface brightness map. Recently, in a sample of 53 galaxy clusters, Laganá, Durret & Lopes (2019) found that 16 clusters classified as cool-core (CC, a drop in the temperature profile towards the centre that is common in relaxed clusters; Ichinohe et al. 2015) by Andrade-Santos et al. (2017) according to concentration, cuspiness, and central density of the surface brightness map have instead characteristics that indicate interactions such as asymmetries and discontinuities in their temperature maps.

Commonly, cases of cluster mergers that are not obvious in the optical data are clearly seen in the analysis of the X-ray images. The interactions between clusters are extremely powerful phenomena that

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release a high amount of energy (of the order of 10^{64} erg, see Sarazin 2002). Such events can cause the intracluster medium (ICM) to exhibit large deviations from spherical symmetry (Roettiger, Burns & Loken 1996), an increase of the temperature in some specific regions via gas compression (Markevitch et al. 2002, 2005; Russell et al. 2010; Macario et al. 2011; Owers et al. 2011; Russell et al. 2012), destruction of the cool-core, and a displacement between the Brightest Cluster Galaxy (BCG) and the X-ray peak (Lopes et al. 2018).

However, since the thermal bremsstrahlung emission of the ICM depends quadratically on the gas density, not all galaxy clusters are luminous enough in X-rays to allow a substructure study at large clustercentric radii; to worsen the situation, the gas can also suffer from projection effects that make it virtually impossible to identify substructures in the line of sight since the ICM is optically thin.

Numerical simulations have often been employed to get a deeper view of cluster mergers (e.g. ZuHone, Markevitch & Johnson 2010; ZuHone 2011; Vazza, Roediger & Brüggen 2012; ZuHone et al. 2013; Iapichino, Federrath & Klessen 2017; Schmidt et al. 2017). In particular, they are also applied to model dissociative merging clusters, in which the gas has been detached from the dark matter peaks, helping to constrain parameters such as the mass ratio, angle of inclination of the collision with respect to the line of sight, and also, the velocity of propagation of the shock fronts in the ICM (e.g. Springel & Farrar 2007; Mastropietro & Burkert 2008; van Weeren et al. 2011; Machado & Lima Neto 2013; Lage & Farrar 2014; Machado et al. 2015; Monteiro-Oliveira et al. 2017). Furthermore, numerical simulations predict that interacting clusters experience an increase in the ram pressure that can originate gas-stripped galaxies with a jellyfish morphology (e.g. Vijayaraghavan & Ricker 2013; Ruggiero et al. 2019). This enhancement occurs because the ram pressure grows with the square of the relative velocity between the ICM and the galaxies, which is higher in merging clusters. Merger events also increase the density of the ICM driving an increase in the ram pressure as well.

In this paper, we study in detail the dynamical state of Abell 2399, a galaxy cluster located at a filamentary intersection of the Aquarius–Cetus supercluster at $z = 0.0579$ (Bregman, Dupke & Miller 2004). It was classified as BM Type III by Bautz & Morgan (1970), meaning that there are no central dominating cluster galaxies. M_{200} and L_X are, respectively, $4.31 \times 10^{14} M_\odot$ and 0.636×10^{44} erg s^{-1} (Lopes et al. 2009b). This galaxy cluster has been the subject of several studies on its dynamical state obtaining conflicting conclusions summarized below.

Abell 2399 was selected in this paper for presenting traces of multimodality in its velocity distribution and in the plane of the sky. In addition, its temperature profile is not isothermal, suggesting that some nongravitational processes, such as an interaction, could be playing a role here. The X-ray morphology of A2399 is suggestive of a possible collision event (e.g. Böhringer et al. 2010; Mitsuishi et al. 2018) and from the inspection of its optical data, we found more than one possible BCG, which is another indication of interaction.

Some authors have classified A2399 as a single cluster, with no interactions. Flin & Krywult (2006), for example, applied a transform wavelet code in an attempt to detect substructures in the projected distribution of the galaxies of A2399 and other 182 Abell clusters. In order to do that, they first combined data from the 10-inch photographic plates taken with the 48-inch Palomar Schmidt Telescope and from the Digitized Sky Survey. Their code did not detect any substructure in this 2D analysis and the system was classified as unimodal.

Abell 2399 was also studied as part of the WIde-field Nearby Galaxy clusters Survey (WINGS), which had as the main goal to investigate the galaxy populations in clusters in the local universe ($0.05 \leq z \leq 0.07$) and the environmental effects on their stellar populations. This survey obtained spectroscopic data for 48 clusters including A2399. On the investigation of substructures in the WINGS clusters, Ramella et al. (2007) applied a kernel-adaptive code, also in the spatial projected distribution of galaxies. In agreement with Flin & Krywult (2006), they found no substructures in the A2399 two-dimensional analyses.

The same conclusion was obtained by Moretti et al. (2017), who applied a $\pm 3\sigma$ clipping method to assign membership to the OmegaWINGS sample. Their analysis did not indicate another structure separated in the line of sight of this system.

On the other hand, other authors have found evidence in different wavelengths that A2399 is a cluster merger. Böhringer et al. (2010) inspected the X-ray emission of A2399 (or RXCJ2157.4-0747) with good details in the *XMM–Newton* image. The authors identified two well-separated components within R_{500} (11.1 arcmin). Contrary to the classifications derived from the optical data available at that time, they classified this cluster as bimodal. It is noteworthy that A2399 has very diffuse, low-surface brightness regions in its centre that make it difficult to automatically identify its local maximum (Böhringer et al. 2010).

In the same direction, Fogarty et al. (2014) reported a decrease in the fraction of elliptical galaxies with a slow rotation towards the denser regions of the cluster, while the two central galaxies are classified as having a fast rotation (one of them is the brightest cluster galaxy, BCG). They then suggested that the slow-rotation galaxies belong to infalling groups and are entering the cluster for the first time. They also state that this is the most complicated cluster in their sample of three clusters and that it is most likely to be a merger.

More recently, Owers et al. (2017) used data from the Sydney-AAO Multi-Object integral field Spectrograph Galaxy Survey (SAMI-GS) to study substructures in galaxy clusters. When performing tests of substructures in 1D (velocity), 2D (sky coordinates), and 3D (velocity and sky coordinates), they found evidence of substructure in all three dimensions. They also found more than one candidate for the central cluster galaxy and that the galaxy closest to the centre of the galaxy density distribution is not the BCG.

Analysing the *XMM–Newton* and *Suzaku* images together, Mitsuishi et al. (2018) found several pieces of evidence for a merger scenario in the plane of the sky. Among them are the offset between temperature peaks and X-ray emission peaks, a significant difference between the location of the optical substructures and X-ray clumps, and discontinuity in X-ray surface brightness and temperature profile in the secondary clump with the possibility of a cold front. According to them, a merger scenario also explains the high entropy in the central region of the cluster found by the authors. In summary, substructure analyses on A2399 have been contradictory, making of this cluster an interesting object for further analysis and determining its dynamical state.

The main goal of this paper was to probe the dynamical state of A2399 circumventing the weaknesses of the optical and X-ray merger identification techniques alone. In order to do that, we applied a careful substructure analysis combining optical and X-ray data to this merger candidate cluster. Optical substructures were identified by using a Gaussian mixture model (GMM), an algorithm that assumes that all the data points are generated from a mixture of a finite number of Gaussian distributions (Duda & Hart 1973), in the high-completeness spectroscopic data from SDSS-DR12 (Alam et al. 2015), WINGS (Cava et al. 2009), and OmegaWINGS Surveys (Moretti et al. 2017). Then, we compared the results with the distribution of the thermodynamic properties of the gas. The

parameters obtained from both analyses were used as input for hydrodynamical numerical simulations that we performed with the purpose of reproducing the observed temperature map and gas distribution. We intended to test the hypothesis that those properties of A2399 might be explained by a bullet-like collision. In Section 2, we present details about the sample and data reduction in the optical and X-ray. In Section 3, we perform a careful substructure analysis in the optical and X-ray and then compare both analyses. Finally, in Section 4, we use the physical parameters obtained from the optical analysis to test if a bullet-like collision scenario would reproduce well the temperature map of the ICM obtained by us in this work.

Throughout this paper, we adopt a ΛCDM cosmology with $H_0 = 70\,\mathrm{km\,s^{-1}\,Mpc^{-1}}$, $\Omega_M = 0.27$, $\Omega_\Lambda = 0.73$. Error bars are quoted within 1σ.

2 SAMPLE AND DATA REDUCTION

We analyse optical and X-ray data available in the region covering the galaxy cluster A2399 in order to look for substructures and clues to its past merger history. The cluster was selected as an interacting candidate cluster after visually inspecting the velocity distribution along with the projected position versus velocity phase space (PS) diagram of the cluster galaxies retrieved from the supplemental version of the Northern Sky Optical Cluster Survey (NoSOCS, Lopes et al. 2004, 2009a). The projected PS diagram is a schematic way of analysing galaxies in the different cluster's regions (infalling, virialized, and backsplash) and understanding the assembly history of galaxy clusters (Jaffé et al. 2015, 2016; Rhee et al. 2017). The distribution of the cluster galaxies in single virialized clusters has a shape that resembles a trumpet (Diaferio & Geller 1997; Diaferio 1999; Oman, Hudson & Behroozi 2013; Haines et al. 2015). When searching for interacting candidates for further analysis, we looked for PS diagrams that have groups of galaxies outside of this trumpet region.

2.1 Optical data

The spectroscopic follow-up program WIde-field Nearby Galaxy-cluster Survey (WINGS-SPE) is a large survey that obtained 48-cluster multifibre spectroscopy between $z = 0.04 - 0.07$ in an intermediate resolution of 6–9 Å (Cava et al. 2009). Its parent sample, the WINGS-OPT, consists of 77 galaxy clusters, with optical imaging, in B and V bands, X-ray selected in a distance from the galactic plane of ($b \geq 20°$; Fasano et al. 2006). Each cluster was observed in a field of view of $34' \times 34'$, $\sim 0.5\,R_{200}$ at $z = 0.05$. In order to extend the sample study to the infall regions of the clusters, the WINGS team observed 46 clusters of the original sample with OmegaCAM/VST imaging instrument in u, B, and V bands covering $1° \times 1°$ of each cluster, $\sim 1.5\,R_{200}$ at $z = 0.05$ (Gullieuszik et al. 2015). Recently, the OmegaWINGS survey obtained spectroscopy of 33 clusters of its sample with the AAOmega spectrograph (Smith et al. 2004; Sharp et al. 2006), with a spectral resolution of 3.5–6 Å full width at half-maximum (Moretti et al. 2017).

We retrieved spectroscopic data from WINGS and OmegaWINGS in a radius of 2.6 Mpc, $\sim 1.68\,R_{200}$ or 37.89′ (R_{200} from Biviano et al. 2017), around A2399's BCG, located at RA $=329.\!^{\mathrm{a}}37258$ and Dec. $=-7.\!^{\mathrm{a}}79569$. We obtained 289 galaxies with $0.05 \leq z_{\rm spec} \leq 0.07$.

We also retrieved spectroscopic data from the Sloan Digital Sky Survey Data Release 12 (SDSS-DR12; Alam et al. 2015) within the same radius and redshift range we used to select the WINGS and OmegaWINGS data. Only the data with flags `p.clean = 1` and `s.z.warning = 0` were used. Those SDSS-DR12 flags ensure a

data set with reliable measurements. The first one removes duplicate objects, galaxies with deblending problems, interpolation problems, and suspicious detections such as saturated sources. The latter ensures that the redshift has been measured with a good signal to noise (S/N), has strong emission or absorption lines, and that the largest part of the pixels is reliable so that we can rely on the redshift measure.

Although in SDSS-DR12 113 galaxies were retrieved, yet in the same region of WINGS and OmegaWINGS for equal redshift range, we obtained 289 galaxies with $z_{\rm spec}$. After performing a match between the two spectroscopic catalogues with a 1 arcsec error, we obtained a combined sample of 299 galaxies. Duplicated galaxies from the SDSS catalogue were excluded from our analysis. Considering only WINGS/OmegaWINGS spectra, the spectroscopic completeness is 60.9 per cent (down to $m_V = 20$ and within a range of $B - V$ between 0.4 and 1.8). After including additional SDSS spectra (which reach an $m_V \sim 18$), the completeness increases to 69.3 per cent. The WINGS and OmegaWINGS spectroscopic data are 1.5 mag deeper than SDSS. This makes these surveys an appropriate source of data for the careful substructure analysis we performed in this work. The left-hand panel of Fig. 1 shows the distribution of the WINGS and SDSS data in the plane of the sky, while in the right-hand panel, the velocity distribution of the two samples is shown along with the final combined sample without duplicate galaxies.

2.2 X-ray data

The *XMM–Newton* observation (ID: 0654440101) was obtained on 2010 June 7. The thin 1 filter was used for all the cameras, in Prime full window for MOS and Prime full window extended for PN.

We reduced the data with the SAS version 17.0.0 (June 2018) and calibration files were updated to 2018 March.

In order to filter background flares, we applied a 2-σ clipping procedure using the light curves in the [1–10] keV energy band. After cleaning the light curves, the resulting exposure times were approximately 53 ks for MOS1, 59 ks for MOS2, and 23 ks for PN. Subsequently, the redistribution matrix file (RMF) and ancillary response file (ARF) were generated with the SAS tasks *rmfgen* and *arfgen* for each camera and for each region that we analysed.

Finally, we had to consider the background contribution of each detector. To account for this, we obtained a background spectrum in an external annulus within 9 and 11 arcmin away from the cluster centre in the [10–12] keV energy band. Then, we compared these spectra with the one obtained from the Read & Ponman (2003) blank sky in the same region and energy band. The annulus background was divided by the blank sky background to obtain a normalization parameter for each detector that was used in all spectral fits (as already presented in Laganá et al. 2008).

Point sources were detected by visual inspection, confirmed in the High Energy Catalogue 2XMMi Source, and excluded from our analysis.

2.2.1 Spectral fit

The cluster's X-ray emission was modelled in XSPEC v12.10.0 (Arnaud 1996), with APEC single temperature plasma model (Smith et al. 2001) using the AtomDB v3.0.9 and WABS to model photoelectric absorption.

We acquired the spectra in the [0.7–5] keV band, with the low-energy cutoff selected such as to diminish contamination from the residual soft emission of the Milky Way. The upper-temperature cut was made in order to increase the S/N ratio since for low-temperature
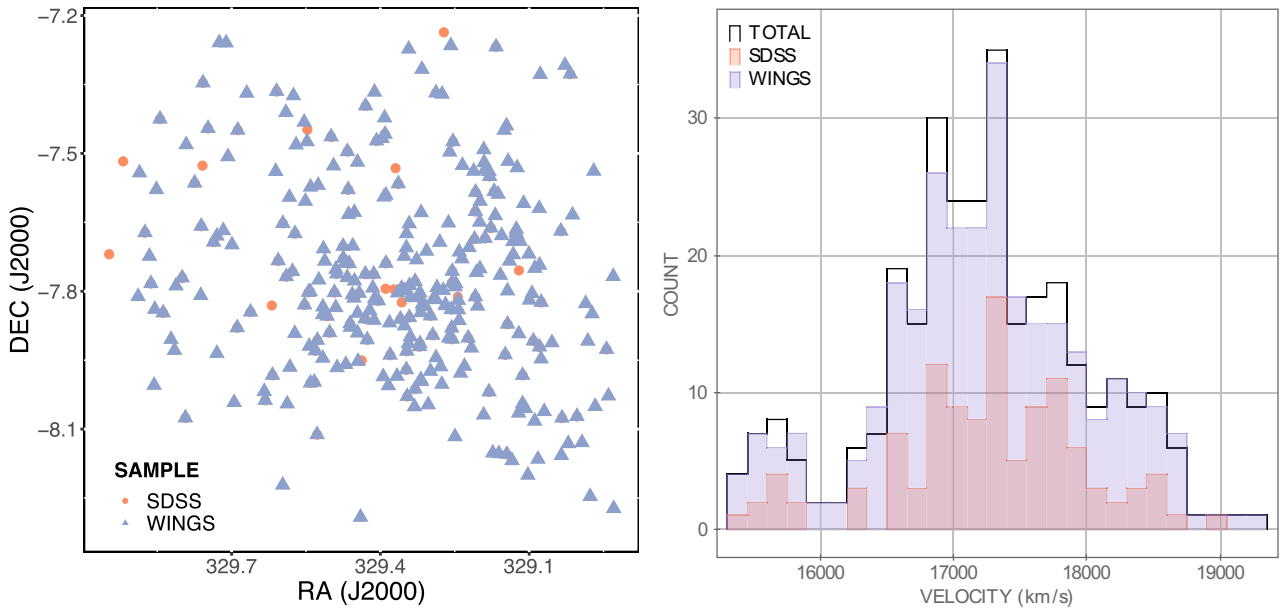
Figure 1. Left-hand panel: The distribution of WINGS and SDSS samples in the plane of the sky. Right-hand panel: The velocity distribution of the two matched samples, SDSS (red) and WINGS (blue), and the resulting sample (black). The duplicated galaxies from SDSS were excluded from our analysis. The SDSS sample alone has gaps in the velocity distribution that could bias the substructure identification.

clusters, such as A2399 with $<kT> = 2.6$ keV (Mitsuishi et al. 2018), no emission is expected above this limit. The energy band from 1.3 to 1.9 keV was excluded to prevent any influence from the strong Al and Si lines.

The redshift was fixed to $z = 0.0579$ (Biviano et al. 2017) for each fit, and the hydrogen column density, nH, was frozen to the value listed in the Leiden/Argentine/Bonn value of 3.05×10^{20} cm^{-2}. We maintained the gas temperature (kT), metallicity (Z), and normalization (N_{apec}) as free parameters in the APEC plasma model. The metallicity was obtained from Asplund et al. (2009) in solar units.

2.2.2 *Projected maps*

A detailed way of getting information on the cluster's history and dynamical state is obtaining spectral maps for different thermodynamic parameters. Laganá et al. (2019) showed that very often clusters classified as cool-core have non-relaxed gas, and non-cool-core clusters have their gas relaxed. Therefore, the mere classification according to the cool-core presence is not sufficient for a reliable determination of the cluster dynamical state. It is then necessary to use detailed techniques such as spectral maps to understand the history of the cluster through comparison with hydrodynamic simulations, as we will see in Section 4.

To obtain the spatially resolved maps, we divided the data into small regions from which spectra were extracted. The 2D maps were made in a grid, where each pixel is 512×512 *XMM–Newton* EPIC physical pixels, corresponding to 12.8 arcsec × 12.8 arcsec. We set a threshold of 900 counts per grid region after background subtraction to assure an S/N of at least 30. The spectra of the MOS1, MOS2, and PN detectors were simultaneously fitted and the best temperature and metallicity values were attributed to the central pixel of each grid (as described in Durret et al. 2010; Durret, Laganá & Haider 2011; Laganá et al. 2015; Laganá et al. 2019). It is important to highlight that since the spectral extraction regions are typically larger than the pixel map, individual pixel values are not independent.

3 SUBSTRUCTURES IDENTIFICATION

3.1 *Mclust*

The *mclust* method is a quite popular *R* package that uses a GMM along with hierarchical clustering to determine the number of clusters in a data set and to assign the data points memberships. In addition to classifying, it provides uncertainties and estimates the density-based finite Gaussian mixture modelling (Scrucca et al. 2016).

GMMs are probabilistic models that fit a finite number of Gaussian distributions with unknown mean and covariance to the data set. These models make use of an Expectation–Maximization (EM) algorithm that acts similarly to *k*-means, but with the advantage of better fitting asymmetric and overlapping data sets, since the only parameter used by *k*-means is the average and it uses Euclidean distances.

At a basic level, the EM algorithm maximizes the likelihood of the density estimation calculating at each iteration the expectation of the log-likelihood function using the current estimate of the parameters (Expectation step) and updating the parameters to maximize the log-likelihood function (Maximization step). The first step of this algorithm considers fixed the parameters that describe the Gaussian and calculates for each point the probability of belonging to each cluster. In the second step, maximization, the probability of each point belonging to a particular cluster, is considered fixed and the Gaussian parameters are weighted by the sum of the probabilities of each point belonging to the clusters. The two steps of the EM increase the log-likelihood of the model, which is the logarithm of the sum of the probability of each point belonging to a given Gaussian of the mixture. The iterations continue until converging.

In summary, *mclust* calculates the log-likelihood for each iteration and then computes the most effective approximation to estimate the mixtures by varying the shape, orientation, and volume of multidimensional Gaussians. *Mclust* uses the Bayesian Information Criterion (BIC) to decide between the 16 templates fitted by the code (see Appendix A). In the model fitting, the likelihood can be increased
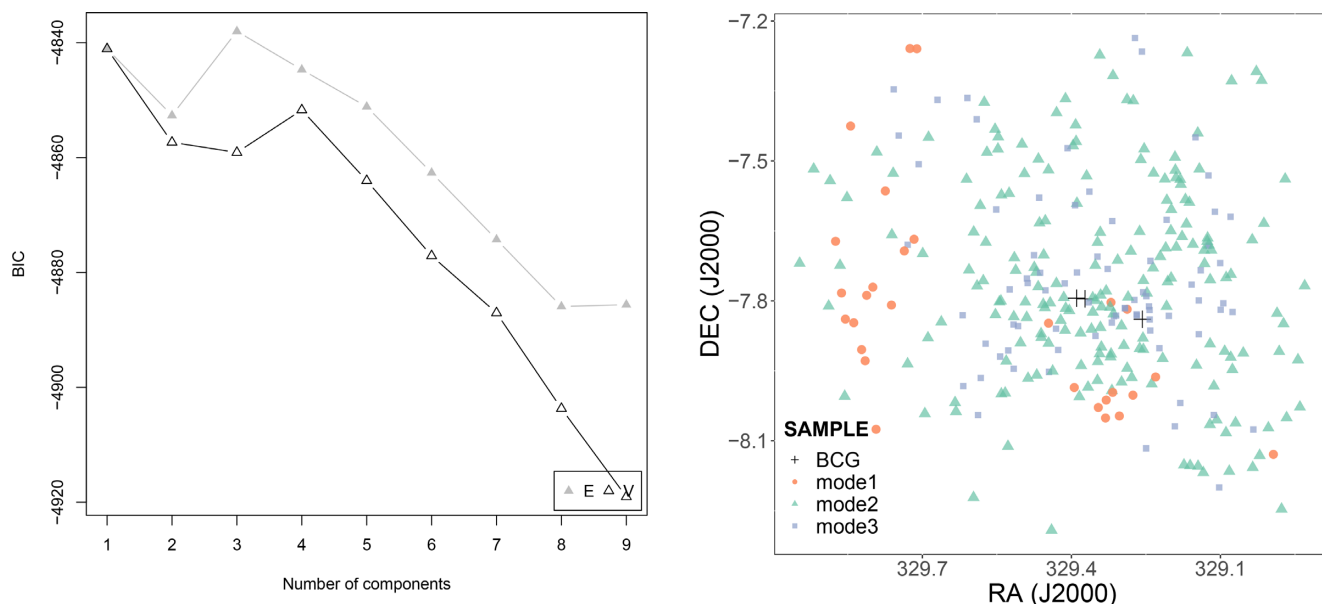
Figure 2. 1D analysis. Left-hand panel: BIC analysis for the different models fitted by *mclust*. Model E with three velocity components is the one to better fit the data. Right-hand panel: Distribution of the velocity groups found by the GMM in the plane of the sky. The black crosses mark the positions of the BCGs.

by adding parameters, but this may cause overfitting and more groups than those that actually exist to be found. The BIC solves this issue by inserting a penalty term for the number of parameters in the model. Although the model selection using the BIC cannot be done blindly because, for models with distinct numbers of free parameters that have their log-likelihood ratio sequence bounded in probability, BIC follows the parsimony principle leading the probability of choosing the model with the fewer parameters to become higher as the number of parameters increases (Findley 1991).

For the optical data, we performed a *mclust* analysis on 1D on the velocity distribution, 2D on RA and Dec., and finally on 3D using RA, Dec., and velocity. Since RA and Dec. coordinates vary by a few degrees, while the velocity coordinates vary by thousands of kilometres per second, we applied the *scale* function of *R* to centre and scale the units in the 3D analysis. The modes identified by the *mclust* in 3D were then isolated and had their centroids calculated for RA, Dec., and z_{spec}. In this work, we chose to use the results of the *mclust* 3D classification for the comparison with the ICM distribution. Unlike the 1D and 2D results, the groups found in the 3D analysis are reliable in the sense that one is well distinguishable in the PS diagram and the other seems to be associated with the western clump observed in the X-ray.

One caveat of this method is that the data distribution is not necessarily described by a mixture of Gaussians. Not all the galaxies in the sample belong to the group that they were assigned by the GMM. In order to neutralize this forced membership attribution, we ran a *shifting gapper* technique (Fadda et al. 1996; Lopes et al. 2009a) to identify objects that are not linked to any structure. In the end, we obtained the cluster's and the groups' members and galaxies that are cinematically in the region but with low connection to the identified groups, the interlopers.

Our *mclust* analysis was performed in the range $cz = 14\,000$–$21\,000\,\mathrm{km\,s^{-1}}$ within $\sim 1.5\,R_{200}$ (~ 2.6 Mpc) around the BCG in the combined SDSS-DR12, WINGS and OmegaWINGS data sets (see Einasto et al. 2012a, b; Ribeiro et al. 2013; Monteiro-Oliveira et al. 2020, for examples of *mclust* applied in substructure analyses).

3.1.1 1D analysis

In a relaxed cluster, a Gaussian velocity distribution can be assumed (e.g. Yahil & Vidal 1977). The 1D analysis of the velocity distribution has been used as the only source of information about the clusters' dynamical state in several studies (see e.g. Hou et al. 2009; Ribeiro, Lopes & Trevisan 2010, 2011; Ribeiro et al. 2013).

Analysing only the velocity distribution, the code identified three velocity modes, suggesting a line-of-sight interaction scenario within a clustercentric radius of 2.6 Mpc. The two best models fitted by *mclust* in one dimension disagree. Model E, that fits Gaussians with equal variance, found three modes, while the variable variance model, V, found only one mode. The differences between the BIC of a unimodal and a trimodal distribution are smaller than 5. Hence, the number of clusters found by the chosen model is not reliable (Kass & Wasserman 1995). The left-hand panel of Fig. 2 shows the variation of the BIC according to the number of components fitted by the two different models. The model with the highest BIC is the one chosen. The right-hand panel shows the distribution of the three velocity subgroups in the plane of the sky. The black crosses mark the position of the BCGs. It is noteworthy that the velocity groups found by the code present low cohesion when plotted in the plane of the sky. For example, mode 1 (red circles) has an isolated subgroup of galaxies closer to the centre of the cluster and another one more on the outskirts. This means that we cannot trust the substructure analysis using the velocity distribution only.

3.1.2 2D analysis

Another way to search for substructures in the optical is to perform an analysis of the distribution of galaxies in the plane of the sky looking for overdensities that may be associated with candidate clusters for interaction.

Our 2D analysis found two groups in the clustercentric radius of 2.6 Mpc. In Fig. 3, the left-hand panel shows the variation of the different BICs. The model with the highest BIC is the VII (spherical Gaussians with variable volume and equal shape) that found two
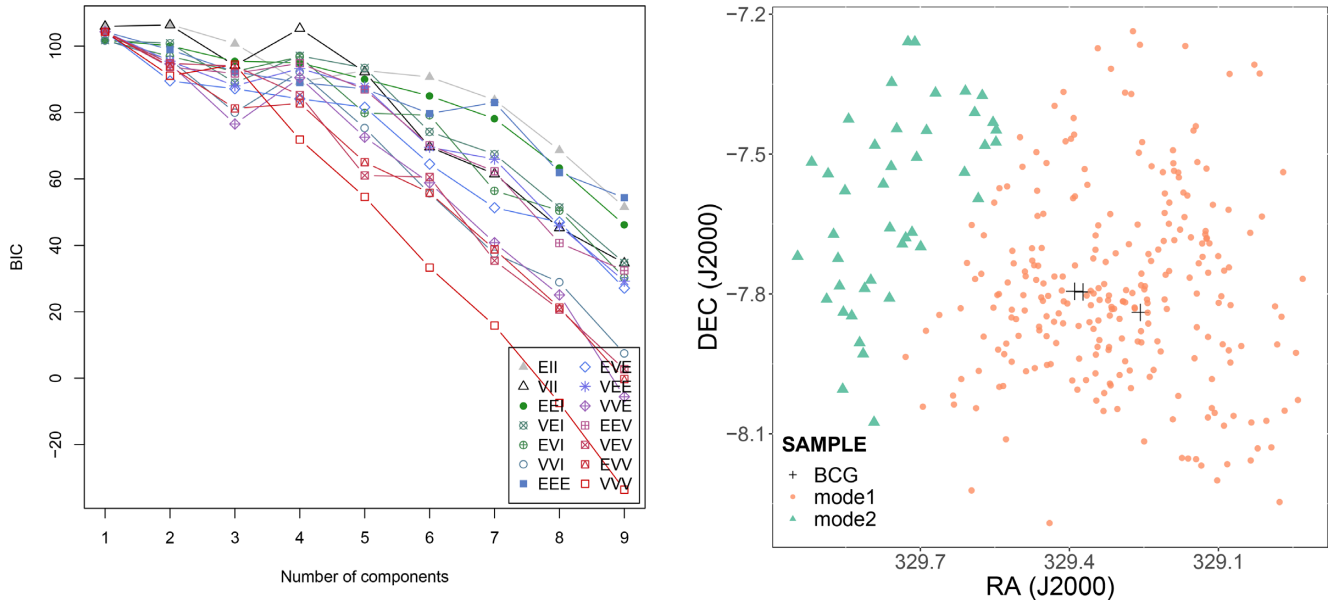
Figure 3. 2D analysis. Left-hand panel: BIC analysis for the different models fitted by *mclust*. The two-component model VII was the one that better fitted the data. Right-hand panel: Two subgroups found by the fitting of 2D Gaussians by the GMM. The black crosses mark the positions of the BCGs.
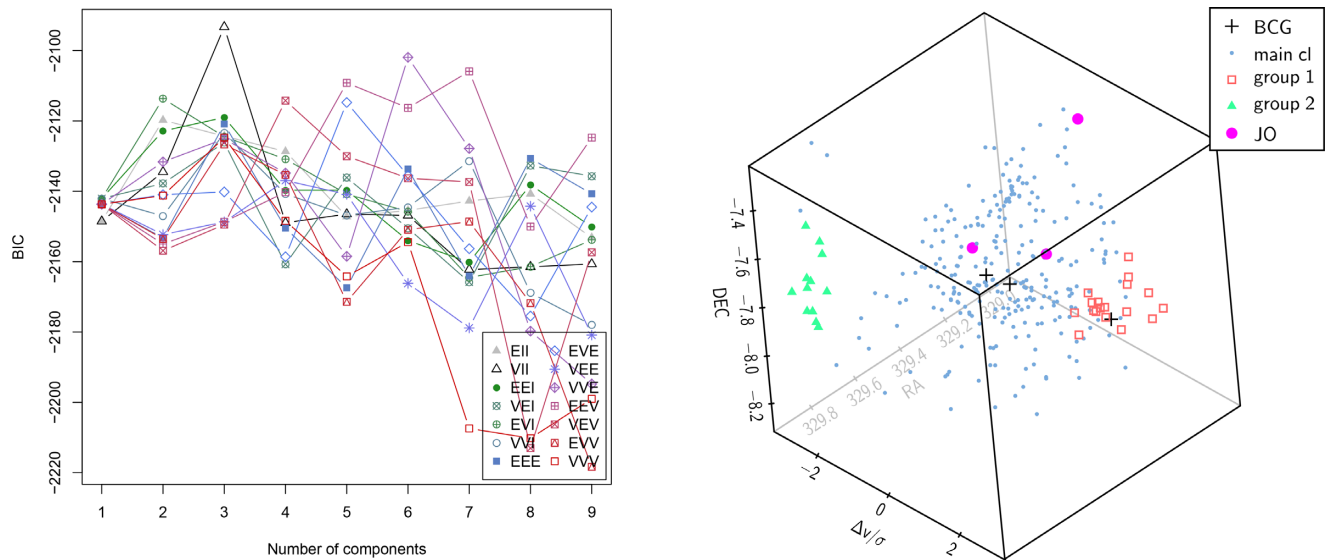

Figure 4. 3D analysis. Left-hand panel: BIC analysis for the different models fitted by *mclust*. The three-component model VII is the one to better fit the data. Right-hand panel: Distribution of galaxy groups found by *mclust* in 3D. The axes were scaled in order to improve the fitting. Jellyfish candidates are from Poggianti et al. (2016) in A2399. BCGs are shown as pink filled circles and black crosses, respectively.

subgroups in the sample. The right-hand panel shows the two groups encountered by the fitting of the 2D Gaussians in the plane of the sky. The black crosses mark the position of the BCGs. However, several models fitted by *mclust* have close BICs, which makes it difficult to obtain a reliable estimate of the number of subgroups in the sample.

3.1.3 3D analysis

To study 3D substructures, one can search for overdensities in the galaxy distribution using the two coordinates of the plane of the sky (RA and Dec.) along with the velocity in the line of sight as a proxy for a third spatial coordinate.

Our 3D analysis found three groups within the 2.6-Mpc clustercentric radius. In Fig. 4, the left-hand panel shows the BICs of the different models fitted. The three-component model VII was the one selected again. The BIC difference from this three-component model and the second larger BIC is greater than 10. This fitting is, therefore, more reliable than the previous cases (Kass & Wasserman 1995). The right-hand panel shows the distribution of the three groups found by *mclust* in 3D. In the 3D analysis, *mclust* retrieved the main cluster (blue dots) of A2399 and two groups in the infalling region (group 1 shown as open red squares and group 2 as filled green triangles). BCGs are shown as the black crosses. The pink filled circles indicate the positions of galaxies under ongoing gas stripping, the so-called jellyfish candidates from Poggianti et al. (2016), which
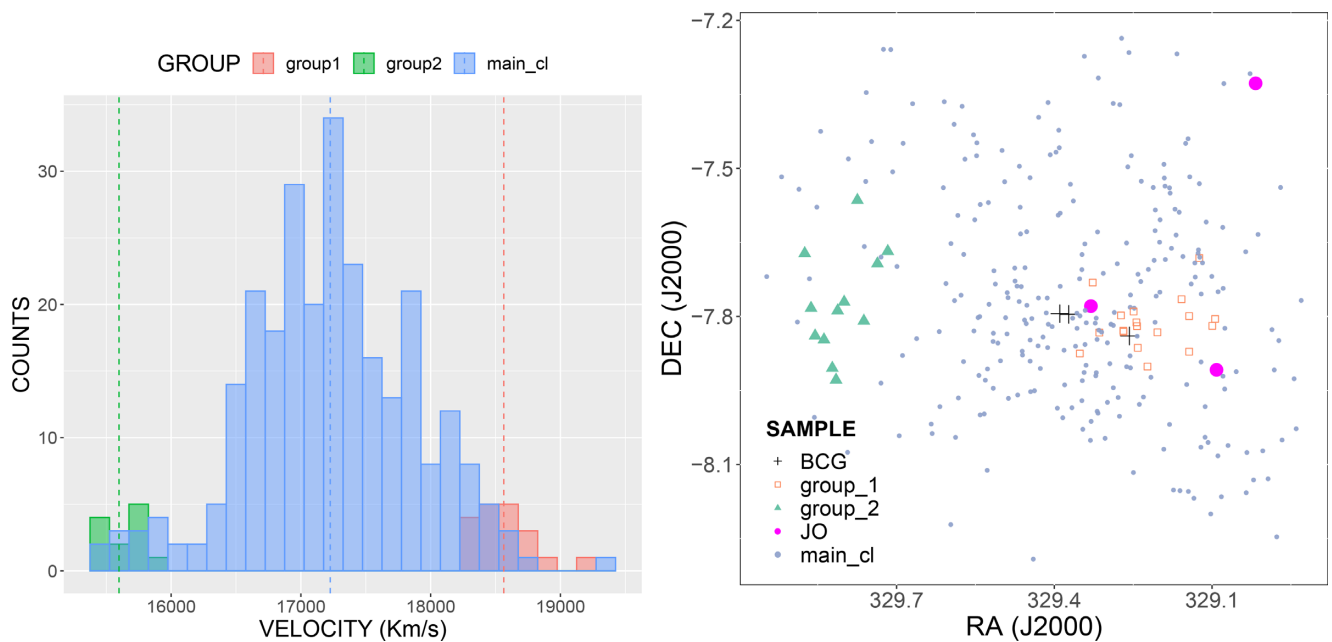
Figure 5. Left-hand panel: Velocity histogram of our sample highlighting the groups found in the 3D optical analysis with different colours. The traced lines indicate the average velocities of each group. Right-hand panel: Plane of the sky distribution of the 3D groups found by *mclust*. The filled pink circles indicate the position of three jellyfish galaxies from Poggianti et al. (2016) in A2399. BCGs are shown as black crosses.

will be discussed later in Section 3.3.1. The axes were scaled in order to improve the fitting since velocity is given in values that are much bigger than RA and Dec. The left-hand panel of Fig. 5 shows the velocity histogram highlighting the classification found by *mclust* in 3D. The right-hand panel shows the same groups distributed in the plane of the sky with the same colours. In Fig. 6, the same groups are plotted in the PS diagram represented by the same colours as before. In the PS diagram, it is clear that the green group is separated from the main body of the cluster. The red group, on the other hand, is more difficult to identify in the PS diagram by eye. However, further on, we will see that this group seems to be associated with a clump in the X-ray emission.

In summary, the 1D and 2D analyses found probable interactions in the line of sight and plane of the sky, respectively, but the BIC gaps between the most and the second most probable models were smaller than 5; therefore, their results were not trustful. Fortunately, in 3D, the best model found the main cluster and two extra groups. In this case, the difference between the first and second highest BICs was higher than 10, so the result was trustful. For this reason, from now on we will just work on further investigation of the substructures found by the 3D analysis.

3.2 Physical parameters of the optical substructures

The *mclust* 3D analysis described above provided the centroids and redshift of each structure identified (i.e. the main cluster and two groups). We then used the *shifting gapper* technique to select only the members of each group. The procedure used to estimate the velocity dispersion σ_P, characteristic radii (R_{500} and R_{200}), and masses (M_{500} and M_{200}) was the same as Lopes et al. (2009a). First, we calculate the robust velocity dispersion for each group using the biweight estimator (Beers, Flynn & Gebhardt 1990). Next, we derive the projected virial radius and then the virial mass is calculated following Girardi et al. (1998). R_{500}, R_{200}, M_{500}, and M_{200} were estimated assuming

an Navarro-Frenk-White (NFW) profile (Navarro, Frenk & White 1997) as in Lopes et al. (2009a).

Table 1 lists the main properties we obtained for the parent cluster and its subgroups within ∼2.6 Mpc. The first column identifies each analysed group. The second, third, and fourth columns contain the centroids of the coordinates (RA, Dec.) and redshift (z_{spec}), respectively. The fifth and sixth columns list the number of galaxies with spectroscopy in each subgroup, N_{spec}, and the number of members selected by the *shifting gapper*, N_{memb}. Columns 7–10 list the characteristic radii and masses of the subgroups M_{200}, R_{200}, M_{500}, and R_{500}. The last column lists the velocity dispersion. The listed values indicate that the main cluster is more than 10 times more massive than the two smaller subgroups.

3.3 Search for interaction in the X-rays

3.3.1 Gas morphology and BCGs

Another way to investigate the dynamical state of galaxy clusters is to inspect how spherically symmetric the distribution of the intracluster gas is.

Fig. 7 shows that the X-ray emission for A2399 (in green) is elongated to the southwest-northeast direction, suggesting an interaction scenario in the plane of the sky, such as in the optical 3D analyses. The X-ray emission has a main concentration and a clump located in the west. The radio intensity maps of the Very Large Array public data are shown as a pink emission. The SDSS coloured image is overlaid on the radio and X-ray ones. Near the centre are two X-ray emission peaks separated by 60 arcsec (∼68.18 kpc) on the plane of the sky. The central BCG, SDSS J215729.42-074744.5, is at the same location as the first X-ray peak and does not emit in radio frequency. The other BCG at the centre, RS1 radio source, is the radio galaxy SDSS J215733.47-074739.2, which coincides with the second X-ray peak in the central region. In the southern part of the X-ray western clump, we can see a third BCG that is also a radio source, RS2.
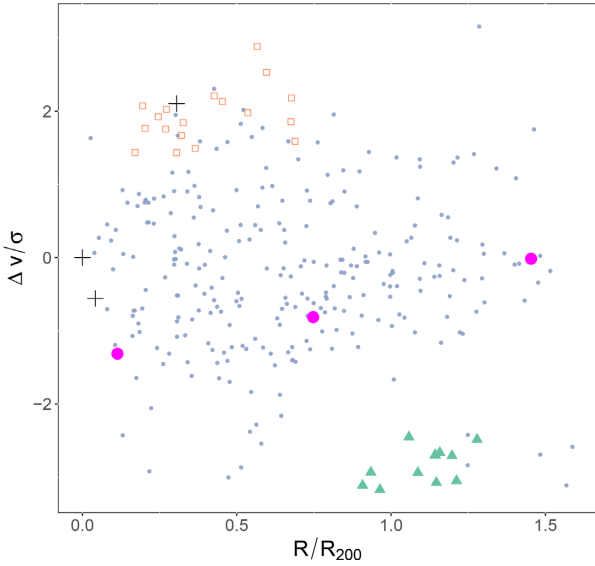
Figure 6. The phase space diagram of A2399 indicating the position of the three jellyfish galaxies from Poggianti et al. (2016) (filled pink circles) and the groups found in the 3D analysis by *mclust*. The green filled triangles represent group 2, easily distinguishable in the phase space diagram, while the red open squares (group 1) are very entangled to the main cluster. BCGs are shown as black crosses.

This galaxy is SDSS J215701.72-075022.5. The bright X-ray source in the lower right corner corresponds to a foreground star. We also highlight JO69, a jellyfish candidate from Poggianti et al. (2016) at RA $=329.\overset{\r{a}}{}36517$, and Dec. $=-7.\overset{\r{a}}{}80358$, with a cyan open circle. The other two jellyfish candidates in this cluster fall outside of the X-ray emission. In the next subsection, we will discuss the location of this jellyfish with respect to the spectral maps.

3.3.2 Structure of thermodynamical maps

Spectral maps are a powerful tool for understanding the dynamical state of a galaxy cluster. Interactions between clusters can leave different imprints in the gas, such as discontinuities and asymmetries in the temperature, pressure, and entropy maps.

Fig. 8 shows the thermodynamic (temperature, pressure, and entropy) and metallicity maps derived from the *XMM–Newton* observation. Overall, the maps have a complex structure. There is a western clump with lower temperature ($\sim$ 2–3 keV), pressure, and entropy, although higher metallicity. Also, we can notice a region extended in the north–south direction, to the left of the western clump, with higher temperature (about 6 keV) and spatially coincident with a high-entropy region. In the pressure map, this region has also higher pressure values, although the region is more elongated, suggestive of a shock region. Near this discontinuity, at the northeast, is the jellyfish candidate JO69, shown as a cyan open circle, which probably had its

morphology modified by the enhancement of the winds near to this shock region. In Owers et al. (2012) and in Vijayaraghavan & Ricker (2013), the authors found a correlation between the ram pressure efficiency and the existence of shocks in the ICM. In fact, two more jellyfish candidates (JO68 and JO70) from Poggianti et al. (2016) were identified in A2399 at larger clustercentric radii.

In recent work, Mitsuishi et al. (2018) have found that A2399 has very high entropy in the centre for a low-surface brightness cluster. The authors concluded that the active galactic nucleus, near to the centre, cannot be responsible for the high entropy in the central region of this cluster. The authors proposed that a merger event caused such an increase. In our pseudo-entropy map, we have regions of higher entropy out of the core that are suggestive of merging as well.

3.3.3 Optical and X-ray comparison

In Fig. 9, we show the main cluster and group 1 found in the 3D analysis of the *mclust* as blue filled circles and red open squares, respectively. The optical groups are plotted on the top of the merged images of the *XMM–Newton* EPIC cameras. Group 2 falls outside the X-ray image. The R_{500} of each group is shown as a dashed circle in the same colour as the corresponding groups. The X-ray emission is highlighted by the black contours and the jellyfish JO69 is shown as a pink open circle. The image suggests that group 1 is associated with the western extension of the X-ray emission. Analysing this image together with the spectral maps, we suspected that A2399 is a bullet cluster-like, where group 1 crossed the main cluster generating the characteristics observed in the thermodynamic maps. To test this hypothesis, we performed a series of hydrodynamical simulations that will be shown in the following section.

It is worth mentioning that *mclust* was successful in finding groups associated with the emission of the main cluster and the western clump in X-ray. If this trend is confirmed in a larger sample, the identification of bullet-like clusters can be automated, increasing the sample of known clusters of this type. This would be of fundamental importance in restricting dark matter parameters such as its cross-section (Markevitch et al. 2002).

4 HYDRODYNAMICAL SIMULATION

In order to evaluate whether A2399 could be the result of a recent collision, we performed a set of hydrodynamical *N*-body simulations. Here, we report on one model that reproduced some of the observed features. The scenario we wish to evaluate is that the clump in the west may have passed through (or near) the core of the main cluster. The main observational constraints the simulation aims to fulfil are the temperature structure (i.e. the presumed shock front with 5.2–6.3 keV) and the current separation between the centroids of the two objects. We assume that the separation between the simulated dark matter peaks in the simulation is a proxy for the

Table 1. Main properties of A2399 and its subgroups.

| Cluster	RA (J2000)	Dec. (J2000)	z_{spec}	N_{spec}	N_{memb}	M_{200} $(10^{14}\,\text{M}_\odot)$	R_{200} (Mpc)	M_{500} $(10^{14}\,\text{M}_\odot)$	R_{500} (Mpc)	σ_p (km s^{-1})
Parent	329.37258	−7.79569	0.0579	299	275	$5.25^{+0.45}_{-0.39}$	1.63	$4.84^{+0.41}_{-0.36}$	1.17	$621.2^{+26.4}_{-23.1}$
Main cl	329.34457	−7.78231	0.0575	268	256	$4.25^{+0.40}_{-0.32}$	1.52	$3.95^{+0.37}_{-0.30}$	1.10	$552.4^{+25.9}_{-21.0}$
Group 1	329.24268	−7.81965	0.0619	19	18	$0.22^{+0.13}_{-0.07}$	0.56	$0.20^{+0.12}_{-0.06}$	0.41	$216.2^{+62.1}_{-32.7}$
Group 2	329.81361	−7.78557	0.0521	12	12	$0.12^{+0.06}_{-0.03}$	0.47	$0.12^{+0.06}_{-0.03}$	0.34	$146.3^{+36.8}_{-13.6}$
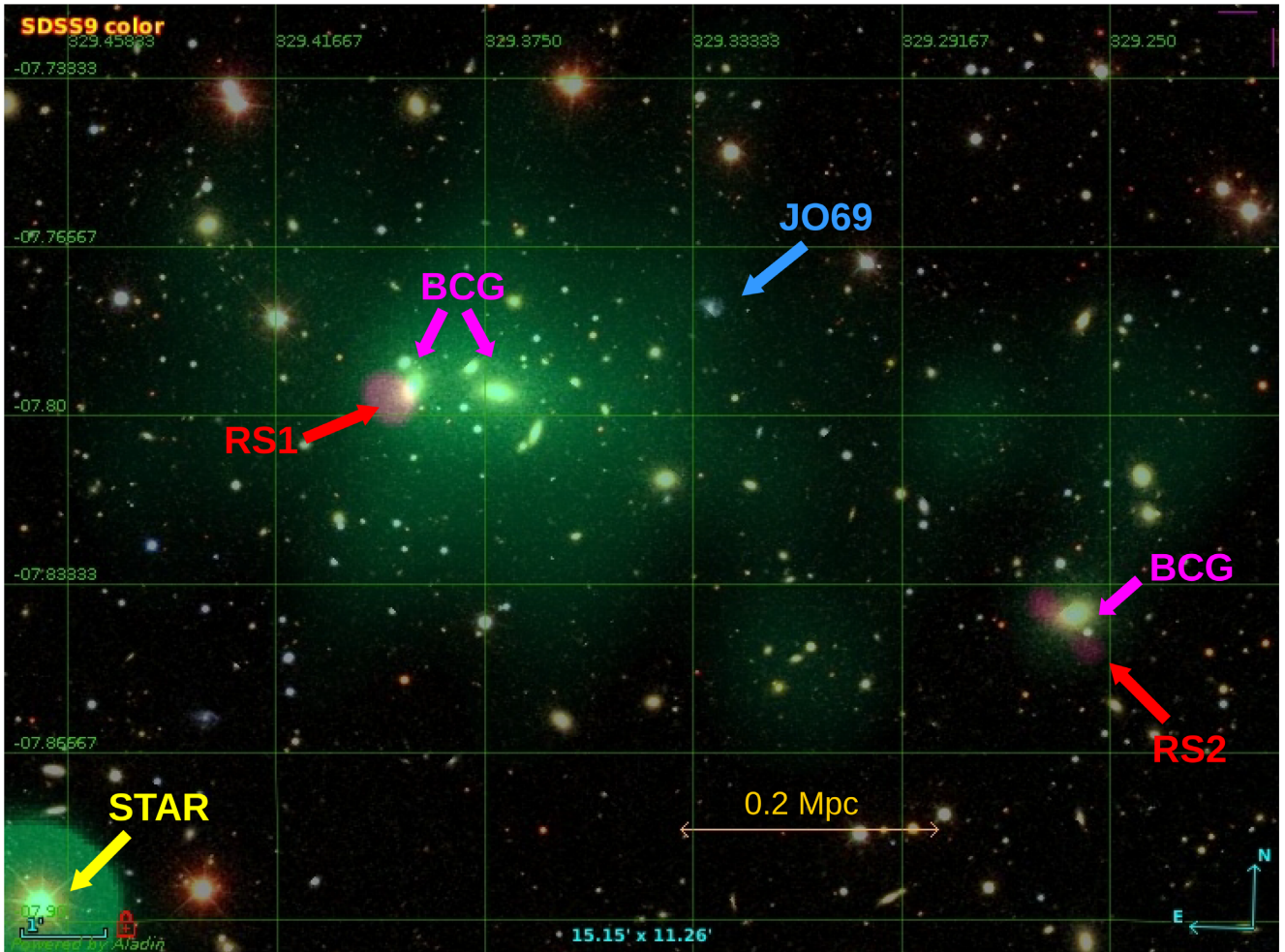
Figure 7. X-ray emission in green and radio emission in red overlaid on the SDSS optical image of A2399. Three BCGs are pointed out by pink arrows and one jellyfish galaxy pointed by a cyan arrow, JO69. This image corresponds to a very central region of A2399 within a radius of 0.5 Mpc.

observed separation between the BCG of the main cluster and the BCG of the western clump.

The simulation setup consists of two idealized galaxy clusters that are initially spherical and relaxed. They are set to collide with a given impact parameter and a given initial velocity. Each galaxy cluster is modelled by gas particles and dark matter particles. In the main cluster, the gas and the dark matter follow a Hernquist (1990) profile. The parameters were chosen such that its virial mass is $M_{200} = 4.0 \times 10^{14} \, M_{\odot}$ and its virial radius is $R_{200} = 1.5$ Mpc. The choice of density profiles fixes the gas temperature profile through the imposition of hydrostatic equilibrium. The temperature around $R \sim 100$ kpc is ~ 3.8 keV but there is a small cool-core. In the western clump, a Hernquist (1990) profile was used for the dark matter and a Dehnen (1993) profile was adopted for the gas. The choice of parameters was such as to give $M_{200} = 0.2 \times 10^{14} \, M_{\odot}$ and $R_{200} = 0.6$ Mpc, which results in a central gas temperature of ~ 2 keV. The initial condition parameters were chosen to be close to the observational estimates from the optical and X-ray. The numerical realizations of these initial conditions were generated with the techniques described in Machado & Lima Neto (2013). In the model presented here, each cluster has 10^6 dark matter particles and 10^6 gas particles. After the two clusters have been created, the collision is prepared in the following way. The clusters are positioned with an initial separation of 3 Mpc along the x-axis and also with an

impact parameter b along the y-axis. The initial velocity between them is v_0, parallel to the x-axis. This impact parameter b refers to the y offset at $t = 0$; later on, at the moment of pericentric passage, the separation is generally much smaller. We found that the initial velocity of $v_0 = -700$ km s^{-1} produced suitable results. The choice of this approximate value was motivated by the result of the two-body dynamical modelling, which provides an estimate of the total mechanical energy of the system. For simplicity, we assume that the collision takes place on the plane of the sky.

The simulations were performed with the hydrodynamical *N*-body code GADGET-2 (Springel 2005). The evolution of the system was followed for 5 Gyr but the relevant phases take place within the first 3 Gyr. Cosmological expansion is ignored due to the relatively small spatial extent of the problem.

One desirable feature of the simulated temperature map is that the presumed shock front (blue region in the temperature map in Fig. 8) needs to be roughly perpendicular to the line connecting the BCGs. This configuration would arise naturally in the case of a head-on encounter, in which the velocity of the clump is entirely radial towards the main cluster and thus the shock front is perpendicular to it. However, in such a perfectly frontal collision ($b = 0$), the outcome of the simulation would be symmetrical to a degree that is not warranted by the complex observed morphologies. At the other extreme, if the impact parameter is too large, the pericentric
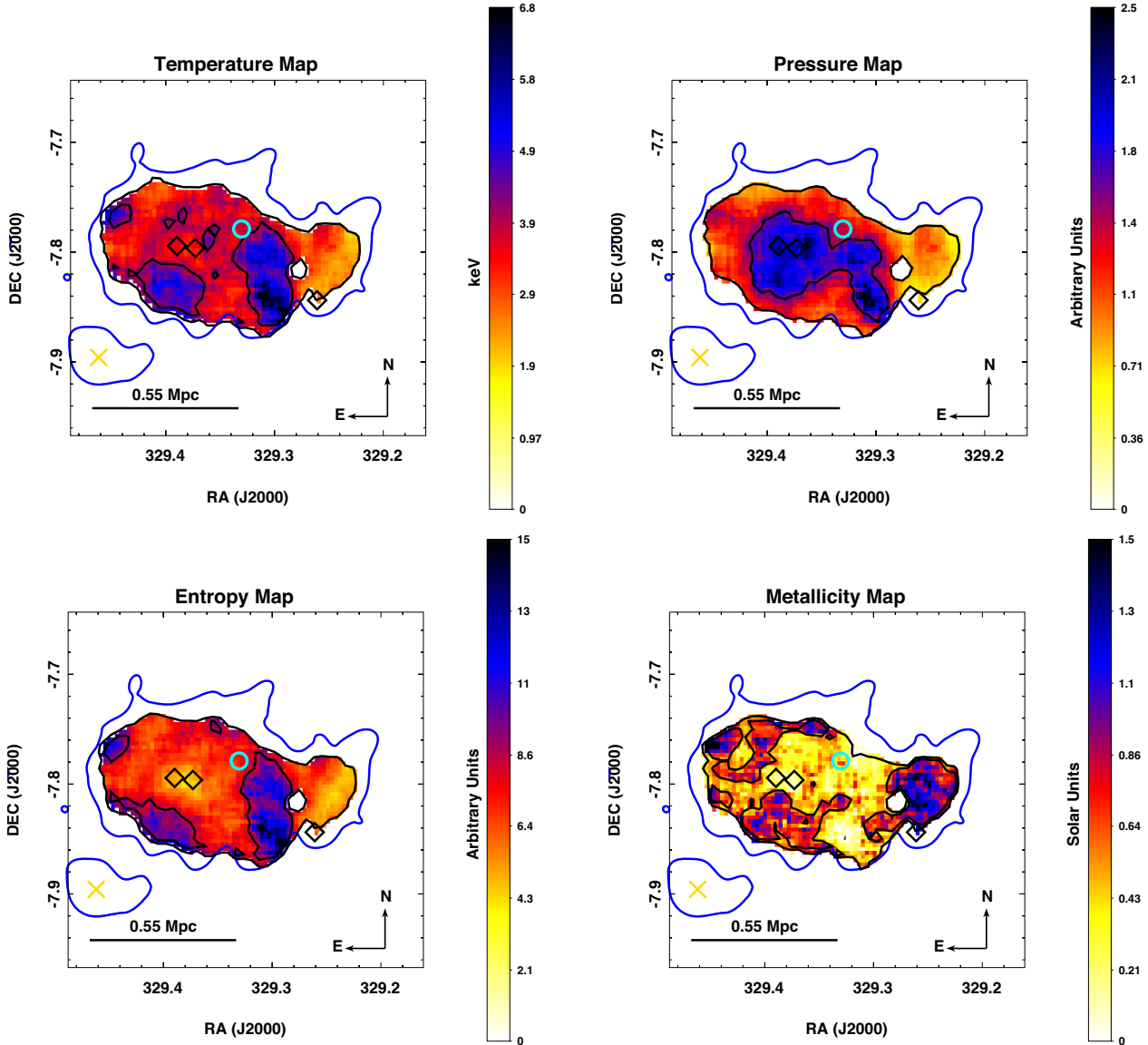
Figure 8. X-ray spectral maps. The open circles in cyan show the position of JO69, a jellyfish galaxy. The open black diamonds indicate where the BCGs are located, and the yellow crux marks the position of a foreground star. The blue contour shows the X-ray emission from *XMM–Newton*. All the maps show that A2399 is a disturbed cluster with a western clump that resembles a bullet.

distance may be considerable. In this regime of large *b*, at the moment when the western clump overtakes the main cluster, its velocity is mostly perpendicular to the line connecting them. Therefore, the shock front would be approximately parallel to the line connecting the two clusters, which is not the desired result. Therefore, a suitable compromise must lie in a choice of *b* that leads to a non-zero but small pericentric separation. By means of a limited but plausible exploration of the parameter space, we found that an initial impact parameter of *b* = 500 kpc gives satisfactory results. At the moment of pericentric passage, the separation between the centroids of the clusters is 110 kpc. This may seem small, but it is known that even such modest pericentric distances are capable of bringing about substantial departures in comparison to perfectly frontal collisions (e.g. Machado & Lima Neto 2015; Doubrawa et al. 2020). Choices of *b* that led to pericentric distances greater than about 200 kpc gave poorer results.

In our best simulation model, the pericentric passage takes place at *t* = 2.35 Gyr. Approximately 0.24 Gyr later, a configuration is reached when the simulated temperature map best reproduces the main features of the observed temperature map. Simultaneously, the separation between the dark matter peaks reaches 580 kpc – within an error of 5 per cent of the observed separation of 550 kpc. Therefore, we regard this instant (*t* = 2.59 Gyr) as the best matching instant of the simulation. The density map and the emission-weighted temperature map of the best instant are shown in Fig. 10. These projected maps were rotated by an angle on the *xy* plane merely to match the position angle of the observational images, making the visual comparison more straightforward. There is no inclination between the orbital plane and the plane of the sky.

The temperature structure of the simulated map in Fig. 10 should be compared to the observed temperature map in Fig. 8. Notice that the region for which observational data are available is restricted, in
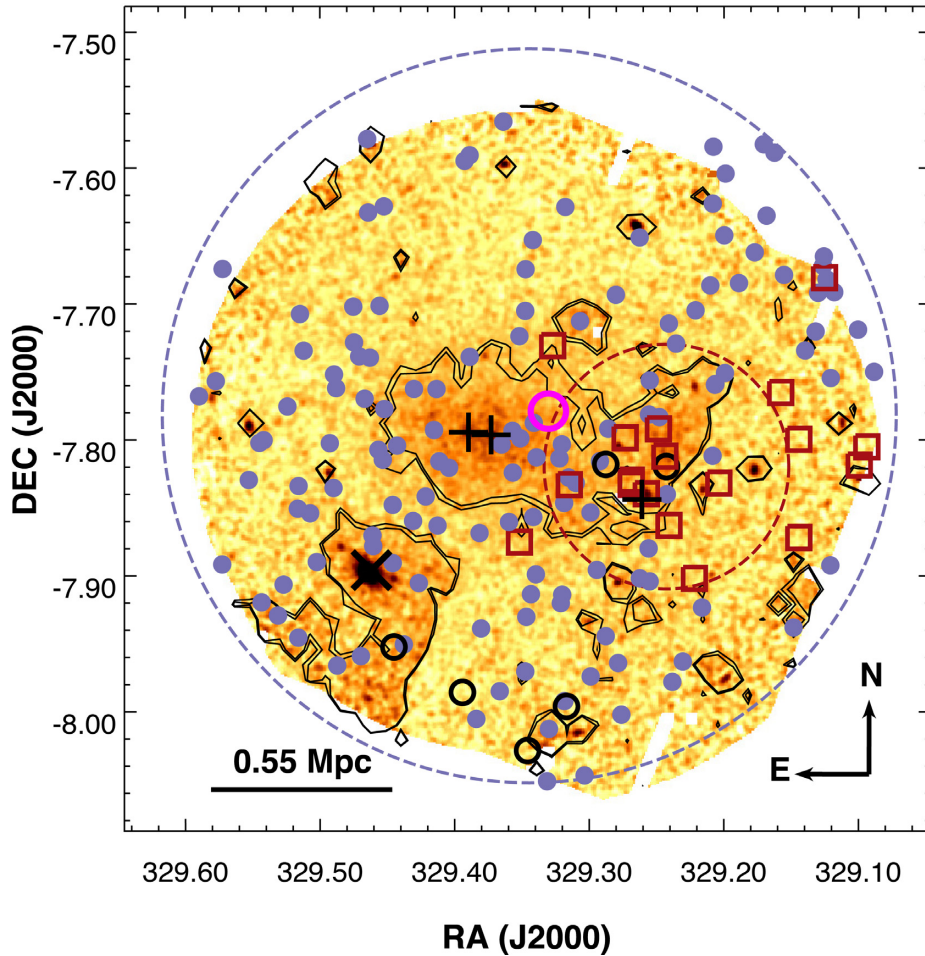
Figure 9. Mclust groups in the central region of A2399 overlaid on the ICM. The filled blue circles represent the main cluster and the open red squares the group 1. Black contours highlight the X-ray emission and black crosses show the positions of the BCGs. Interlopers are marked by black open circles. The dashed blue and red circles show R_{500} of the main cluster and group 1, respectively. The open magenta circle shows JO69.

comparison to the entire simulated frame. In order to produce a fair comparison, a semi-transparent mask was applied to the simulated temperature map in Fig. 10, outlining a region with the same size and shape as the region where observational data are available. One should also bear in mind that this type of simulation is highly idealized. Nevertheless, this simple simulation succeeds in capturing some of the overall morphological features of the temperature map. There is a hotter region of ∼6 keV, elongated roughly along the north–south direction and which is flanked on both sides by cooler gas; the gas on the side of the main cluster has on average ∼4 keV, and that the gas on the side of the western clump has ∼2–3 keV.

One of the major shortcomings of idealized binary collisions is the lack of cosmological context and therefore of realistic substructures taking part in the previous history of the clusters. The assumption that the two clusters were initially in hydrostatic equilibrium is adopted for simplicity and also due to lack of prior knowledge of the state of the system. In particular, A2399 could have undergone an earlier collision with the eastern substructure (group 2), perhaps pre-mixing the metals as it passed through the core. This is not modelled by our current simulation. Including more than two structures is not often attempted when modelling individual clusters, because it greatly increases the possible combinations of parameters; a few instances exist (e.g. Brüggen, van Weeren & Röttgering 2012; Ruggiero et al.

2019; Doubrawa et al. 2020), but they remain rare. Furthermore, in the particular case of A2399, the problem would be even more underconstrained due to the lack of X-ray data at the position of the eastern structure. For these reasons, it would be unreasonable to demand that such simplified simulations reproduce the detailed small-scale structure of a given cluster. Nevertheless, it is interesting to note that the simulation exhibits a plume of hot gas behind the shock front, in a position which is comparable to a similar region of hot gas in the observations. While we cannot claim that they correspond to the same physical process, we can attempt to describe how that feature arose in the simulation. As the shock front passes close to the main cluster core, the nose of the shock is able to advance, but the gas in the western edge of the shock front encounters the denser part of the main cluster core and its motion is hindered, leaving behind a detached wisp of hot gas that will later dissipate. This can be seen in the time evolution of the temperature maps shown in Fig. 11.

While the simulation succeeds in recovering some global properties of the cluster, it does not reproduce all morphological features in great detail. For example, a more exhaustive exploration of the parameter space of initial conditions might have resulted in an improved model, in which the western clump would have retained more of its gas in the final configuration. Such fine-tuning is beyond
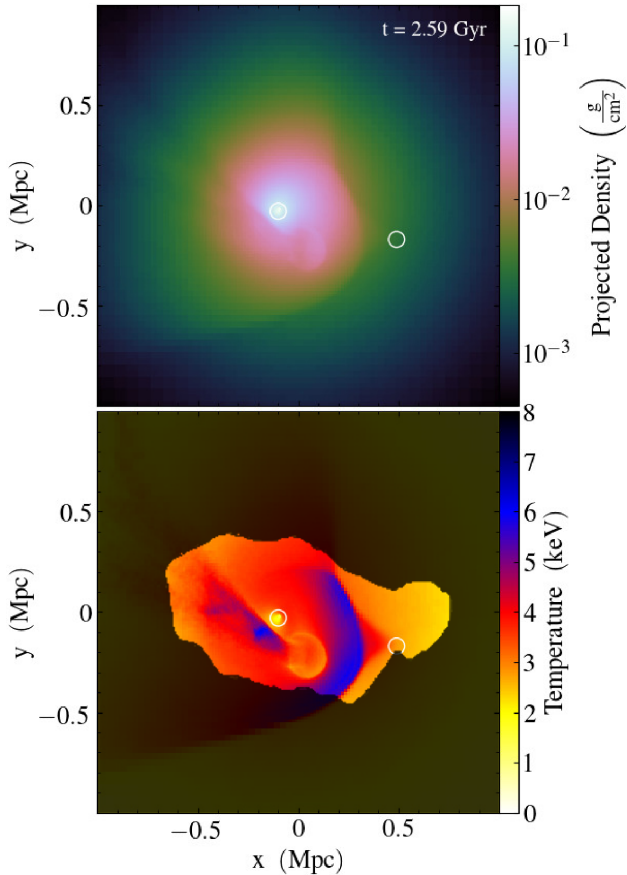
Figure 10. The best matching instant of the simulation. Top: projected density map. Bottom: emission-weighted projected temperature map; the overlaid mask outlines a region comparable to the region where observational data are available (Fig. 8). The white circles mark the positions of the dark matter peaks of each cluster.

the scope of this work. By offering one possible solution, we argue that this type of collision is a plausible scenario to explain the current state of A2399.

5 DISCUSSION: ON THE IMPLICATION OF THE METALLICITY DISTRIBUTION

The ICM is not only composed by primordial elements such as hydrogen and helium but also enriched with heavier elements such as iron, which indicates that material processed inside stars is related to this enrichment. There are several mechanisms that can contribute to driving metals from galaxies into the ICM (see Schindler & Diaferio 2008, for a review).

Although a strong enhancement in the abundance is found in the central regions of the cool-core clusters (De Grandi & Molendi 2001), Laganá et al. (2019) showed that several metallicity peaks are observed also out of the core in the metallicity maps of many clusters. Such peaks may be related to galaxies that recently introduced metals into the ICM by galactic winds or ram-pressure stripping (RPS) but still did not have time to mix with the rest of the gas (Kapferer et al. 2007).

A plausible hypothesis for explaining the excess of metals in the western clump, which may also account for this 'lack' of metals in the centre of the main cluster, is that a possible frontal collision of the western clump with the main cluster enhanced the RPS in that region

causing the galaxies of group 1 to insert more gas and metals in there. Some hydrodynamical simulations support the idea that collisions increase the RPS (Owers et al. 2012; Vijayaraghavan & Ricker 2013). In fact, in clusters at low redshift like A2399 where the ICM is already formed and the galaxies have already had their star formation peaks around redshift one (Madau, Pozzetti & Dickinson 1998), ram pressure is the most important mechanism of gas enrichment (Schindler & Diaferio 2008). To test this hypothesis, we would need to obtain from the simulations the increase in RPS in the galaxies of the western clump. This will be the subject of future studies.

Usually, metallicity and temperature are anticorrelated (De Grandi & Molendi 2001). However, we can observe a region in the southeast and another small region in the northeast where temperatures are higher and metallicities too. Such as in the case presented in Durret, Lima Neto & Forman (2005) for A85, the higher metallicities in those regions cannot be explained by the anticorrelation between metallicity and temperature; they are very likely mergers features. In summary, our findings are consistent with the hypothesis that this cluster is a postmerger, but more detailed studies with simulations are required to verify if the RPS enhancement over the galaxies of group 1 was responsible for the observed metallicity excess in the western clump.

6 SUMMARY AND CONCLUSIONS

Identifying interactions accurately in optical and X-ray provides us with clues about the merger history of galaxy clusters that help to better understand the influence of extremely violent environments on the evolution of galaxies. A large number of contradictory conclusions about the dynamical state of Abell 2399 was shown in our literature review (Flin & Krywult 2006; Ramella et al. 2007; Böhringer et al. 2010; Fogarty et al. 2014; Moretti et al. 2017; Owers et al. 2017; Mitsuishi et al. 2018). In this work, we analysed the substructures at different wavelengths and used the results as an input for a hydrodynamical *N*-body simulation. First, we isolated the substructures in the optical data using a GMM code and then we matched them with the X-ray substructures through the comparison with the surface brightness and spectral maps. In the optical analysis, when running *mclust* in 1D, 2D, and 3D, we found that the 3D analysis yielded the most robust results retrieving three separate structures: the main cluster, and two groups. After having isolated those substructures from the optical, we obtained new measures for characteristic radius, mass, and the velocity dispersion of the main cluster and the other two optical groups. In the X-ray analysis, we found the main gas concentration associated with the main cluster and a western clump associated with an optical substructure. Both wavelengths, optical and X-ray, indicated that A2399 is unrelaxed.

The optical group 1, the one that matches the western X-ray clump, has lower pressure, entropy, and temperature than the main cluster but higher metallicity. We also found that the jellyfish candidate JO69 from Poggianti et al. (2016) is near a shock region of the gas. A possible scenario to explain these characteristics is that this subgroup crossed the main cluster, increasing the RPS (that may have influenced on the formation of the jellyfish) in that region and also on the release of metals from the galaxies of that subgroup in the western clump, explaining its higher metallicity.

Using hydrodynamical *N*-body simulations, we offered one specific model that reproduces some of the observed features of A2399 to an approximate degree. The results of the simulation suggest that A2399 could be the outcome of a recent collision taking place on the plane of the sky. This is consistent with the hypothesis put forth by Mitsuishi et al. (2018), but in the context of our simulation, the
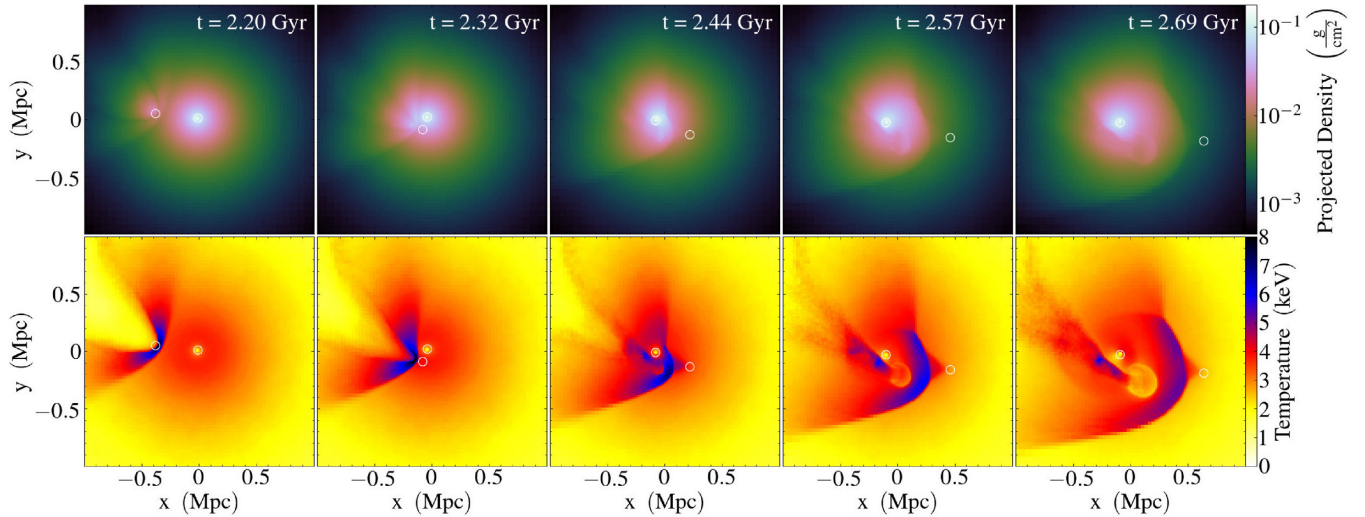
Figure 11. Time evolution of the simulation. Upper panels show projected density maps and lower panels show emission-weighted projected temperature maps. The white circles mark the positions of the dark matter peaks of each cluster.

temperature discontinuity is understood as a shock front. In our best simulation model, the western clump passed through the main cluster 0.24 Gyr ago, with a pericentric distance of 110 kpc. It is currently 580 kpc ahead of the main cluster core, in an outgoing configuration, and has left behind a ∼6 keV shock front and also a detached strand of hot gas. The dark matter peak of the western clump has been mostly dissociated from its gas. We propose that this is a plausible scenario to explain the current state of A2399. However, this reconstruction is not unique, even if it is physically well motivated. Alternative solutions and even an incoming configuration cannot yet be confidently ruled out. Gravitational weak lensing analyses would be needed to map out the dark matter distribution and thus provide more robust constraints for future dedicated simulations.

DATA AVAILABILITY

The data underlying this article were derived from sources in the public domain and are available in Vizier, at https://dx.doi.org/10.1051/0004-6361:200810997 and https://dx.doi.org/10.1051/0004-6361/201630030.

REFERENCES

Alam S. et al., 2015, ApJS, 219, 12
Andrade-Santos F. et al., 2017, ApJ, 843, 76
Arnaud K. A., 1996, in Jacoby G. H., Barnes J., eds, ASP Conf. Ser. Vol. 101, Astronomical Data Analysis Software and Systems V. Astron. Soc. Pac., San Francisco, p. 17
Asplund M., Grevesse N., Sauval A. J., Scott P., 2009, ARA&A, 47, 481
Bautz L. P., Morgan W. W., 1970, ApJ, 162, L149
Beers T. C., Flynn K., Gebhardt K., 1990, AJ, 100, 32
Biviano A. et al., 2017, A&A, 607, A81
Boch T., Fernique P., 2014, in Manset N., Forshay P., eds, ASP. Conf. Ser. Vol. 485, Astronomical Data Analysis Software and Systems. Astron. Soc. Pac., San Francisco, p. 277
Böhringer H. et al., 2010, A&A, 514, A32
Bonnarel F. et al., 2000, A&AS, 143, 33
Bregman J. N., Dupke R. A., Miller E. D., 2004, ApJ, 614, 31

Brüggen M., van Weeren R. J., Röttgering H. J. A., 2012, MNRAS, 425, L76
Cava A. et al., 2009, A&A, 495, 707
Cohen S. A., Hickox R. C., Wegner G. A., Einasto M., Vennik J., 2014, ApJ, 783, 136
De Grandi S., Molendi S., 2001, ApJ, 551, 153
Dehnen W., 1993, MNRAS, 265, 250
Diaferio A., 1999, MNRAS, 309, 610
Diaferio A., Geller M. J., 1997, ApJ, 481, 633
Dooley G. A., Griffen B. F., Zukin P., Ji A. P., Vogelsberger M., Hernquist L. E., Frebel A., 2014, ApJ, 786, 50
Doubrawa L., Machado R. E. G., Laganá T. F., Lima Neto G. B., Monteiro-Oliveira R., Cypriano E. S., 2020, MNRAS, 495, 2022
Duda R. O., Hart P. E., 1973, Pattern Classification and Scene Analysis. John Willey & Sons, New Yotk
Durret F., Lima Neto G. B., Forman W., 2005, A&A, 432, 809
Durret F., Laganá T. F., Adami C., Bertin E., 2010, A&A, 517, A94
Durret F., Laganá T. F., Haider M., 2011, A&A, 529, A38
Einasto M. et al., 2010, A&A, 522, A92
Einasto M. et al., 2012a, A&A, 540, A123
Einasto M. et al., 2012b, A&A, 542, A36
Fadda D., Girardi M., Giuricin G., Mardirossian F., Mezzetti M., 1996, ApJ, 473, 670
Fasano G. et al., 2006, A&A, 445, 805
Findley D., 1991, Anna. Inst. Stat. Math., 43, 505
Flin P., Krywult J., 2006, A&A, 450, 9
Fogarty L. M. R. et al., 2014, MNRAS, 443, 485
Fraley C., Raftery A. E., Murphy T. B., Scrucca L., 2012, Technical report, mclust version 4 for r: normal mixture modeling for model-based clustering, classification, and density estimation. Technical report. University of Washington, Washington D.C.
Girardi M., Giuricin G., Mardirossian F., Mezzetti M., Boschin W., 1998, ApJ, 505, 74
Gullieuszik M. et al., 2015, A&A, 581, A41
Haines C. P. et al., 2015, ApJ, 806, 101
Hernquist L., 1990, ApJ, 356, 359
Hou A., Parker L. C., Harris W. E., Wilman D. J., 2009, ApJ, 702, 1199
Hudson D. S., Mittal R., Reiprich T. H., Nulsen P. E. J., Andernach H., Sarazin C. L., 2010, A&A, 513, A37
Hurier G., 2019, preprint (arXiv:1904.06951)
Hurier G., Singh P., Hernández-Monteagudo C., 2019, A&A, 625, L4
Iapichino L., Federrath C., Klessen R. S., 2017, MNRAS, 469, 3641
Ichinohe Y., Werner N., Simionescu A., Allen S. W., Canning R. E. A., Ehlert S., Mernier F., Takahashi T., 2015, MNRAS, 448, 2971
Jaffé Y. L., Smith R., Candlish G. N., Poggianti B. M., Sheen Y.-K., Verheijen M. A. W., 2015, MNRAS, 448, 1715
Jaffé Y. L. et al., 2016, MNRAS, 461, 1202
Kapferer W. et al., 2007, A&A, 466, 813
Kass R. E., Wasserman L., 1995, J. Am. Stat. Assoc., 90, 928
Katayama H., Hayashida K., Takahara F., Fujita Y., 2003, ApJ, 585, 687
Kelkar K. et al., 2020, MNRAS, 496, 442
Laganá T. F., Lima Neto G. B., Andrade-Santos F., Cypriano E. S., 2008, A&A, 485, 633
Laganá T. F., Lovisari L., Martins L., Lanfranchi G. A., Capelato H. V., Schellenberger G., 2015, A&A, 573, A66
Laganá T. F., Durret F., Lopes P. A. A., 2019, MNRAS, 484, 2807
Lage C., Farrar G., 2014, ApJ, 787, 144
Lopes P. A. A., de Carvalho R. R., Gal R. R., Djorgovski S. G., Odewahn S. C., Mahabal A. A., Brunner R. J., 2004, AJ, 128, 1017
Lopes P. A. A., de Carvalho R. R., Kohl-Moreira J. L., Jones C., 2009a, MNRAS, 392, 135
Lopes P. A. A., de Carvalho R. R., Kohl-Moreira J. L., Jones C., 2009b, MNRAS, 399, 2201
Lopes P. A. A., Trevisan M., Laganá T. F., Durret F., Ribeiro A. L. B., Rembold S. B., 2018, MNRAS, 478, 5473
Macario G., Markevitch M., Giacintucci S., Brunetti G., Venturi T., Murray S. S., 2011, ApJ, 728, 82
Machado R. E. G., Lima Neto G. B., 2013, MNRAS, 430, 3249
Machado R. E. G., Lima Neto G. B., 2015, MNRAS, 447, 2915
Machado R. E. G., Monteiro-Oliveira R., Lima Neto G. B., Cypriano E. S., 2015, MNRAS, 451, 3309
Madau P., Pozzetti L., Dickinson M., 1998, ApJ, 498, 106
Markevitch M., Gonzalez A. H., David L., Vikhlinin A., Murray S., Forman W., Jones C., Tucker W., 2002, ApJ, 567, L27
Markevitch M., Govoni F., Brunetti G., Jerius D., 2005, ApJ, 627, 733
Mastropietro C., Burkert A., 2008, MNRAS, 389, 967
McGee S. L., Balogh M. L., Bower R. G., Font A. S., McCarthy I. G., 2009, MNRAS, 400, 937
Mitsuishi I., Babazaki Y., Ota N., Sasaki S., Böhringer H., Chon G., Pratt G. W., 2018, PASJ, 70, 112
Monteiro-Oliveira R., Cypriano E. S., Machado R. E. G., Lima Neto G. B., Ribeiro A. L. B., Sodré L., Dupke R., 2017, MNRAS, 466, 2614
Monteiro-Oliveira R., Doubrawa L., Machado R. E. G., Lima Neto G. B., Castejon M., Cypriano E. S., 2020, MNRAS, 495, 2007
Moretti A. et al., 2017, A&A, 599, A81
Navarro J. F., Frenk C. S., White S. D. M., 1997, ApJ, 490, 493
Oman K. A., Hudson M. J., Behroozi P. S., 2013, MNRAS, 431, 2307
Owers M. S., Randall S. W., Nulsen P. E. J., Couch W. J., David L. P., Kempner J. C., 2011, ApJ, 728, 27
Owers M. S., Couch W. J., Nulsen P. E. J., Rand all S. W., 2012, ApJ, 750, L23
Owers M. S. et al., 2017, MNRAS, 468, 1824
Poggianti B. M. et al., 2016, AJ, 151, 78
Press W. H., Schechter P., 1974, ApJ, 187, 425
Ramella M. et al., 2007, A&A, 470, 39
Read A. M., Ponman T. J., 2003, A&A, 409, 395
Rhee J., Smith R., Choi H., Yi S. K., Jaffé Y., Candlish G., Sánchez-Jánssen R., 2017, ApJ, 843, 128
Ribeiro A. L. B., Lopes P. A. A., Trevisan M., 2010, MNRAS, 409, L124
Ribeiro A. L. B., Lopes P. A. A., Trevisan M., 2011, MNRAS, 413, L81
Ribeiro A. L. B., de Carvalho R. R., Trevisan M., Capelato H. V., La Barbera F., Lopes P. A. A., Schilling A. C., 2013, MNRAS, 434, 784
Roettiger K., Burns J. O., Loken C., 1996, ApJ, 473, 651
Ruggiero R., Machado R. E. G., Roman-Oliveira F. V., Chies-Santos A. L., Lima Neto G. B., Doubrawa L., Rodríguez del Pino B., 2019, MNRAS, 484, 906
Russell H. R., Sanders J. S., Fabian A. C., Baum S. A., Donahue M., Edge A. C., McNamara B. R., O'Dea C. P., 2010, MNRAS, 406, 1721
Russell H. R. et al., 2012, MNRAS, 423, 236
Sanderson A. J. R., Edge A. C., Smith G. P., 2009, MNRAS, 398, 1698
Sarazin C. L., 2002, in Feretti L., Gioia I. M., Giovannini G., eds, Astrophysics and Space Science Library, Vol. 272, Merging Processes in Galaxy Clusters. Kluwer Academic Publishers, Dordrecht, p. 1
Schindler S., Diaferio A., 2008, Space Sci. Rev., 134, 363
Schmidt W., Byrohl C., Engels J. F., Behrens C., Niemeyer J. C., 2017, MNRAS, 470, 142
Scrucca L., Fop M., Murphy T. B., Raftery A. E., 2016, R J., 8, 289
Sharp R. et al., 2006, McLean I. S., Iye M., Vol. 6269, Ground-based and Airborne Instrumentation for Astronomy, SPIE, Bellingham, p. 152
Smith R. K., Brickhouse N. S., Liedahl D. A., Raymond J. C., 2001, ApJ, 556, L91
Smith G. A. et al., 2004, in Moorwood A. F. M., Iye M., eds, Proc. SPIE Conf. Ser., Vol. 5492, AAOmega: a multipurpose fiber-fed spectrograph for the AAT. SPIE, Bellingham. p. 410
Springel V., 2005, MNRAS, 364, 1105
Springel V., Farrar G. R., 2007, MNRAS, 380, 911
Stroe A., Sobral D., Röttgering H. J. A., van Weeren R. J., 2014, MNRAS, 438, 1377
Taylor M. B., 2005, in Shopbell P., Britton M., Ebert R., eds, ASP Conf. Ser., Vol. 347, Astronomical Data Analysis Software and Systems XIV. Astron. Soc. Pac., San Francisco, p. 29
van Weeren R. J., Brüggen M., Röttgering H. J. A., Hoeft M., 2011, MNRAS, 418, 230
Vazza F., Roediger E., Brüggen M., 2012, A&A, 544, A103
Ventimiglia D. A., Voit G. M., Donahue M., Ameglio S., 2008, ApJ, 685, 118
Vijayaraghavan R., Ricker P. M., 2013, MNRAS, 435, 2713
Yahil A., Vidal N. V., 1977, ApJ, 214, 347

ZuHone J. A., 2011, ApJ, 728, 54
ZuHone J. A., Markevitch M., Johnson R. E., 2010, ApJ, 717, 908
ZuHone J. A., Markevitch M., Brunetti G., Giacintucci S., 2013, ApJ, 762, 78

APPENDIX A: GOODNESS OF FITTING

A probability distribution that consists of multiple Gaussians depends on a covariance matrix Σ_k, which can be understood as the variance in multiple dimensions. The eigen-decomposition of the form $\Sigma_k = \lambda_k D_k A_k D_k^\top$ provides parsimonious parametrizations of the covariance matrices for each model fitted by *mclust*. The volume of the ellipsoid described by Σ_k is regulated by the scalar λ_k, the shape of the isodensity contours is outlined by the diagonal matrix

A_k, and the ellipsoid orientation is determined by the orthogonal matrix D_k of the kth component.

The parameters of the covariance matrices defining each different model fitted by *mclust* are described in Table A1. Each letter in the model identifier column encodes a different meaning according to its position. The first letter gives information about the volume, the second describes the shape, and the third specifies the orientation of the multidimensional Gaussians. The letters used for describing those features are E (equal), V (varying across clusters), and I (related to the identity matrix in specifying shape or orientation). In a model with identifier VEI, for example, the retrieved clusters will present varying volumes, equal shapes, and the identity I as the orientation (Fraley et al. 2012).

Table A1. Models adjusted by *mclust*. The 'x' in the third and fourth columns represents the availability in the column Hierarchical Cluster (HC) and Expectation–Maximization (EM). Adapted from Scrucca et al. (2016).

Id	Σ_k	HC	EM	Distribution	Volume	Shape	Orientation
E		x	x	(univariate)	Equal		
V		x	x	(univariate)	Variable		
EII	λI	x	x	Spherical	Equal	Equal	NA
VII	$\lambda_k I$	x	x	Spherical	Variable	Equal	NA
EEI	λA		x	Diagonal	Equal	Equal	Coordinate axes
VEI	$\lambda_k A$		x	Diagonal	Variable	Equal	Coordinate axes
EVI	λA_k		x	Diagonal	Equal	Variable	Coordinate axes
VVI	$\lambda_k A_k$		x	Diagonal	Variable	Variable	Coordinate axes
EEE	$\lambda D A D^\top$	x	x	Ellipsoidal	Equal	Equal	Equal
EVE	$\lambda D A_k D^\top$		x	Ellipsoidal	Equal	Variable	Equal
VEE	$\lambda_k D A D^\top$		x	Ellipsoidal	Variable	Equal	Equal
VVE	$\lambda_k D A_k D^\top$		x	Ellipsoidal	Variable	Variable	Equal
EEV	$\lambda D_k A D_k^\top$		x	Ellipsoidal	Equal	Equal	Variable
VEV	$\lambda_k D_k A D_k^\top$		x	Ellipsoidal	Variable	Equal	Variable
EVV	$\lambda D_k A_k D_k^\top$		x	Ellipsoidal	Equal	Variable	Variable
VVV	$\lambda_k D_k A_k D_k^\top$	x	x	Ellipsoidal	Variable	Variable	Variable

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