The intra-cluster light as a tracer of the total matter density distribution: a view from simulations

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ABSTRACT
By using deep observations of clusters of galaxies, it has been recently found that the projected stellar mass density closely follows the projected total (dark and baryonic) mass density within the innermost ∼ 140 kpc. In this work, we aim to test these observations using the Cluster-EAGLE simulations, comparing the projected densities inferred directly from the simulations. We compare the iso-density contours using the procedure of Montes & Trujillo (2019), and find that the shape of the stellar mass distribution follows that of the total matter even more closely than observed, although their radial profiles differ substantially. The ratio between stellar and total matter density profiles in circular apertures, shows a slope close to −1, with a small dependence on the cluster’s total mass. We propose an indirect method to calculate the halo mass and mass density profile from the radial profile of the intra-cluster stellar mass density.

Key words: galaxies: clusters: general – methods: numerical

1 INTRODUCTION
Ten to more than thirty percent of the stellar light of clusters of galaxies comes from a diffuse distribution of stars emitting the so called intra-cluster light (ICL), the inferred fraction depending on the definition of the border between the brightest central galaxy and the diffuse stellar component, the radial extent at which the stellar mass distribution is integrated and the relaxation state of the clusters (e.g., Krick & Bernstein 2007; Gonzalez et al. 2013; Mihos et al. 2017; Jiménez-Teja et al. 2018; Zhang et al. 2019). This distribution is produced by the stripping of stars from galaxies undergoing mergers and tidal interactions during their evolution in the cluster environment (see Mihos 2015, for a review). Due to its low surface brightness, the observational study of the stellar population producing the intra-cluster light has been challenging. An increasing effort towards deep imaging of clusters of galaxies in the recent years, both through individual cluster imaging, up to z ∼ 1.5, (e.g., Mihos et al. 2005; Montes & Trujillo 2014; Burke et al. 2015; Morishita et al. 2017; Ko & Jee 2018; Jiménez-Teja et al. 2018; Montes & Trujillo 2018; DeMaio et al. 2018, 2020), and by stacking observations of multiple clusters (e.g., Zibetti et al. 2005; Zhang et al. 2019) has allowed new insights into the ICL.

One recent, remarkable result achieved with deep imaging is the tight correlation between the distribution of the stellar surface density, inferred from its surface brightness, and the surface density of the total mass, measured by modelling the gravitational lensing signal (Montes & Trujillo 2019, hereafter MT19). MT19 proposed that the surface density of the stellar mass not bound to galaxies should settle in the potential well of the cluster similarly to the dark matter. This could be used to trace the total matter distribution of clusters within a cluster-centric distance set by the depth of the observations. They also compared their result with total mass surface densities inferred from the X-ray emission of the intra-cluster medium, and concluded that this method is limited by the misalignment of the gaseous component with respect to the dark matter and stellar mass in non-relaxed clusters. Their quantitative analysis made use of the Modified Hausdorf Distance (MHD) (Dubuisson & Jain 2000).
Figure 1. Projected density of stars (top) and matter (bottom) for three different clusters of the C-EAGLE simulations at $z = 0.352$. The secondary density peak in the Cluster CE-21 (middle panel) can be due to a recent major merger event which has stripped stars from the interacting galaxies, or numerical artifacts produced by SUBFIND not being able to correctly assign stellar particles to substructures.

1994) to quantify the deviation between iso-density contours of stars and total matter. They found that, in general, the stellar surface density has smaller MHD values than that of the intra-cluster medium where both are compared with the iso-density contours of total mass.

In this Letter, we test this observational result with state-of-the-art cosmological, hydrodynamic simulations of the Cluster-EAGLE project (C-EAGLE, Barnes et al. 2017; Bahé et al. 2017). We give a brief description of the simulations in the next section and a description of the analysis in section 3. The main results of this work are shown in section 4 and discussed in section 5, along with some concluding remarks.

2 SIMULATIONS

We have used the set of 30 zoom-in cluster simulations performed within the C-EAGLE project. The simulated clusters are uniformly distributed in the mass range $10^{14} < M_{\text{200}}/M_\odot < 10^{15.4}$, where $M_{\text{200}}$ is the halo mass. The simulations were performed with the EAGLE model for galaxy formation and evolution, with the AGNDT9 calibration (Schaye et al. 2015). They provide a physical spatial resolution of $\epsilon = 0.7 \text{kpc}$ (at $z < 2.8$) and baryonic mass resolution of $m_{\text{gas}} \approx 1.81 \times 10^6 M_\odot$. For more information on the EAGLE model and its comparison with global relations of the observed galaxy population, the reader is referred to Schaye et al. (2015) and Crain et al. (2015).

For more details on the numerical algorithms describing photo-ionization equilibrium cooling, star formation, stellar evolution, stellar feedback, black hole growth and feedback, and the hydrodynamic scheme we refer the reader to Wiersma et al. (2009a), Schaye & Dalla Vecchia (2007), Wiersma et al. (2009b), Dalla Vecchia & Schaye (2012), Rosas-Guevara et al. (2015), and Schaller et al. (2015), respectively.

For the results presented here, we have used the particle data, friends-of-friends and SUBFIND (Dolag et al. 2009) groups at $z = 0.352$ to match the average redshift of the Hubble Frontier-Fields clusters (Lotz et al. 2017). Furthermore, the same analysis was performed at $z = 0$, and we found no significant difference. Throughout the paper we assume the cosmological parameters of the C-EAGLE simulations, $(\Omega_\text{m}, \Omega_\Lambda, h, n_s, \sigma_8) = (0.307, 0.693, 0.6777, 0.961, 0.8288)$ (Planck Collaboration et al. 2014), where $\Omega_\text{m}$ and $\Omega_\Lambda$ are the matter and dark energy fractions, $h$ is the Hubble constant in units of $100 \text{kmMpc}^{-1} \text{s}^{-1}$, $n_s$ and $\sigma_8$ are the spectral index and the power spectrum normalisation used to generate the initial conditions.

In the analysis, we have used all particles belonging to the main halo of the largest friends-of-friends group in each simulation, i.e., we excluded all particles bound to satellite galaxies and substructures within the same friends-of-friends group. Maps of projected stellar and total matter density were produced with a spatial resolution of 5 kpc, in order to mimic the spatial resolution employed in the analysis of the observational data ($3 \times 3 \text{arcsec}^2$ at $z \approx 0.35$). We have re-

\footnote{$M_{\text{200}}$ is the mass enclosed in a sphere of radius $r_{\text{200}}$, whose mean density equals 200 times the critical density of the Universe.}
There is no evidence for a correlation with halo mass. The solid line marks the average value of $f_{\text{ICL}}$, and the dashed lines the spread around it. We repeated the analysis with higher ($3.75$ kpc) and lower ($7.5$ kpc) resolution without finding any remarkable difference. The main advantage with respect to observations is that there is no need of masking the light of satellite galaxies. However, debris from tidal interactions between galaxies will be included in the projected matter density. Furthermore, there are biases due to SUBFIND failing to assign stellar particles to satellites (Bahé et al., in prep).

Examples of the projected stellar and total mass density are shown in Fig. 1 for three simulated clusters of increasing virial mass. The top row corresponds to the projected density of stars, while the bottom row shows the density of total matter (dark and baryonic).

Uncertainties on the amount of ICL mass produced and its radial distribution may arise from the modelling of the star formation rate and the spatial and mass resolution of numerical simulations. The EAGLE model matches quite accurately the observed stellar mass and luminosity functions (Schaye et al. 2015; Trayford et al. 2015). Moreover, it reproduces the evolution of the stellar mass function and the observationally inferred density of stars in the universe up to high redshift ($z = 7$) (Furlong et al. 2015). However, while the reference simulation matches the observed sizes of galaxies over several decades in stellar mass, the AGNnt9 calibration yields an offset in the relation towards more compact galaxies (Schaye et al. 2015). This last point seems to be relevant in the interpretation of the ICL mass fractions described in the next section, where the inferred values are on the low side of the distribution of those derived from observations (see references in section 1): compact galaxies are less prone to stripping. On the other hand, Henden et al. (2019) noted that having too large in size galaxies in their simulations boosts the effect of tidal stripping, increasing the fraction of stellar mass in the ICL, and that uncertainties in galaxy sizes are the major contributors to the uncertainty in the determining the fraction of mass in the ICL in simulations.

3 ANALYSIS

Before describing the methodology used in the analysis of the simulation data, we briefly discuss a consistency check for the simulated clusters. We computed the fraction of stellar mass in the ICL, $f_{\text{ICL}}$, and compared it with expected observational and theoretical values. For the sake of ease, we adopted the methodology of (Rudick et al. 2011). The mass fraction has been computed as the stellar mass with projected stellar density below some threshold surface brightness, $\mu$, with respect to the total stellar mass within $r_{200}$. As in (Rudick et al. 2011), we have converted the stellar surface density into surface brightness assuming a constant mass-to-light ratio of $5 M_{\odot}/L_{\odot}^{\text{B}}$, and set $\mu = 26.5$ mag arcsec$^{-2}$ as the threshold.

We show in figure 2 the computed $f_{\text{ICL}}$ as function of halo mass. We find that $f_{\text{ICL}} = 0.091 \pm 0.013$ (solid and dashed lines), with no significant correlation with the total mass of the clusters (the Pearson correlation coefficient is 0.0063). The result is consistent with that of (Rudick et al. 2011). Although the range of halo masses in our sample is rather narrower, similar fractions and the lack of correlation have been reported by (Pillepich et al. 2018), when using a definition of the ICL related to the size of the central galaxy. The result is consistent with previous simulations (Rudick et al. 2011; Contini et al. 2014), where they applied semi-analytical models to N-body simulations, and hydrodynamical cosmological simulations (Pillepich et al. 2018; Henden et al. 2019). Finally, observations using similar thresholds have reported as well similar mass fractions (Krict & Bernstein 2007; Montes & Trujillo 2014).

We have followed a methodology similar to MT19 to extract iso-density contours. We computed circularly averaged radial profiles of the density of the stellar and total mass. For this, we take the position of the minimum of the potential energy as centre of the cluster (McAlpine et al. 2016). The projected densities for drawing the contours were selected interpolating the profiles at radii of $50, 75, 100, 125, 140$ kpc (the distances used by MT19) for the inner part, and of $170, 220, 300, 460, 620, 780, 940, 1100$ kpc for the outer regions, and only up to $r_{200}$. At large distance from the centre of the clusters ($r > 140$ kpc), we down-sample the images merging $4 \times 4$ pixels, thus degrading the spatial resolution to 20 kpc, to smooth the otherwise very noisy contours. The contours of the projected densities are shown in Fig. 3, for the same three clusters as depicted in Fig. 1. The projected total mass density contours are drawn with blue dotted lines, and the projected stellar density contours with red dashed lines, where a darker colour indicates a smaller radius. The top row is a close-up view of the contours near the centre of the clusters, out to 140 kpc, whereas the contours at larger distances are shown in the bottom row.

We measured projected radial distances from the centre of the cluster instead of elliptical distances to the centre of

2 We used the contour function of matplotlib to compute the contours.
the brightest central galaxy, as usually done in observations. This simplification is not crucial to derive the iso-density contours, as it only changes the values of density at which the contours will be drawn. In practice, this means that the distances we use are systematically different from those of MT19, the difference depending on the eccentricity of the brightest central galaxy, or the presence of more than one central galaxy, that we excluded from the analysis, or both. As this is only an exploratory analysis we ignore these differences.

As in MT19, to compare the shape of the contours, we estimated the Modified Hausdorff distance (MHD) defined by Dubuisson & Jain (1994):

\[
d_{\text{MH}}(X, Y) = \max \left( d(X, Y), d(Y, X) \right),
\]

where

\[
d(X, Y) = \frac{1}{N_X} \sum_{x \in X} \min_{y \in Y} \|x - y\|.
\]

The two samples, \( X \equiv \{x_1, x_2, \ldots, x_{N_X} \} \) and \( Y \equiv \{y_1, y_2, \ldots, y_{N_Y} \} \), contains the points defining two contours, and \( \| \cdot \| \) is the Euclidean norm. As we may have different closed contours for the same density value, we select for each distance the contour composed by the largest number of segments. The selected contours are shown in Fig. 3.

When measuring the MHD close to the virial radius of the clusters, we would expect an increase of its value, as the outskirts of clusters are not dynamically relaxed and fewer stellar particles are populating it, producing noisier contours. In order to compare the MHD across different distances, we define the relative MHD as

\[
\zeta = \frac{d_{\text{MH}}(r)}{r},
\]

where \( r \) is the distance at which the iso-density contours have been computed. This way, we are measuring deviations as fraction of the distance. We find that this definition removes almost entirely the correlation with distance.

4 RESULTS

In the left panel of Fig. 4, we show with the blue, solid line the mean value of \( d_{\text{MH}} \), the shaded area depicts the 1-\( \sigma \) confidence interval. We overplotted the MHDs calculated by MT19, as well as their 1-\( \sigma \) area, in green. For sake of clarity, observational points for individual cluster are slightly displaced along the x-axis. From that panel, we can highlight that:
pared the density profiles of the stellar component. In Fig. 5 total matter and stellar mass distribution, we have also com-
lensing models is limited. non-continuous contours and the spatial resolution of the
dnegligible with coarse datasets, as two similar shapes can have a non-
putation, but it has to be taken into account when dealing
not continuous segments. This obviously simplifies the com-

(i) the $d_{\text{MH}}$ from both simulations and observations are
of the same order of magnitude;
(ii) they show the same trends with radius;
(iii) and simulations have a $\sim 50\%$ lower $d_{\text{MH}}$ than observ-
ations, with smaller scatter.

As $d_{\text{MH}}$ increases monotonically with the distance at
which it is computed, we introduced the relative MHD, $\zeta$,
to obtain a distance-free similarity measurement. We show
in Fig. 4 (right panel) the distribution of $\zeta$ for all contours
and clusters, in blue, and the $\zeta$ extracted from MT19’s data,
in green. Most of the values of $\zeta$ are lower than those ob-
erved: 96 percent of the relative MHDs are below the mean
observed value. The shape of the distribution is remarkably
close to a Gaussian distribution in logarithmic space, with
mean $\langle \zeta \rangle = 0.107$ and dispersion $\sigma_{\zeta} = 0.080$, indicating that
$\zeta$ is a solid, scale-free estimate of the similarity of contours
at any cluster-centric distance.

We would like to highlight two relevant issues with the definition
of $d_{\text{MH}}$ that can bias the observational values to-
wards higher values. The $d_{\text{MH}}$ is defined based on points and
not continuous segments. This obviously simplifies the com-
putation, but it has to be taken into account when dealing
with coarse datasets, as two similar shapes can have a non-
negligible $d_{\text{MH}}$. Second, each point’s contribution is defined
positive and with respect to the other set of points. This
provides a distance that increases monotonically with noise
(Dubuisson & Jain 1994), thus special care must be taken
when dealing with data with low signal-to-noise or large un-
certainties. Both these points could be driving the observed
$d_{\text{MH}}$ towards higher values, as masking galaxies introduces
non-continuous contours and the spatial resolution of the
lensing models is limited.

In addition to the study of the similarity between the
total matter and stellar mass distribution, we have also com-
pared the density profiles of the stellar component. In Fig. 5
(left panel) we show circularly averaged density profiles of
the stellar particles. They follow a power-law behaviour up
to $\sim 500$ kpc for the lightest halos, and $\sim 1$ Mpc for the more
massive ones. At such distances, the interactions between
substructures are weaker, and fewer particles get ejected to
the intra-cluster medium, thus they can no longer success-
fully trace the potential well. In the right panel of Fig. 5 we
show the ratio between the stellar and total matter density
profiles. This ratio is close to a power law with scatter of
0.1 dex and a slope of about $-1$. We have performed a fit
to all the profiles at once, with and without normalising the
radial distance using $r_{200}$, yielding the relations:

$$
\log_{10} \Sigma_{\text{tot}} = \log_{10} \Sigma_{\star} + (1.115 \pm 0.005) \log_{10} r - (0.25 \pm 0.01), \quad (4)
$$

$$
\log_{10} \Sigma_{\text{tot}} = \log_{10} \Sigma_{\star} + (1.085 \pm 0.004) \log_{10}(r/r_{200}) + (3.144 \pm 0.005). \quad (5)
$$

The residuals of both fits have a similar scatter: 0.147 and
0.127 dex for equations 4 and 5, respectively. We recall that
the AGNdT9 feedback calibration, used in the C-EAGLE
simulations, yields more compact galaxies than the reference
model for stellar masses $M_{\star} > 10^{10} M_{\odot}$. The less efficient
tidal stripping may therefore deposit more stellar mass closer
to the centre of the cluster, resulting in a steeper density
profile. However, this bias may be of secondary importance,
at least within the central 100 kpc (Bahé et al., in prep.).

We propose a new, indirect way of measuring a cluster’s
mass knowing its stellar density profile in the innermost re-
gion. First, via deep imaging as that performed by MT19,
the stellar density profile can be obtained and extrapolated
up to $r_{200}$ assuming a power law. Then, using equation 4 or
5, the total mass density profile can be computed. This pro-
file can be integrated to obtain an estimation of the cluster’s
total mass.

This procedure would be similar to that proposed by
Pillepich et al. (2018). In that case, however, only the power law slope of the 3D stellar mass density profile was used to infer the total halo mass, in our case we use more information (the 2D stellar density profile and equation 4 or 5), expecting less scatter in the mass estimate.

5 DISCUSSION & CONCLUSIONS

We have studied the similarity of the projected stellar and total matter distributions in the halos of massive galaxy clusters using the C-EAGLE set of 30 zoom-in simulations of clusters of galaxies. In the analysis, we considered as constituents of the diffuse distribution of stellar mass only particles in the friends-of-friends group that were not assigned to any substructure by the SUBFIND algorithm.

We can summarise our results as follows:

(i) we confirm the finding of MT19: the projected distribution of stars closely follow the projected distribution of the total mass, although their radial profiles differ substantially;

(ii) the ICL, approximated as those stars in the region where \( \mu > 26.5 \text{ mag arcsec}^{-2} \) (\( \Sigma_\ast \approx 1.4 \times 10^6 \text{ M}_\odot \)), accounts for \( \sim 10 \) percent of the stellar content of the cluster within \( r_{200} \); this fraction does not show any correlation with the mass of the cluster;

(iii) the ratio between the surface density profiles of the stellar to the total matter follows a simple power-law up to the virial radius, equations 4 and 5; as the slope and amplitude of the stellar surface density profile can be extracted from observation, we proposed a method to estimate the total mass surface density profile, thus the mass of the halo;

(iv) the similarity between the stellar and total matter distributions in the cluster halo is even higher in the simulations than that observed by MT19 (Fig. 4); This indicates that stars closely trace the underlying gravitational potential;

(v) in order to show any self-similarity, we have introduced the relative measure, \( \zeta = d_{\text{BH}}/r \), whose distribution resembles a log normal when using all the clusters and contours pairs; the parameter \( \zeta \) could be used to study the relaxation state of a cluster; the maximum of this distribution is located at \( \zeta \sim 0.1 \), thus the typical \( d_{\text{BH}} \) is about 10\% of the distance at which it is computed.

The study of the spatial distribution of the ICL can be used to infer, in high detail, the distribution of the underlying dark matter in clusters of galaxies. Moreover, the average density profile of total matter can be extracted, and extrapolated up to the virial radius, only by measuring the slope of the stellar mass density profile and its normalisation close to the centre of the cluster. This is complementary to the study of Pillepich et al. (2018), where only the total halo mass was given as function of the slope of the 3D stellar density profile, with larger uncertainty.

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