Characterizing the Uncertainty in Cluster Magnetic Fields Derived from Rotation Measures

A. R. Johnson1,2, L. Rudnick2, T. W. Jones2,3, P. J. Mendygral2,4, and K. Dolag5,6
1 Optum, Inc., 15625 Technology Dr., Eden Prairie, MN 55344, USA
2 Minnesota Institute for Astrophysics, School of Physics and Astronomy, University of Minnesota, 116 Church St., SE, Minneapolis, MN 55455, USA; larry@umn.edu
3 Minnesota Supercomputing Institute, University of Minnesota, Minneapolis, MN 55455, USA
4 Cray Inc., 2131 Lindau Lane, Suite 1000, Bloomington, MN 55425, USA
5 Universitäts-Sternwarte, Fakultät für Physik, Ludwig-Maximilians Universität München, Scheinerstrasse 1, D-81679 München, Germany
6 Max-Planck-Institut für Astrophysik, Karl-Schwarzschild Strasse 1, D-85748 Garching bei München, Germany
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Abstract

Magnetic fields play vital roles in intracluster media (ICMs), but estimating their strengths and distributions from observations is a major challenge. Faraday rotation measures (RMs) are widely applied to this task, so it is critical to understand inherent uncertainties in RM analysis. In this paper, we seek to characterize those uncertainties given the types of information available today, independent of the specific technique used. We conduct synthetic RM observations through the ICM of a galaxy cluster drawn from a magnetohydrodynamic cosmological simulation in which the magnetic field is known. We analyze the synthetic RM observations using an analytical formalism based on commonly used model assumptions, allowing us to relate model physical variables to outcome uncertainties. Despite the simplicity of some assumptions, and known physical parameters, we are able to extract an approximate magnitude of the central magnetic field within an apparently irreducible uncertain factor of ≈3. Principal, largely irreducible, uncertainties come from the unknown depth along the line of sight of embedded polarized sources, the lack of robust coherence lengths from area-constrained polarization sampling, and the unknown scaling between ICM electron density and magnetic field strength. The RM-estimated central magnetic field strengths span more than an order of magnitude including the full range of synthetic experiments.

Unified Astronomy Thesaurus concepts: Intracluster medium (858); Galaxy clusters (584); Extragalactic magnetic fields (507); Polarimetry (1278)

1. Introduction

The hot, diffuse media of galaxy clusters (intracluster media, ICMs) are magnetized (e.g., Carilli & Taylor 2002) and very likely turbulent (e.g., Schuecker et al. 2004; Sanders et al. 2011; Miniati & Beresnyak 2015; Vazza et al. 2018, 2017). The strength and structure of the magnetic fields play central roles in determining the turbulent and thermodynamical properties of the ICMs. For instance, these properties may control the scale and isotropy of transport processes such as viscosity and thermal conduction, even if the magnetic, Maxwell stresses are insignificant on cluster scales. Magnetic fields or their induced anisotropic transport characteristics can stabilize structures such as cold fronts (e.g., Zuhone et al. 2010) or lead to instabilities that, for example, influence cluster thermal structure (e.g., Parrish et al. 2012). ICM magnetic fields also control the acceleration and transport of relativistic particles within the ICMs (e.g., Brunetti & Jones 2014). For these reasons, much effort has gone into observational estimates of ICM magnetic field properties. As discussed below, there are key physical properties of the field and of the accompanying thermal plasma that are not well constrained observationally, and therefore lead to irreducible uncertainties in the structure and strength of the derived magnetic fields. It is the goal of this paper to characterize the most important uncertainties in this process.

Cluster magnetic fields reveal themselves through the diffuse synchrotron emission found in a good many clusters (such as giant radio halos, e.g., Feretti et al. 2012; Brunetti & Jones 2014) and through the Faraday rotation of linearly polarized radio emission propagating through the ICM. Faraday rotation is based on circular birefringence, in which the plane of polarization of a linearly polarized signal rotates along one with the line of sight by an amount that is proportional to the square of the plane of polarization of a linearly polarized signal rotates along the line of sight (i.e., $\Delta \phi = RM \times \lambda^2$). The constant of proportionality, the rotation measure (RM), is determined by the integral along the line of sight of the plasma electron density, $n_e$, times the projection of the vector magnetic field onto the line of sight, $B_\parallel$; namely over path length $\ell$,

$$\text{RM}(\ell) = 812 \int_0^\ell n_e(s)B_\parallel(s) \, ds \, \text{rad m}^{-2},$$

where $n_e$ is expressed in units of cm$^{-3}$, $B_\parallel$ is in units of $\mu$G, and the differential line of sight path, $ds$, is in units of kpc. We assume, going forward, that foreground RM contributions (e.g., Galactic) have been removed and that we can ignore extraneous modifications to the ICM due, for example, to active galactic nucleus (AGN) outflows.

RM analyses of ICMs have been applied extensively, using polarized synchrotron sources embedded in the cluster (e.g., Taylor & Perley 1993; Feretti et al. 1999; Eilek & Owen 2002; Govoni et al. 2006, 2017; Taylor et al. 2006; Guidetti et al. 2008, 2010; Laing et al. 2008; Vacca et al. 2012) as well as multiple unresolved polarized sources behind galaxy clusters. (e.g., Clarke et al. 1999; Feretti et al. 1999; Bonafede et al. 2010) RM measurements have typically suggested maximum magnetic field strengths at cluster centers $\sim 1-15 \mu$G, with strengths steadily decreasing outward. Several recent
magnetohydrodynamic (MHD) cosmological simulation studies (e.g., Donnert et al. 2009; Xu et al. 2012) have obtained evolved cluster magnetic field distributions that are relatively insensitive to the model details for the seed fields, and that are qualitatively consistent with the reported observations.

So far, however, neither the simulated nor the RM-based “observed” field distributions provide the sufficiently robust information needed to constrain ICM physics and evolution. In this paper we aim to examine some of the factors that limit the accuracy of ICM RM studies, specifically targeting limitations coming from inherent, irreducible uncertainties in such analyses. In order to do this we have adopted a simple analytic approach that clearly exposes such limitations. We emphasize at the start that our intent is not to promote any particular analysis method, but to identify inherent model and physical uncertainties that seem to restrict accuracy to something close to an order of magnitude.

Application of Equation (1) to establish the magnetic field distribution along a given line of sight within an ICM is, in fact, not straightforward. First, it requires that one isolate the electron density distribution, \( n_e(l) \), from the magnetic field distribution. That can often be done reasonably well in clusters by modeling thermal X-ray measurements. A more serious issue is revealed by the fact that measured ICM RM distributions are patchy and irregular, often including RMs of both signs, telling us that the magnetic fields are disordered. Thus, \( B_{\parallel} \) contributions to Equation (1) are stochastic; a statistical analysis is mandated. In practice one usually tries to estimate the dispersion of the magnetic field strength distribution, \( \sigma_{B_{\parallel}} \).

On the reasonable and common assumption that the ICM magnetic field averaged over large volumes is isotropically disordered by turbulence with some magnetic field coherence scale, \( L_{\text{coh}} \ll \ell \), with \( \ell \) the full path length, an ensemble of statistically independent lines of sight lead to \( \langle \text{RM} \rangle \to 0 \) while the RM dispersion, \( \sigma_{\text{RM}} = \sqrt{\langle (\text{RM})^2 \rangle - \langle \text{RM} \rangle^2} \to \sqrt{\langle \text{RM}^2 \rangle} \). More to the point of the exercise, \( \sigma_{\text{RM}} \propto \sigma_{B_{\parallel}} = \sqrt{(B_{\parallel}^2 - \langle B_{\parallel} \rangle^2) \to \sqrt{(B_{\parallel}^2) / 3} \). In the idealized case of a homogeneous medium, and independent lines of sight, \( \sigma_{\text{RM}} \propto \sqrt{L_{\text{coh}} \ell} \). There are many related metrics in the literature for magnetic field characteristic scales, as outlined in the Appendix. These include, for example, the magnetic field autocorrelation scale, \( L_{\text{aut}} \), which is the power-spectrum-weighted mean length associated with the magnetic field variations (Cho & Ryu 2009). Each has been applied in the literature to model associated RM distributions in MHD turbulence. For an isotropic, homogeneous, turbulent magnetic field the two lengths are simply related by \( L_{\text{aut}} = 2L_{B} \).

If we choose a characteristic scale, \( \Lambda = (3/2) L_{\text{aut}} = (3/4) L_{\text{int}} \), then the relation between \( \sigma_{\text{RM}} \) and \( \sigma_{B_{\parallel}} \) conveniently takes the familiar normalization in Equation (1); namely (e.g., Lawler & Dennison 1982; Tribble 1991; Feretti et al. 1995; Felten 1996)

\[
\sigma_{\text{RM}} = 812 \bar{n}_e \sigma_{B_{\parallel}} \sqrt{\Lambda} \approx 812 \bar{n}_e \sigma_{B_{\parallel}} \Lambda \sqrt{\ell / \Lambda},
\]

where the units are as in Equation (1). We will adopt this common convention below. We also need to clarify the difference between \( \Lambda \), the projected, 2D-observable RM coherence length and 3D magnetic field structures, which are not directly observable. In fact, as noted by Enßlin & Vogt (2003), even for an isotropic magnetic field in a homogeneous medium, the 3D magnetic field correlation length is not a direct proxy for \( \Lambda \) in Equation (2). Furthermore, note from the beginning that, unless the various lines of sight sample regions with uncorrelated magnetic fields, so include separations \( > \Lambda \), but regions with similar electron densities, the calculated RM dispersion measure from Equation (2) is not identical to \( \sigma_{\text{RM}} \) among different lines of sight. This equivalence cannot be taken for granted in observational analyses based on sparse RM information (e.g., Murgia et al. 2004).

There are a number of important challenges in applying Equation (2) to actual data; these lead to limitations in the accuracy with which we can characterize the magnetic field strengths in clusters. These challenges include:

1. the inhomogeneous and anisotropic magnetic field structure in realistic ICMs, as illuminated in MHD cluster formation simulations;
2. the unknown position of cluster-embedded polarized sources along the line of sight through the ICM;
3. the unknown scaling between magnetic field and thermal plasma density—the latter, in turn, must be modeled using X-ray or microwave observations;
4. the incomplete sampling coming from limited availability of polarized emissions of appropriately large scales;
5. for sight lines penetrating the core, the bulk of the emission comes from a very small number of high emissivity regions, so the sampling of independent field regions is reduced;
6. the unknown radial dependence of characteristic spatial scales in the ICM turbulence;
7. the fact that \( \sigma_{\text{RM}} \) and \( \Lambda \) are inherently statistical quantities, with distributions that can only be sampled, but not fully measured; Newman et al. (2002) have argued that the inherent uncertainties in their estimation are comparable to their magnitudes;
8. modification of the local magnetic field and density structure by radio lobe interactions with the ICM, as discussed briefly below.

The contribution of RM variations local to an embedded radio galaxy, as opposed to RM variations that characterize the unperturbed ICM, is an open issue. Guidetti et al. (2011, 2012), for example, reported anisotropic RM patterns associated with several cluster AGNs, implying, they argued, influence of the AGNs on the RM distribution. Rudnick & Blindell (2003) argued that the RM distributions observed in front of several embedded cluster sources showed biases suggesting influence from plasma entrained within the emitting radio lobes. Enßlin et al. (2003) argued, however, that those biases were not statistically robust. On the theoretical front, Huarte-Espinosa et al. (2011) conducted synthetic RM observations of MHD simulations of high-powered, FR II AGN jets driven into a model ICM with an initially isotropic, random magnetic field, and found that the jet interactions modified the RM statistics in ways that biased the inferred magnetic field values upward by as much as 70%. The complex issue of physical modification of the local ICM interacting with a radio source is beyond the scope of this paper and not addressed.

It is the purpose of this paper to use extensions of Equation (2) to explore and quantify many of these uncertainties. Our approach is to carry out synthetic RM measurements of a magnetized ICM extracted from a high-
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resolution MHD cosmological simulation (see Mendygral et al. 2012 and references therein). The usefulness of such synthetic observations from cosmological simulations is demonstrated, e.g., by Xu et al. (2012). In our case, the simulated cluster magnetic field evolved from an initially weak but uniform field seeded at high redshift (\(z = 20\)). Since the cluster and its magnetic field formed dynamically according to the current cosmological paradigm, there are no artificial biases that would influence our analysis other than what result from finite numerical resolution in the magnetic field distribution and omission of ICM radiative or conductive cooling and galaxy feedback. Those details are largely irrelevant to the exercise at hand, which primarily aims to establish the reliability of estimations for existing ICM magnetic fields using standard RM methods.

We report two complementary sets of analysis experiments. In the first experiments, we obtain and analyze results for a fully sampled background polarized screen, which offers the optimal information potentially available to an observer. We also analyze a fully sampled screen placed halfway along the line of sight through the cluster (the “mid-plane” experiment), in order to confirm assumed scaling relations. In the second set of experiments, we examine and compare results based on RM measurements for discrete sets of finite-sized, polarized sources, embedded as passive objects in the cluster. We refer to these sources as “masks” in our analysis.

We note that some recent methods (Vacca et al. 2012; Govoni et al. 2017) of estimating intracluster magnetic field properties from observed RMs use Monte Carlo simulations and a Bayesian analysis with an assumed power-law distribution of magnetic field fluctuations. This approach offers distinct advantages by characterizing some, although not all, of the uncertainties associated with the above issues. We will discuss these further in the context of our findings in Section 5.1.

The outline of the remainder of our paper is as follows: in Section 2, we describe the physical properties of our simulated cluster and its ICM, along with beta-law fits for the electron density and magnetic field distributions for use in modeling analysis of the synthetic RM observations. In Section 3 we outline the procedures for translating the RM statistics for a fully sampled background screen into ICM magnetic field properties. Magnetic field estimates derived from discrete polarized patches (both “mid-plane masks” and “background masks”) for RM measurements are presented in Section 4. Further discussion of analysis issues is presented in Section 5, while our conclusions are outlined in Section 6. In the Appendix, we discuss the expected relationships between RM and magnetic field coherence lengths and our methods for estimating these.

2. The Test Cluster Properties

2.1. ICM Evolution

The ICM used in our study is the diffuse baryonic component of the \(1.5 \times 10^{14} M_\odot\) cluster g676 (Dolag et al. 2009; Stasyszyn et al. 2013) extracted at a redshift of \(z \approx 0\) from a very high-resolution ΛCDM cosmology simulation (\(h = 0.7, \Omega_M = 0.3,\) and \(\Omega_\Lambda = 0.7\)). The simulation was carried out with an MHD implementation of the SPH GADGET-3 code (Dolag & Stasyszyn 2009). The magnetic field in that cluster developed from a uniform, primordial field of strength \(10^{-11} G\) at \(z = 20\). For our study here we mapped and centered the evolved cluster ICM and, to allow continued, short-term dynamical evolution mentioned in Section 4, applied a spherical approximation to the total gravitational potential onto a 1 Mpc × 1 Mpc Cartesian grid (called “the cluster analysis box” below) of uniform resolution with \(\Delta_x = \Delta_y = \Delta_z = 1\) kpc. Details of the g676 cluster and the brief simulation extension that produced the data we used can be found in Mendygral et al. (2012) and references therein, used for the study of jet propagation. This cluster was chosen for its lack of recent mergers (the last major merger occurred \(7\) Gyr prior to the epoch of our RM experiments) and a relatively relaxed morphological appearance, based on synthetic thermal X-ray observations (Mendygral et al. 2012).

Despite this rather long period devoid of major disruptions and the relatively relaxed morphology, the g676 ICM at \(z \approx 0\) still contains significant dynamical features; in particular, there are large-scale “sloshing” motions due to gravitational interactions with subhalos, with flow velocities approaching the cluster sound speed (\(c_s \approx 650 \text{ km s}^{-1}\)), so Mach \(\approx 1\). The sloshing velocity field is evident in Figure 5 of Mendygral et al. (2012). Spiral density and magnetic field structures expected from the sloshing (e.g., Ascasibar & Markevitch 2006) are obvious in Figure 1 of the present paper. Small, isolated density clumps visible in Figure 1 reveal subhalos present in the original cluster formation simulation. Figure 6 in Mendygral et al. (2012) provides a synthetic X-ray image of this cluster along the same axis at approximately the same time. Although the cluster is somewhat aspherical in that image, the sloshing structures visible in our Figure 1 are not obvious.

2.2. Spherically Modeled Distributions

While the various deviations from spherical symmetry outlined above are natural and relate to the dynamical state of the cluster, they are not dominant, and for this RM model study we follow the standard practice of constructing symmetric averages for the density and magnetic field distribution. Existing asymmetries will, of course, contribute largely irreducible errors to the RM analyses, although on scales of order the cluster core size, both electron density and magnetic field distributions are actually reasonably spherically symmetric and, in the latter case, reasonably isotropic. Based on spherical averages, the core of the g676 ICM has a radius \(r_c \approx 40\) kpc, central mass density \(\rho \sim 10^{-23} \text{ g cm}^{-3}\) (\(n_e \sim 4 \times 10^{-2} \text{ cm}^{-3}\)), temperature \(kT \approx 1.6\) keV, and pressure \(P_g \approx 2 	imes 10^{-10} \text{ dyne cm}^{-2}\).

The spherically averaged ICM mass density at \(r = 500\) kpc is \(\rho \sim 3 \times 10^{-28} \text{ g cm}^{-3}\), so this radius very roughly corresponds to \(R_{500}\). The g676 ICM is modestly turbulent with core turbulent velocities \(\sim 50\) km s\(^{-1}\) (Zharulava et al. 2011). These properties correspond to core turbulent pressures \(<1\%\) of the total pressure. Outside the core, turbulent velocities increase to \(\sim 100\) km s\(^{-1}\), but still contribute \(<5\%\) of the total pressure. The cluster magnetic field has values locally as large as \(12\) \(\mu G\) in some strong central filaments visible in Figure 1, but field strengths more typically fall into a range \(B \sim 0.5\)–4 \(\mu G\). The rms core field strength is \(\approx 2\) \(\mu G\), corresponding to a magnetic pressure in the cluster core \(P_B \sim 10^{-13} \text{ dyne cm}^{-2} \sim 0.1\% P_g\). Thus, the central plasma \(\beta_p = P_p/P_B \sim 10^1\). Field strengths decrease outward, so that \(B \sim 0.1\) \(\mu G\) near \(r \sim 500\) kpc, producing a magnetic pressure that is again \(\sim 0.1\% P_g\). Evaluating Equation (1) while assuming \(B_\parallel \sim B \sim 2\) \(\mu G\) and \(L \sim 2 \times 40\) kpc as uniform values
through the core would produce \( \text{RM} \sim 4000 \text{ rad m}^{-2} \) as a fiducial RM. The analogous result in the cluster outskirts would be roughly two orders of magnitude smaller. Disorder in the magnetic field then reduces the observed RM values from these estimates by factors of a few within the core and by much larger factors over the full cluster (Figure 5).

Spherically averaged ICM characteristics, represented as functions of cluster radius, \( r \), and in projection functions of projected radius, \( a \), are a standard approach to cluster RM analysis, and underlie our modeling as well. Of course, this overlooks real ICM structures that limit the comparisons between the standard models and the physical ICM. As we set up this basic model, we stress that the specific ICM properties of this cluster are not themselves central to our subsequent, analysis-based conclusions, since the task is to explore uncertainties in analysis outcomes. That is, the underlying questions being explored have to do with the uncertainties inherent in the observational analysis methods to recover the actual cluster properties, whatever they may be. It is also important to note that the model parameters described below are only used to interpret the data; the cluster simulation is not based on such symmetry assumptions.

### 2.2.1. ICM Electron Density

For analysis modeling purposes we express the electron density radial distribution, \( n_e(r) \), in terms of a spherical beta-law profile (Cavaliere & Fusco-Femiano 1976),

\[
n_e(r) = \frac{n_{e,0}}{1 + \left( \frac{r}{r_c} \right)^{2\beta_e}} \quad (3)
\]

where \( n_{e,0} \) is the central electron density, \( r_c \) is the ICM core radius, and \( \beta_e \) represents the galaxy–gas velocity dispersion ratio in the cluster, which is nominally assumed spherically symmetric and isothermal. We display in the upper left panel of Figure 2 the spherically averaged profile of \( n_e(r) \), as well as the best fit to Equation (3). The core radius in the fit is \( r_c = 41 \text{ kpc} \), while \( \beta_e = 0.75 \). Both are consistent with observed properties of real ICMs (e.g., Vikhlinin et al. 2006).

At this point it is useful to emphasize that observable measures generally correspond to integrals along lines of sight; that is, they represent projections. We generally cannot determine 3D quantities directly, but depend on modeling projections. In a spherically symmetric model the relevant positional variable is the projected radius from the cluster center, \( a \), rather than the spherical coordinate, \( r \). It is necessary, for example, to distinguish between \( \sigma_B(r) \), an intensive 3D quantity, and \( \sigma_{RM}(a) \), a projected quantity. We may, in principle, be successful in determining \( \sigma_{RM}(a) \) from observations, but must model the projection to estimate \( \sigma_B(r) \). In this context and an isothermal ICM approximation, the projected electron density squared, \( \Sigma_{n_e^2}(a) \) provides a convenient proxy for the (observable and projected) thermal X-ray surface brightness distribution as a function of \( a \). With a beta-law density profile model

\[
\Sigma_{n_e^2}(a) = \frac{C}{1 + \left( \frac{a}{r_c} \right)^{2\beta_e} \frac{1}{\Gamma(3\beta_e) - 1/2}} \quad (4)
\]

where \( C = 2 \int_0^\infty n_e(r)^2 dr = n_{e,0}^2 \frac{\sqrt{\pi}}{\Gamma(3\beta_e - 1/2)} \) is the square of the electron density integrated along the line of sight through the cluster center. The upper right panel in Figure 2 shows radial profiles of the projected, azimuthally averaged electron density-squared distribution along the three principal axes of our grid along with the beta-law model curve determined by Equation (4) using the azimuthally averaged electron profile shown in the upper left panel of the figure. Deviations in the projected squared density fit to Equation (4), coming especially from the spherical asymmetries noted above, are as large as 15% near the cluster center, but less than 5% at projected distances beyond the core radius.

Note that in order to derive empirical estimates for the central magnetic field, the values of \( n_{e,0} \), \( \beta_e \), and \( r_c \) must first be derived from X-ray observations (real or synthetic, as described above).

### 2.2.2. ICM Magnetic Field

We assume any foreground (e.g., Galactic) RM has been removed, so that the integral in Equation (1) is along the line of sight from the near edge of the cluster to the polarized emission of an embedded source or to the far edge of the cluster for a background source. For the best case scenario, we also assume no other contributions to the RM, e.g., from the immediate environment of the embedded source, or along the post-cluster path to a background source. Although there are several long magnetic filaments related to ICM sloshing motions in the g676 cluster, for modeling purposes, we assume the magnetic field is
locally disordered and isotropic. This is a good assumption as long as RMs can be averaged over sufficiently large scales. As seen in the lower right panel of Figure 2, the magnitude of the mean vector magnetic field, \(|\mathbf{B}|\), is an order of magnitude smaller than the rms field strength, \(\sqrt{\langle B^2 \rangle}\) or its dispersion \(\sigma_B\).

In practice, effective isotropy will be valid provided the RM observations relate to volumes comparable to or larger in size than the cluster core or, as it turns out, equivalently, than the correlation length of the magnetic field, as established below. However, observations of individual radio galaxies may not cover sufficiently large scales for this assumption to apply; over smaller scales the \(|\mathbf{B}|\) within the observed volume can remain large compared to \(\sigma_B\), while the observed estimate for \(\sigma_{RM}\) will be too small to provide reliable information about the physical dispersion in the local magnetic field strength, \(\sigma_B\).

A different modeling issue arises on scales larger than the core radius. Since the characteristic strength of the magnetic field (e.g., \(\sigma_B\)) decreases systematically with distance from the cluster core, as noted in the previous subsection, this behavior must be included for effective modeling. Such a decrease with distance is expected generally in clusters, except where local activity (e.g., AGN jets) may have recently injected a large amount of magnetic flux. The cluster-scale variation in the magnetic field strength has been commonly expressed in terms of a scaling relation between the magnetic field strength and the ICM density (e.g., Dolag et al. 2001; Ryu et al. 2008; Kunz et al. 2011). There are several plausible physical arguments for such a dependence. For instance, flux freezing of a tangled magnetic field during compression would yield \(B \propto \ell^{-2} \propto V^{-2/3} \propto n_e^{2/3}\), where \(V \sim \ell^3\) is the volume containing the magnetic flux. Accounting for work done on a disordered field by the ICM during adiabatic compression would yield \(B \propto V^{-1} \propto n_e\) while, alternatively, assuming during compression that the magnetic energy maintains a fixed ratio with turbulent energy would lead to \(B \propto n_e^{3/2}\). Rather than assume a particular scaling choice, in this paper we will consider \(\sigma_B \propto n_e^{\eta}\), with \(1/2 \leq \eta \leq 1\). Various MHD cluster formation simulations have shown results roughly consistent with such scalings (e.g., Dolag et al. 2005; Vazza et al. 2018).

In general, RM observers would not know \(\eta\) a priori and would, therefore, have to estimate \(\eta\) based on theoretical considerations or simulation results. In cases where there are extensive RM data, one can attempt to estimate \(\eta\) by comparing the radial dependence of \(\sigma_{RM}\) to the radial dependence of \(n_e\) (e.g., Dolag et al. 2001; Guidetti et al. 2008; Bonafede et al. 2010; Govoni et al. 2017). However, the ability to do this depends on the unknown distribution of polarized sources along the line of sight, the assumption of an unchanging...
coherence scale, and the lack of any local effects around the embedded radio galaxies. For 867, since we have full knowledge of \( B \), we can actually determine an approximate value of \( \eta \) as follows. First, we calculate \((n_e(r))\) and \( \sigma_B(r) \) within spherical shells of fixed thickness and uniform logarithmic spacing in \( r \), then compute a least-squares fit for \( \log \sigma_B(r) \) versus \( \log (n_e(r)) \) within the cluster. The result, shown in the lower left panel of Figure 2, produces \( \eta = 0.5 \), with a central, core magnetic field strength dispersion, \( \sigma_B = 1.9 \mu G \).

Using such a density scaling, along with the beta-law density profile model in Equation (4), we can express the spherically averaged magnetic field strength dispersion as a function of radius, \( r \), by

\[
\sigma_B(r) = \frac{\sigma_{B,0}}{\left(1 + \left(\frac{r}{r_c}\right)^{2}\right)^{1/2}},
\]

where \( \beta_c \) and \( r_c \) are obtained from the electron density distribution as described in the previous subsection. The radial profile of \( \sigma_B \) for 867 is shown in the lower right panel of Figure 2 along with a least-squares fit to Equation (5) allowing both \( \eta \) and \( \sigma_{B,0} \) to vary. As expected from the excellent fits for the beta-law density form and the previously established \( \log \sigma_B \) versus \( \log n_e \) form, the best-fit parameters in 867 are consistent again with \( \sigma_{B,0} = 1.9 \mu G \) and \( \eta = 0.5 \).

3. Estimating the Magnetic Field Distribution from the RM Distribution

3.1. The Basic Model

The previous section outlined the actual 3D magnetic field and electron density distributions in our simulated test cluster, 867, and described beta-law model, spherically symmetric radial fits for those distributions. The observational challenge is to recover the (spherically averaged) magnetic field properties as a function of radius, \( r \), in a cluster from an observed distribution of RM, which is a 2D, projected quantity. In particular, the objective is to obtain estimates for \( \sigma_B(r) \) and \( \sigma_{RM}(r) \), as a function of projected radius, \( a \). If the magnetic field is disordered and isotropic, Section 1 showed that this problem reduces to obtaining reliable measures for the RM dispersion, \( \sigma_{RM}(a) \), along with an estimate of the magnetic field coherence length, \( \Lambda(r) \), within the cluster, and assuming some value or range for \( \eta \). We emphasize that, in practice, one cannot directly determine \( \Lambda(r) \), which is a property of the 3D magnetic field, but must rely on estimating the projected RM coherence length, \( \Lambda(a) \) and modeling a connection between these lengths. Generally, if the magnetic field is isotropic, the magnetic field and RM power spectra are proportional (e.g., Enblin & Vogt 2003). Thus, it is possible, in principle, to derive the shape of the magnetic field power spectrum from the RM distribution (e.g., Murgia et al. 2004; Guidetti et al. 2008; Laing et al. 2008; Vacca et al. 2012; Govoni et al. 2017). In this work we do not attempt to determine any specific form of the magnetic power spectrum. Rather, we rely on the simple assumption, \( \Lambda(a) \approx \Lambda(r) = (3/4)\Lambda_{in}(r = a) \).

To summarize, the procedure we used in this work to estimate from observations the rms value of the magnetic field strength at the cluster center, \( \sigma_{B,0} \), is to:

1. determine the central density \( n_e,0 \) and \( \beta_c \) and \( r_c \) from X-ray observations;
2. estimate \( \sigma_{RM}(a) \) and fit it to a beta-law model in order to estimate the central \( \sigma_{RM,0} \);
3. determine an estimated value for the RM coherence length, \( \Lambda \), and make an assumption about whether it is a constant or a function of cluster radius;
4. assume a value for the magnetic field density scaling parameter, \( \gamma \);
5. derive from these inputs \( \sigma_{B,0} \), as outlined below.

To estimate the theoretical central RM dispersion, \( \sigma_{RM,0} \), for our cluster we average RM and \( RM^2 \) azimuthally with respect to the cluster center as functions of projected cluster radius, \( a \), and calculate \( \sigma_{RM}(a) = (\sqrt{RM^2})_a - (RM)_a^2 \). Assuming an isotropic magnetic field along with beta-law radial electron density and magnetic field models, we integrate through the full cluster to obtain from Equation (5) the actual, theoretical radial RM dispersion distribution. Following Dolag et al. (2001), with a fixed \( \Lambda = \Lambda_0 \) within the cluster and observational path lengths through the cluster, \( \ell \gg \Lambda_0 \), this would give

\[
\sigma_{RM}(a) = \frac{\sigma_{RM,0}}{\left[1 + \left(\frac{a}{r_c}\right)^{2}\right]^{1/2}},
\]

where the exponent \( \alpha_1 = \frac{3}{2}(1 + \eta)\beta_c - \frac{1}{2} \) and \( \sigma_{RM,0} \) is the central RM dispersion. Note also that for Equation (6) to be consistent with \( \sigma_{RM} \) values determined from observations that will be used in practice to estimate \( \sigma_B \), the observational distribution of RM sight lines must span projected scales significantly exceeding \( \Lambda_0 \).

Figure 3 displays the RM distributions in 867 (at 1 kpc resolution) obtained using a polarized screen with a 100% covering factor behind the cluster, viewed alternately along the three analysis grid axes. We can now use the central RM dispersion, \( \sigma_{RM,0} \), in Equation (6) to derive the central magnetic field dispersion, \( \sigma_{B,0} \), in Equation (5) for constant \( \Lambda = \Lambda_0 \) by the relation

\[
\sigma_{B,0} = \frac{\sqrt{3}}{812} \pi^{1/4} \sigma_{RM,0}\Lambda_0^{-1/2} r_c^{-1/2} \frac{\Gamma(1/2)}{\Gamma(\alpha_1)}. \]

The rms variation in magnetic field as a function of radius \( \sigma_{B,r} \) then follows from Equation (5).

We point out in the Appendix that \( L_{in}(r) \) (and therefore \( \Lambda(a) \)) actually increases with radius in our test cluster or, as a conveniently simple alternative expression, decreases with spherically averaged mean density (or for \( \Lambda(a) \), projected mean density). If we adopt the scaling relation suggested in Section 2, namely \( L_{in} \propto n_e^{-3/2} \), we can model the projected RM coherence scale, \( \Lambda(a) \), as

\[
\Lambda(a) = \Lambda_0 \left[1 + \left(\frac{a}{r_c}\right)^{2}\right]^{\frac{1}{2}\frac{3}{4}},
\]

where \( \Lambda_0 \), now refers to \( \Lambda(a = 0) \). Equation (6) then becomes slightly modified to

\[
\sigma_{RM}(a) = \frac{\sigma_{RM,0}}{\left[1 + \left(\frac{a}{r_c}\right)^{2}\right]^{\gamma}},
\]
where \( \alpha_2 = \frac{1}{2}(1 + \frac{3}{4}\eta)\beta_c - \frac{1}{4} = \alpha_1 - \frac{1}{8}\eta/\beta_c \). For characteristic parameters \( \beta_c \sim 3/4, \eta \sim 1/2 \), the difference between \( \alpha_2 \) and \( \alpha_1 \) is roughly 10%. \( \sigma_{\text{RM},0} \) can still be obtained using Equation (7), provided \( \Lambda_0 \) is found from Equation (8) and \( \alpha_2 \) replaces \( \alpha_1 \) in the \( \Gamma \) function arguments. The latter substitution leads to a small renormalization of the rhs of Equation (7) (\(<3\%\) for the above characteristic \( \beta_c, \eta \)). On the other hand, we note that, in the case of g676, \( \Lambda(a) \) varies by a factor of 3 between \( a = r_c \) and \( a = 10r_c \), so including the variation of \( \Lambda \) with radius in fitting for \( \sigma_{\text{RM},0} \) can significantly change the estimates of \( \sigma_{B,0} \).

We emphasize again that the above relationships assume both that the magnetic field itself is reasonably isotropic, so that \( \langle |B|^2 \rangle \ll \langle B \rangle^2 \), and that the RM values used in computing \( \sigma_{\text{RM}} \) broadly sample independent parts of the RM distribution. If the RM distribution is not well sampled on large enough scales (\( \gtrsim \Lambda \)), then even for an isotropic magnetic field the measured properties will lead to \( \langle (\text{RM})^2 \rangle \sim \langle \text{RM} \rangle^2 \). In particular, it is critical to include lags, \( |\Delta a| \), between sampling points that satisfy \( |\Delta a| \gtrsim \Lambda \). Murgia et al. (2004) show the dependence of \( \langle \text{RM} \rangle /\sigma_{\text{RM}} \) as a function of the utilized sampling space. In the case where the sampling region is too small, the estimated \( \sigma_{\text{RM}} = \sqrt{\langle (\text{RM})^2 \rangle} - \langle \text{RM} \rangle^2 \) will generally be reduced, so will lead to underestimates for \( \sigma_{B,0} \) through Equation (7).

As noted above, both \( \beta_c \) and \( r_c \) in such an analysis are established observationally from the X-ray surface brightness distribution. From synthetic X-ray observations of our simulated cluster we found in Section 2.2 values of \( \beta_c \approx 0.75 \) and \( r_c = 41 \text{ kpc} \). The remaining parameters in our RM models are \( \sigma_{\text{RM},0}, \eta, \) and \( \Lambda_0 \). We defer discussion of estimates for \( \Lambda_0 \) to the following subsection. The central RM dispersion, \( \sigma_{\text{RM},0} \) is obtained from observations by fitting an empirical RM distribution to Equation (6) (or alternatively Equation (9), if \( \Lambda \) is a function of radius).

The \( \eta \) parameter needed in Equation (10) comes from interpreting the fitted slope as \( \alpha_1 \) for a fixed RM coherence length, \( \Lambda(a) = \Lambda_0 \), or as \( \alpha_2 \), if the RM coherence length follows Equation (8). For reference we recall here that in Section 2 we obtained \( \eta \approx 0.5 \) from the 3D density and magnetic field distributions in our cluster. Using \( \beta_c = 3/4 \) and \( \eta = 1/2 \) as nominal parameters, the exponents in Equations (6), (8), and (9) become \( \alpha_1 = 23/16 \approx 1.44, \quad \alpha_2 = 83/64 \approx 1.30 \) respectively. Outside the cluster core, where \( a/r_c \gg 1 \) the theoretical RM dispersion would scale with projected radius as \( \sigma_{\text{RM}} \propto (a/r_c)^{3/2} \) for constant \( \Lambda \). Including the previously outlined density scaling for \( \Lambda \) would lead to \( \sigma_{\text{RM}} \propto a/r_c^{0.56} \) and \( \sigma_{\text{RM}} \propto a/r_c^{0.56} \). These are rather strong radial scalings, especially for the RM dispersion. One obvious consequence that we address in the next subsection is that measurements depending on \( \sigma_{\text{RM}} \) that include the cluster core will be dominated by contributions from the core.

In preparation for the discussion below, we also point out the sensitivity of solutions for \( \sigma_{B,0} \) to the \( \eta \) parameter. That sensitivity comes through its presence in the RM distribution shape parameters \( \alpha_{1,2} \) in Equations (6) and (9). In particular, as \( \eta \) decreases, both \( \alpha_{1,2} \) decrease, so the radial variation in \( \sigma_{\text{RM}} \) is reduced. Thus, given values of \( \sigma_{\text{RM}} \) at finite radii \( a, \) smaller \( \eta \) lead to smaller values for \( \sigma_{\text{RM},0} \) and thus smaller values for \( \sigma_{B,0} \).

Finally, for convenience, we rewrite Equation (7) in terms of approximate values for the g676 cluster. In particular,

\[
\sigma_{B,0} \approx 2 \mu G \left( \frac{\sigma_{\text{RM},0}}{1000 \text{ rad m}^{-2}} \right) \left( n_0/0.04 \text{ cm}^{-3} \right)^{1/2} \left( \Lambda_0/17 \text{ kpc} \right)^{1/2} \left( r_c/41 \text{ kpc} \right)^{1/2}.
\]

(10)
All our analysis fits below assume a central electron density, \( n_0 = 4 \times 10^{-2} \text{ cm}^{-3} \) and cluster core radius, \( r_c = 41 \text{ kpc} \).

4. RM Experiments Using Discrete Regions

In the previous section, to obtain a “theoretical” distribution for \( \sigma_{RM} \) we assumed that data were available from a well-sampled background polarized screen spanning the entire cluster. This is not achievable in practice, so we conducted a series of experiments restricting those conditions in ways that mimic common experience. In each case, we generated a finite set of discrete background regions (patches), each with their own values of projected radius, \( a_i \equiv (1/A_i) \int dA \) with \( A_i \) the area of an individual patch. For each patch, there was an associated “measured” \( \langle \text{RM} \rangle \) and \( \sigma_{\text{RM},i} \).

In Section 4.1 we will look at fully sampled background screens, using discrete regions in the form of annuli surrounding the cluster center. Although the cluster is then still fully sampled, the discretization of the RM information can introduce problems, as discussed below. In Section 4.2 we will use rectangular masks of various sizes placed at different locations along the line of sight, to approximate RM observations of individual radio galaxies embedded in the cluster.

In order to reduce “pixelation” issues associated with the 1 kpc\(^2\) discrete voxel size of the MHD simulation that produced the ICM being modeled, all subsequent 2D images presented below are averaged over 2 \( \times \) 2 pixels, for an effective observed resolution of 2 kpc. This is still sufficient to sample the RM structure, but does impose a fine scale cutoff, for example, to a structure function analysis we carried out on synthetic RM data (Figure 4).

4.1. Discrete Background Screens

Here we assume the observer can sample RMs on discrete background screens. Rather than the full sampling available in our “theoretical” RM scenario in Section 3, discrete patches do not necessarily allow full sampling of the RM distribution. The coverage depends on how the patches/screens are constructed. Information can be lost that may or may not influence the estimation of \( \sigma_{RM,0} \). Most important in this is the maximum length of the available lag vectors, \( |\Delta a| \), since without large lags, \( |\Delta a| \gg \Lambda \), all the RM values within a patch will be correlated. In that case, \( \langle \text{RM} \rangle^2 \sim \langle \text{RM} \rangle^2 \) within each patch, even if the underlying magnetic field is disordered and isotropic on scales beyond \( \Lambda_{\text{min}} \approx \Lambda \). Then \( \sigma_{RM} \) will be reduced from its true, physical value needed to estimate \( \sigma_B \) properly.

For these experiments we first create ideal background patches in the form of annuli around the cluster center, uniformly spaced in \( \log a_i \) with the minimum \( a_i \) placed somewhat inside the projected core radius. We explored the consequences of varying ring thicknesses, \( \delta a \), and found converged results so long as \( \delta a \geq 10 \text{ kpc} \). Thus, we limit our discussion to the illustrative \( \delta a = 10 \text{ kpc} \) case. Note that, due to the logarithmic spacing, there is some overlap in the annuli at the smallest radii, and gaps between the rings at the largest radii.

The two panels of Figure 5 summarize results of the \( \delta a = 10 \text{ kpc} \) annular rings viewed along the three primary grid axes. The top panel shows measured \( \sigma_{RM}(\delta a) \) values plus fits to the form in Equation (9) for variable \( \Lambda \) allowing \( \eta \) as a free parameter (see Table 1 for fitting summaries).

We checked whether the annuli have adequately sampled the largest scales of the RM fluctuations by examining the ratio \( \langle |\text{RM}| \rangle /\sigma_{RM} \), which in reality should be small. The computed values for each annulus are shown in the bottom panel of Figure 5. The median \( |\text{RM}| /\sigma_{RM} \approx 0.2 \). Even with the apparent scatter, the values are small enough that the associated \( \sigma_{RM} \) still represent reasonably appropriate measures for an isotropically disordered magnetic field behavior.

Table 1 provides an analysis summary from the annular background screen experiment. Results are given both for constant \( \Lambda = \Lambda_0 \) and radially varying \( \Lambda = \Lambda(\delta a) \). In each case fits are shown with fixed (preset) value of the magnetic field density scaling parameter, \( \eta = 0.5 \) and also with \( \eta \) as a free fitting parameter. These estimates for \( \Lambda_0 \) using annular background screens are consistent with what we found from the full background screen in the previous section, for the same assumptions. The results in Table 1 show that if we fix \( \eta \), then we get a range of only 10%–20% in the values derived for \( \sigma_{B,0} \) from the various projections. (Note, however, that the average for \( \sigma_{B,0} \) in the fixed \( \Lambda \) case is off by a factor of 1.6.) We see further that if \( \eta \) is allowed to vary in the fitting, estimates of \( \sigma_{B,0} \) will span a range \( \approx 3 \). Thus, with full RM coverage and independent knowledge of \( \eta \) it is possible to derive estimates of the magnetic field strength only within a range \( \approx 3 \).

The strong relationship that exists between the derived \( \sigma_{B,0} \) and \( \eta \) from these solutions is illustrated in Figure 6. This arises from the sensitivity of \( \sigma_{RM}(\delta a) \) to \( \eta \) through the \( \alpha_{1,2} \) shape parameter in Equation (6) or Equation (9). Smaller \( \eta \) leads to smaller \( \alpha_{1,2} \), which makes the form of \( \sigma_{RM}(\delta a) \) “stiffer.” Thus, for example, a fit to \( \sigma_{RM}(\delta a) \) in Equation (9) using \( \eta = 0.5 \) increases between \( a/r_c = 5 \) and \( a/r_c = 0 \) by a factor of 4 larger than it does using \( \eta = 0 \). An additional contribution to the \( \eta \) dependence of \( \sigma_{B,0} \) in Equation (10), when \( \Lambda \) varies with cluster radius, comes from the \( \eta \) dependence of \( \Lambda(\delta a) \) itself in Equation (8). This important degeneracy between the magnetic field model parameters has been extensively discussed in the literature (e.g., Murgia et al. 2004; Guidetti et al. 2008;
Bonafede et al. 2010; Vacca et al. 2012). In general, higher $\sigma_{R0}$ correlate with higher $\eta$.

Thus, even estimates using near-perfect RM coverage are not reliable to better than a range of 3 unless $\eta$ is assumed correctly within an unmeasurable range of values. Looking more closely at the variable-$\Lambda$ fits in Figure 6, we see that there is little change from the derived value of $\sigma_{R0}$ compared to the fixed $\Lambda$ case, although $\eta$ increases by $\sim 0.1$–0.2. If $\eta$ were to be held fixed in the fits, however, as we suggest below, then the assumption of a variable $\Lambda(a)$ can result in a factor of up to 2 decrease in $\sigma_{R0}$.

We performed an additional test of the robustness of the annular sampling procedure by placing the polarized screens at the mid-plane, instead of behind the cluster. We again viewed the screens along the three principal grid axes from both directions. When the computed $\sigma_{RM,0}$ values for each mid-plane experiment were renormalized by a factor $\sqrt{2}$ to adjust for the shorter path (factor 1/2) to the mid-plane, $\sigma_{R0}$ estimates were consistent with those for the above annular background screens within the statistical uncertainties. This good match was possible because paths through half the cluster still incorporate multiple RM coherent lengths and, just as important, we know the actual path to the screens. The rescaled mid-plane annular ring screen partition $\sigma_{R0}$ values are shown in Figure 6 with mid-plane identifications.

To examine whether these results were dependent on the specific use of annuli, we repeated the experiment by breaking up the background screen into a set of square patches. As long as the square patches were larger than $\Lambda_0$ ($\sim 20$ kpc), the average derived values of $\sigma_{R0}$ were consistent with the actual values, although with larger scatter between the different projections than the annular results, which spanned scales $\gg \Lambda_0$. When the square patches were smaller than $\Lambda_0$, the derived $\sigma_{R0}$ values were consistently too small (a factor of 2 for 12 kpc boxes), consistent with expectations. As noted earlier, the value of $|\langle RM \rangle|/\sigma_{RM}$, $\geq 1$ provided a good indicator that the larger scales of RM variations were not being adequately sampled by the 12 kpc boxes, since they are smaller than the actual magnetic field coherence length.

4.2. Embedded Masks

To approximate the type of information available from RM observations of cluster-embedded radio galaxies, we conducted a series of experiments using rectangular “masks” that provide a sparse but observationally realistic sampling of the cluster RM distribution. The first type of experiment simply sampled the RM distribution of rectangular masks placed randomly along and at right angles to the line of sight at different projected positions within the cluster. The second type of experiment looked at the RM distribution along the lines of sight to a central radio galaxy that was actually evolved by MHD simulation within this ICM. We defer to a subsequent study the RM consequences from physical displacement of adjacent ICM. Here we simply insert a 2D mask into the undisturbed ICM that matches the silhouette of the radio galaxy.

Our analysis of these embedded masks was similar to those discussed above. Statistics from synthetic RM observations were computed across individual masks to determine values for $\sigma_{RM}(a)$, with $a_i$ the mean projected cluster radius of each mask. The ensemble $\sigma_{RM}(a)$ distributions from the masks were then fit to Equation (9) to find $\sigma_{RM,0}$. They were also incorporated into a second-order RM structure function (Equation (12) in the Appendix) to find $\Lambda(a)$. Then $\Lambda(a)$ as represented in Equation (8) was used to estimate $\Lambda_0$. $\sigma_{RM}(a_i)$ errors were assigned for the purposes of finding $\sigma_{RM,0}$ including a statistical component $\sim 1/\sqrt{N_{\text{obs}}}$, where $N_{\text{obs}}$ represents the number of independent observing beams across a mask plus a constant 5 rad m$^{-2}$ component representing the uncertainty in the measurement of $\sigma_{RM}$. Although somewhat arbitrary, most observations would have uncertainties at least this large. These two error components were added in quadrature to get the total error used. Finally, Equation (10), renormalized by a factor $\sqrt{2}$.
with central electron density, \( \Lambda \) cluster translate to account for the assumed mid-plane location, was used to Figure 6. Empirically estimated \( \eta \) and \( \Lambda \) annuli 22 3.33 0.96 Fig. 6. Empirically estimated \( \sigma_{B,0} \) and \( \eta \) using \( \sigma_{RM} \) calculated from mid-cluster (five-point symbols) and background (three-point symbols) annular screens. Constant coherence scale \( \Lambda \) solutions are green, while density-dependent \( \Lambda \) solutions are red. The black circular point marks the actual, 3D \( g676 \) global magnetic field properties.

to account for the assumed mid-plane location, was used to translate \( \sigma_{RM,0} \) and \( \Lambda_0 \) into \( \sigma_{B,0} \).

### 4.2.1. Embedded Passive Masks

In this experiment, we randomly distributed multiple polarized, rectangular planar masks within our cluster, with the mask normals aligned to the line of sight. Individual rectangular masks had variable aspect ratios. Their side lengths ranged between 5 and 50 kpc representing the extent of typical rectangular masks had variable aspect ratios. Their side lengths \( r_x \approx 41 \) kpc. The fit to the actual 3D magnetic field in the cluster yields \( \sigma_{B0} = 1.9 \mu \text{G} \) and \( \eta = 0.5 \).

Table 1

| \( \langle \Lambda_0 \rangle \) (kpc) | \( \langle \sigma_{B,0} \rangle \) (\( \mu \text{G} \)) | range \( \langle \sigma_{B,0} \rangle \) | \( \langle \sigma_{B,0} \rangle \) (\( \mu \text{G} \)) | range \( \langle \sigma_{B,0} \rangle \) | \( \langle \eta \rangle \) | range(\( \eta \)) |
|---|---|---|---|---|---|---|
| Fixed \( \eta = 0.5 \) | \( \Lambda = \Lambda_0 \) | \( \Lambda = \Lambda_0 \) | \( \Lambda = \Lambda_0 \) | \( \Lambda = \Lambda_0 \) | \( \Lambda = \Lambda_0 \) | \( \Lambda = \Lambda_0 \) |
| \( \Lambda _{\Lambda (a)} \) annuli | 22 | 3.33 | 0.96–1.07 | 2.15 | 0.62–1.70 | 0.38 | 0.31–0.50 |
| \( \Lambda = \Lambda_0 \) | 16 | 1.96 | 0.94–1.06 | 2.30 | 0.61–1.72 | 0.51 | 0.41–0.67 |

**Note.** Averages and errors are calculated including all three viewing directions for a given experiment. The \( \eta \) parameter defines the scaling \( \sigma_B \propto n_e^{\eta} \). The assumed form for \( \Lambda(a) \) is given in Equation (8). The \( \sigma_{B,0} \) values on the left of the table assumed \( \eta = 0.5 \), while \( \sigma_{B,0} \) values on the right correspond to the associated, fitted \( \langle \eta \rangle \), with central electron density, \( n_0 = 4 \times 10^{-4} \text{ cm}^{-3} \) and ICM core radius, \( r_x = 41 \) kpc. The fit to the actual 3D magnetic field in the cluster yields \( \sigma_{B0} = 1.9 \mu \text{G} \) and \( \eta = 0.5 \).

![Figure 6](image_url)

**Figure 6.** Empirically estimated \( \sigma_{B,0} \) and \( \eta \) using \( \sigma_{RM} \) calculated from mid-cluster (five-point symbols) and background (three-point symbols) annular screens. Constant coherence scale \( \Lambda \) solutions are green, while density-dependent \( \Lambda \) solutions are red. The black circular point marks the actual, 3D \( g676 \) global magnetic field properties.

along all three grid principal axes from both directions (so a total of six views). To derive \( \sigma_{RM,0} \), we assumed the masks to be in the cluster mid-plane, although in fact they existed at random displacements with respect to that plane. Averaged over all views, the sources are centered around the mid-plane.

Figure 8 presents the \( \sigma_{RM} \) statistics for the eight random embedded mask RM distribution shown in Figure 7 along with curves representing two fits of Equation (9) to the data. The solid curve includes \( \eta \) as a fitting parameter, so that the shape parameter, \( \alpha_2 \), of \( \sigma_{RM}(a) \) is part of the fit. It is obvious in this case that the available data are simply inadequate to obtain a meaningful value for \( \eta \) as part of the fitting effort. The best-fit value, \( \eta = -0.37 \), is unphysical, while the accompanying value for \( \sigma_{RM,0} \approx 152 \text{ rad m}^{-2} \) is only about one-fifth the values of \( \approx 700 \text{ rad m}^{-2} \) obtained from the background screen. The dashed curve in Figure 8 represents a fit to the same random mask RM data, but with fixed \( \eta = 0.5 \), matching the physically determined value in this cluster. The associated estimate for \( \sigma_{RM,0} = 400 \text{ rad m}^{-2} \) is an improvement, but still only about 60% of the expected value.

We restrict our remaining analysis of the random mask experiments to the physically established \( \eta = 0.5 \) (along with a radially dependent \( \Lambda(a) \) represented in Equation (8)). Table 2 summarizes those experiments. It lists the \( \langle \Lambda_0 \rangle \) and \( \langle \sigma_{RM,0} \rangle \)
values obtained from six combined views of each embedded source ensemble and associated representative estimates for \( \langle \sigma_{B,0} \rangle \) along with ranges of these values for different views after trimming the two most extreme values from the calculation. This provides a very conservative estimate of the uncertainties. Several points are clear. All of the experiments produce \( \sigma_{\text{RM}} \) estimates (and consequently \( \sigma_{B,0} \) estimates) that are both highly uncertain and significantly reduced from correct values for this ICM. Similar to our full-screen experiments with square box partitions (Section 4.1), this results from the fact that these masks are too small to sample independent portions of the RM distribution adequately. Accordingly, the ratio \( [\text{RM}] / (\sigma_{\text{RM}}) \) is simply too large to allow reliable translation of \( \sigma_{\text{RM}} \) into \( \sigma_B \) through Equation (10).

The results in Table 2 from the eight-source experiments are no better than from the three-source experiments. This demonstrates that the undersampling within each patch of the larger-scale fluctuations, rather than their limited number, is the critical limitation. Additional scatter is introduced in this experiment because the masks are not actually at the assumed fixed mid-plane location.

4.2.2. Embedded Central AGN Jet-formed Cavity Masks

In another set of experiments we measured the foreground RMs in front of the cavities produced in numerical simulations where intermittent bipolar AGN jets were injected at the center of the cluster (run g676 in Mendygral et al. 2012). The use of intermittent jets, with a 50% duty cycle, resulted in distinct, “fat” cavities with axial ratios of \( \sim 2:1 \). In the experiments we describe here, we used those AGN simulations only to define the silhouette of the two cavities; the ICM for our purposes was exactly the same as in our previous experiments, with no central AGN. The cavity silhouettes thus define our “masks” within which to measure RMs. They evolved in time, with a greatest extent of approximately 120 kpc on each side. At each of two times, 79 and 92 Myr, we used the corresponding ICM, evolved without the presence of the radio galaxy; in this way our results reflect only the time-dependent behavior of the ICM, without the influence of the radio galaxy. Analysis of the RM structure with the actual jet-modified ICM will be discussed in a future work. The orientation of the jet axis was arbitrarily set to \( \sim 45^\circ \) from the z-axis of the analysis grid, and the mask planes included the major axis of each cavity and the joint normal to that axis and the line of sight. We observed the bipolar masks bidirectionally along all three analysis grid axes, and from two arbitrary directions normal to the jet axis (so eight views in total) at each time. Figure 9 shows RM distribution maps of the masks viewed along the z-axis at the two times mentioned above.

For the RM analysis, we partitioned each projected mask into annular sectors centered on the position of the AGN at the cluster center. We present the results utilizing the radially dependent \( \Lambda(a) \) in Equation (8) with \( \eta = 0.5 \). In the translation of \( \sigma_{\text{RM},0} \) to \( \sigma_{B,0} \) both sides of the AGN structure were assumed to be in the mid-plane of the cluster, so that a renormalization factor \( \sqrt{2} \) was applied to Equation (10). In actuality, for views down the grid axes, the distance along the line of sight varied across the mask, but in a way which the average line of sight was approximately equivalent to the mid-plane line of sight. Those details turn out not to be particularly important, compared to the differences in RMs from the different viewing angles.

Table 3 lists the values for \( \langle \sigma_{B,0} \rangle \) averaged over all eight views, as well as the range in \( \sigma_{B,0}/\langle \sigma_{B,0} \rangle \) for the individual views. Once again the two most extreme values of \( \sigma_{B,0} \) were excluded from each range. At both observation times, the mean value, \( \langle \sigma_{B,0} \rangle \), comes within roughly 10% of the associated physical \( \sigma_{B,0} \) of the cluster. This is a consequence of the masks covering spatial scales significantly larger then \( \Lambda_0 \). Although the averages are quite accurate, the individual derived values \( \sigma_{B,0} \) span more than a factor of two. This reflects the fact that the ICM is not, in reality, statistically homogeneous, even on scales larger than \( \Lambda_0 \), with distinct features such as the magnetic filaments evident in Figure 1. Different views therefore can yield quite different estimates of \( \sigma_{B,0} \).

5. Discussion

The most important finding from these studies is that the strength of the central magnetic field can only be determined to within a range of \( \sim 3 \) even in the ideal, practically unrealistic case of a fully sampled background of RMs. In any more realistic situations, with very partial sampling, estimates of the central field strength are much more uncertain, ranging up to a factor of 30 in our experiments. The critical factors leading to these uncertainties are as follows.

(a) ICM inhomogeneity. The richness of magnetic field structures, even in clusters that have not experienced a recent significant merger, is evident in Figure 1. In an ideal world, these real inhomogeneities would become statistically inconsequential on large enough scales. In practice, clusters will have inhomogeneities that span scales as large as the cluster...
Table 2
Statistics for Randomly Embedded Masks

|                | \(\langle \Lambda_0 \rangle\) (kpc) | \(\langle \sigma_{\text{RM},0} \rangle\) (rad m\(^{-2}\)) | trimmed \(\langle \sigma_{\text{RM},0} \rangle\) (rad m\(^{-2}\)) | \(\langle \sigma_{\Lambda,0} \rangle\) (\(\mu G\)) | trimmed \(\langle \sigma_{\Lambda,0} \rangle\) (\(\mu G\)) |
|----------------|---------------------------------|-----------------------------------|---------------------------------|-----------------|-----------------|
| Three sources  | 16                              | 526                               | 0.71–1.2                        | 1.5             | 0.73–1.2         |
| Eight sources  | 21                              | 462                               | 0.74–1.2                        | 1.2             | 0.75–1.2         |

Note. Averages include all six viewing directions for a given experiment. \(\sigma_{\text{B},0}\) is computed from \(\sigma_{\text{RM},0}\) using Equations (10) and (8), with \(\eta = 0.5\), \(\Lambda_0 = \langle \Lambda_0 \rangle\), central electron density, \(n_0 = 4 \times 10^{-2} \text{ cm}^{-3}\), ICX core radius, \(r_c = 41 \text{ kpc}\), and assuming the sources are all in the cluster mid-plane. Trimmed ranges exclude the two most extreme values, so would represent 67% probabilities if the distributions were Gaussian. Full ranges for these experiments and those reported in Table 3 were typically an order of magnitude or more.

Figure 9. RM maps of the embedded, passive bipolar AGN masks viewed down the grid z-axis at \(t = 79 \text{ Myr}\) (top) and \(t = 92 \text{ Myr}\) (bottom). Colorbar units are rad m\(^{-2}\). The AGN jets are active at the lower time, but inactive at the upper time. Details of the mask construction are in the text.

Table 3
Fits from Passive AGN Masks

| \(t\)          | \(\langle \Lambda_0 \rangle\) (kpc) | \(\langle \sigma_{\text{B},0} \rangle\) (\(\mu G\)) | \(\langle \sigma_{\text{B},0} \rangle\) (\(\mu G\)) |
|----------------|---------------------------------|---------------------------------|-----------------|
| \(t = 79 \text{ Myr}\) | 16                              | 1.95                            | 0.72–1.7         |
| \(t = 92 \text{ Myr}\) | 16                              | 1.80                            | 0.63–1.5         |

Note. Each \(\langle \sigma_{\text{B},0} \rangle\) is derived from RM data along eight viewing directions using the same fitting parameters as for values in Table 2. The physical 3D ICX magnetic field properties evolved to the observation time 79 Myr (92 Myr) were \(\sigma_{\text{B},0} = 1.79 \mu G\), \(\eta = 0.41\) (\(\sigma_{\text{B},0} = 1.60 \mu G\), \(\eta = 0.39\)). See Table 2 notes regarding trimmed values.

assumption that the physical parameters that characterize the distribution, \(\eta\) and \(\Lambda_0\), are well-determined. Even then, any individual experiment will typically yield a value within a range of no less than a factor \(\approx 3\).

(b) Magnetic field, density scaling. The derived \(\sigma_{\text{B},0}\) is sensitive to the magnetic field density scaling parameter, \(\eta\) (\(B \propto n_0^\eta\)). While MHD cluster formation simulations show that there can be a characteristic scaling, \(\sigma_0 \propto n_0^\beta\), in practice, \(\eta\) cannot be determined accurately from observations. Fitting for \(\eta\) depends on subtle shape differences in the profile of \(\sigma_{\text{RM}}(\alpha)\), as described in Equations (6) and (9). Even for fully sampled background screens, allowing \(\eta\) to vary in fitting leads to a large range in derived \(\sigma_{\text{B},0}\) as seen in Figure 6. The problem is greatly exacerbated for sparse sampling similar to that available from actual observations. In that case, the shape of \(\sigma_{\text{RM}}(\alpha)\) is poorly constrained, and the unknown position along the line of sight of a source (see point \(d\)) adds a further large complication. Since many simulations indicate that \(\eta \approx 0.5\), our experiments suggest that assuming that value, or something similar, may yield the most accurate estimates of \(\sigma_{\text{B},0}\).

(c) RM coherence length. Using Equations (2) and 10, the magnetic field estimates scale as the assumed value of \(\sqrt{\Lambda}\). We estimated \(\Lambda\) from the RM structure function, identifying the scale, \(\Lambda_{SF}\), at which the slope of the structure function first reached zero from small scales as \(\Lambda\). In the ideal case of the mid-plane screen experiments, we found a range of \(\sim 25\%\) in \(\Lambda\), which therefore makes only a minor contribution to the error budget. Randomly placed embedded source experiments have more limited sampling, and the range of derived \(\Lambda_{SF}\) and estimated central cluster value \(\Lambda_0\) is 10–30 kpc. But since \(\sigma_{\text{B},0}\) depends only on \(\sqrt{\Lambda}\), this has relatively modest impact compared to \(a\) or \(b\). We note that while this procedure was successful in the current experiment, one can envisage other magnetic field configurations, such as very flat magnetic power spectra, where only the minimum scale is accessible, and a different way of utilizing the RM structure function would be needed (e.g., Laing et al. 2008).

(d) Line-of-sight uncertainties. Since the positions of individual, embedded sources along the line of sight are unknown and unknowable (except for cluster center sources), the application of Equation (2) must be based on some assumption. In our derivations, we assumed that all RMs were integrated through the entire cluster or, for our mid-plane experiments, half the cluster. For a large enough sample of sources with RMs, the mid-plane would be a good approximation. However, as each individual source contributes to the measurements \(\eta\) and \(\Lambda(\alpha)\), the assumption of a mid-plane location can lead to large uncertainties and even apparent nonphysical behaviors (e.g., a derived negative fit for \(\eta\)). In the best case, where we fix \(\eta\) and the \(\Lambda(\alpha)\) behavior, Table 2 shows that...

...itself due to the ongoing intermittent accretion along filaments, etc. We studied the influence of these inhomogeneities on \(\sigma_{\text{B},0}\) determination by viewing the cluster from different directions under a range of circumstances. These included the ideal of a fully sampled polarized background screen spanning the cluster, and the more realistic case of sparse sampling. When averaging over multiple, individual experiments in a class, an accurate value of \(\sigma_{\text{B},0}\) can sometimes be recovered (e.g., 10%–20%), although this commonly involves the unrealistic

...
a range of 3 in the fit for the trimmed \( \sigma_{B,0} \) can be achieved. The variations contain contributions from both cluster inhomogeneities and the assumption that the sources were all at mid-plane, when actually they were sampling only the path length appropriate to their 3D position in the cluster and the particular projected view.

5.1. Comparison to Current Cluster Analyses

One of the best data sets for cluster RM analysis comes from Hydra A, the central radio source in a cool core cluster, mapped by Taylor & Perley (1993), and further analyzed by Vogt et al. (2005) and Vogt & Enßlin (2005). The latter work presents a Bayesian analysis of the northern lobe of the source, which extends \( \sim 40 \text{kpc} \) from the center, comparable to the inner core radius of the X-ray distribution. They derive the magnetic field power spectrum, finding a portion with slope \( \sim \frac{5}{3} \), a characteristic RM coherence length of \( \sim 5 \text{kpc} \), and a central field strength of \( 7 \pm 2 \mu \text{G} \). Although they describe these errors as reflecting the systematic uncertainties in the distance along the line of sight (reflected in the inclination of the radio source) and in the range in acceptable values of \( \eta \) from 0.1 to 0.8, results are only reported holding one or the other of these parameters fixed. Their equivalent \( \lambda_0 \) is held fixed, and since the full spatial resolution is only utilized over \( \sim 20 \text{kpc} \), because of signal-to-noise concerns, there is no information available on any radial variation in the coherence length. However, the most important shortcoming of this analysis is that it applies only to the northern lobe. The southern lobe was explicitly excluded from this analysis because it has different RM properties, with a much stronger RM power spectrum, leading to RM values reaching \( \sim 12,000 \text{rad m}^{-2} \), with associated strong depolarization. A later analysis of Hydra A, including both lobes and assuming that the lobes had created cavities in the surrounding medium, was performed by Laing et al. (2008). The latter authors find (a) less gas, (b) a different preferred value for \( \eta \), (c) a magnetic field autocorrelation length twice as high, and (d) a central magnetic field strength 2.5\times higher than in Vogt & Enßlin (2005). The characterization of the central magnetic field strength with high accuracy in Vogt & Enßlin (2005) therefore does not reflect the uncertainties that are present, as shown in this paper, in the modeling of the magnetic field in this cluster.

Vacca et al. (2012) use a different type of analysis to measure the magnetic power spectrum in A2199. Their approach is also Bayesian, and involves comparing a range of simulations with the observed RM distributions of 3C 338, a central radio galaxy with RM measurements extending over \( \lesssim 40 \text{kpc} \). The maximum fluctuation scale is only approximately characterized at \( 35 \pm 28 \text{kpc} \), which is expected given the limited RM sampling available. They explicitly show the sensitivity of the central magnetic field strength to the maximum scale, as we also discuss here. Their final estimates of \( \eta \) are from 0.4 to 1.4, and a central magnetic field strength of \( 2.7-20.7 \). This large range is consistent with the results of our experiments.

A study of A194 using RM observations from the Sardinia Radio Telescope and the VLA is presented by Govoni et al. (2017). They use a Bayesian-type analysis based on 3D simulations assuming a power-law magnetic field spectrum, while utilizing both the RM distribution and fractional polarization information. They find \( \eta = 1.1 \pm 0.2 \), a maximum scale of \( 64 \pm 24 \text{kpc} \), and a central magnetic field of \( 1.5 \pm 0.2 \mu \text{G} \). Their polarization information extends over \( 260 \text{kpc} \), so we would expect the variations in the field to be very well sampled. However, the observed RM distribution has a mean value of \( 15.2 \text{rad m}^{-2} \) with \( \sigma_{\text{RM}} = 14.4 \text{rad m}^{-2} \). In this case, \( \langle |\text{RM}| \rangle / \sigma_{\text{RM}} \) is not small, as required by the analytical modeling described above, and only a Monte Carlo-type analysis such as performed by Govoni et al. (2017) can be used. In A194, most of the polarized emission comes from 3C 40B, associated with a luminous galaxy in the cluster core, so the position along the line of sight is reasonably well constrained. Remaining uncertainties, such as discussed in the Govoni et al. (2017) analysis, include possible non-power-law magnetic field distributions and/or cavities produced by the radio galaxy lobes (e.g., Laing et al. 2008).

In summary, we find that existing derivations of cluster fields from RM observations reflect the same kinds of fundamental uncertainties that underlie our analysis. In some published observational analyses, the derived central field strengths are reported with uncertainties consistent with our findings, while in other cases, the extent of the uncertainties is not adequately addressed, and may be seriously underestimated.

6. Conclusions

The derivations of central magnetic field strengths in clusters of galaxies are subject to a number of important uncertainties. From a physical standpoint, cluster magnetic fields are likely to be statistically inhomogeneous, being influenced by disturbances due to the continuing growth and evolution of the cluster, even when no major cluster encounters have occurred recently. Then, even in this “best case” situation, a specific “central field value” may not be appropriate or adequate to address key science questions associated with the ICM. This needs to be evaluated on a case-by-case basis.

Using as a test base the known magnetic field distribution in a relatively quiescent cluster formed during cosmological simulations, we have found that magnetic field determinations, even in the perfect but unrealizable case of a fully sampled RM distribution, are limited to a range of \( \sim 3 \). In the case of actual observations of real clusters, several practical, irreducible limitations can cause estimates of the central field strengths to span at least an order of magnitude. These limitations include the unknown scaling of magnetic field strength with ICM electron density, RM sampling limitations that lead to underestimates of the magnitude of RM fluctuations, and possible but unknown variations of magnetic field structure scales with distance from the cluster center. These limitations in obtaining fits to these model parameters and the desired magnetic field strengths are amplified by uncertainties such as the unknown positions along the line of sight of polarized sources used to determine RMs. It will thus always be necessary to introduce both physical and sampling assumptions into any derivations of cluster fields. It is important that all such assumptions are clearly stated, and that the uncertainties in those assumptions be reflected in the final derived values.

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Appendix

In this Appendix, we outline some issues surrounding the use of the RM coherence length (\(\Lambda_0\) and \(\Lambda(a)\)) needed for the derivation of \(\sigma_{B,a}\). For a magnetic field that is isotropically disordered through turbulence, there will be some effective RM coherence scale along a line of sight of length \(\ell\), where \(\Lambda \ll \ell\). It has been shown that for an isotropic, turbulent field distribution the length \(\Lambda\) needed in Equation (2) can be expressed in terms of the magnetic field autocorrelation length, \(\Lambda_B\), as \(\Lambda = (3/2) \Lambda_B\) (Enßlin & Vogt 2003) or alternatively in terms of the so-called integral length Cho & Ryu (2009) as

\[
\Lambda = \frac{3}{2} L_B = \frac{3}{4} L_{\text{int}} = \frac{3}{4} \int P_B(k) \langle k \rangle^{-1} \, dk.
\]

(11)

These relationships are related, and depend only on isotropy of the magnetic field and a well-defined magnetic field power spectrum. In Equation (11) \(P_B(k)\) is the 1D power spectrum of the 3D, isotropic magnetic field, with the wave number, \(k\), defined without the usual 2\(\pi\) factor.

The numerical factors 3/4 and 3/2 connecting \(\Lambda\) to \(L_{\text{int}}\) and \(L_B\) reflect somewhat different weights that turbulent magnetic field fluctuations contribute to the specific statistical length measures. Indeed, a number of authors have emphasized in realistic models of RM properties associated with disordered magnetic fields that while \(\Lambda \propto L_B\) (or similarly \(\Lambda \propto L_{\text{int}}\)), \(\Lambda \gg L_B\) (or similarly \(\Lambda \gg L_{\text{int}}\)).

In practice, \(\Lambda\) must be estimated from the RM distribution’s 2D characteristic coherence scale. The observed RM distribution is a projected rather than a local measure. Although the projected RM \(\Lambda\) is related to the 3D magnetic field measure, \(\Lambda(r)\), it is not generally equivalent. As pointed out above and emphasized by Enßlin & Vogt (2003), in a homogeneous isotropic magnetic field setting \(\Lambda\) can be directly related to the magnetic field correlation length, \(L_B\), by the RM distribution power spectrum (see also Vogt & Enßlin 2003), which can, in principle, be established from observations. However, obtaining both \(L_B\) and \(\Lambda\) in this way is a complex and difficult procedure in practice, especially with restricted RM sampling (Vogt & Enßlin 2005). So a more common strategy has been to assume for modeling purposes a magnetic field power spectrum, usually a power law with inner and outer scales (e.g., Murgia et al. 2004; Huarte-Espinoza et al. 2011). On the other hand, MHD simulations reveal that magnetic field distributions evolved through the turbulent dynamo from a weak seed field are poorly represented by power-law power spectra (e.g., Cho & Ryu 2009; Porter et al. 2015). That is also apparently the case for the magnetic fields that evolve in clusters formed in cosmological simulations (e.g., Xu et al. 2011; Wittor et al. 2016), which certainly is the case for the simulated cluster we use in these experiments (Section 2, Figure 10). Consequently, we apply in our work here a relatively simpler approach to estimating \(\sigma_B\) that does not require assumptions about or computation of power spectra. In particular, as discussed immediately below, we estimate a projected RM coherence length from the second-order RM structure function, \(\Lambda_{\text{SF}}\), and associate that length with \(\Lambda\) in

![Figure 10. Power spectrum of the kinetic energy (red) and magnetic energy (blue) of the g676 ICM within the (1 Mpc)^3 analysis box. The solid black line represents a Kolmogorov spectrum, \(P(k) \propto k^{-5/3}\). The broad magnetic power spectrum turnover, \(k \sim 8–20\) Mpc^\(-1\), represents scales \(k^{-1} \sim 50–120\) kpc.](image-url)
sufficiently large areas (see Figure 5), so \( S(|\Delta a|) \) should approach \( 2 \times \sigma_{RM} \) for large \(|\Delta a|\). As detailed below, we designate the length \(|\Delta a|\) on which \( S(|\Delta a|) \) reaches its maximum as \( \Lambda_{SF} \), which we then use as our estimate for \( \Lambda \).

We should keep in mind, of course, that while our simulated ICM provides information on the necessary scales, actual cluster RM observations may not provide sufficient sampling, and those approximations may not be meaningful. However, attempts have been made to directly fit the observed structure functions to high-quality RM images using the Hankel transform (e.g., Laing et al. 2008).

The behavior of \( S(|\Delta a|) \) computed using background screens spanning our full (1 Mpc)\(^3\) analysis grid in the g676 cluster is shown in Figure 4. For each projection, \( S(|\Delta a|) \) increases from small scales, then plateaus near values \( S(|\Delta a|) \approx 5000-10,000 \text{ rad}^2 \text{ m}^{-4} \), corresponding to (cluster-wide) \( \sigma_{RM} \approx 50-70 \text{ rad}^2 \text{ m}^{-2} \). Note that this value is an order of magnitude below \( \sigma_{RM,0} \), such as the values in Figure 5, because it represents an average over the entire (1 Mpc)\(^3\) grid, and is dominated by the very weak fields far from the cluster center. Nonetheless, the shape of \( S(|\Delta a|) \) is a good representation of the magnetic field structure, so can be used to estimate the coherence length.

We define \( |\Delta a|_{\Lambda} \) as the smallest \( |\Delta a| \) for which \( d(S(|\Delta a|)) / d(|\Delta a|) \approx 0 \), and justify this now as our estimated \( \Lambda \). Applied to the full (1 Mpc)\(^3\) analysis box (Figure 4) from an average of the three principal axis orientations, we would obtain in this way \( |\Delta a|_{\Lambda} \approx 25 \text{ kpc} \). A similar analysis of the \( S(|\Delta a|) \) distributions computed over (100 kpc)\(^2\) areas centered on the cluster core results in a coherence scale, \( |\Delta a|_{\Lambda} \approx 20 \text{ kpc} \).

Equation (10) specifically requires an estimate of the central coherence length, \( \Lambda_0 \). So it is necessary to establish the relationship between observationally derived \( \Lambda \) values over a selected projected cluster area and the central \( \Lambda_0 \). If we assume a constant \( \Lambda \) in evaluating the RM structure function data using fully sampled regions spanning the entire (1 Mpc)\(^3\) analysis box from cluster g676, we would obtain \( \Lambda_0 = \Lambda \approx 25 \text{ kpc} \). The fully sampled core alone would yield \( \Lambda_0 = \Lambda \approx 20 \text{ kpc} \). These two \( \Lambda_0 \) estimates are similar enough that, taken at face value, they would lead to estimates for \( \sigma_B \) from Equation (6) differing by only roughly 10%. However, that they do differ points back to the previously discussed issue that the magnetic field integral length is not a constant value, but increases with distance from cluster center (that is, \( L_{int} = L_{int}(r) \), with \( dL_{int} / dr > 0 \)). As pointed out previously, this can have a substantial impact on the derivation of valid estimates for \( \sigma_{B,0} \), depending on how the estimates are made.

Specifically, our analysis in Section 2 of the 3D magnetic field properties of g676 established that \( \Lambda = (3/4) L_{int} \approx 17 \text{ kpc} \) in the central 100 kpc\(^3\) box (so \( r \approx 50 \text{ kpc} \)). Since this (50 kpc)\(^3\) volume is roughly the size of the core, we, therefore, obtain directly from the field itself (not via RM measurements) the estimate \( \Lambda_0 \approx 17 \text{ kpc} \). On the other hand, we similarly found \( \Lambda = (3/4) L_{int} \approx 40 \text{ kpc} \) within the full (1 Mpc)\(^3\) box (so \( r \approx 500 \text{ kpc} \)). That is an increase from the core to the outskirts of the cluster of a factor \( \sim 2.4 \), representing a difference of \( \sim 50\% \) in the translation between \( \sigma_{B,0} \) and \( \sigma_{RM,0} \) in Equation (10). This enhanced impact comes from the fact that the RM values depend on both the magnetic field distribution and the electron density distribution. The latter has a steep radial dependence (\( r^{-5/4} \) at large radii), which leads to a strong central bias in RM contributions. Thus, substantially better estimates for \( \Lambda_0 \) should come from application of the nonuniform model represented in Equation (9) than from fixed values for \( \Lambda \) represented in Equation (6).

A.2. Some Power Spectrum Issues

As emphasized in the discussion after Equation (2), the RM dispersion, \( \sigma_{RM} \), for a medium with an isotropically disordered magnetic field scales over long paths as the square root of the number of independent magnetic structures along the path, so depends inversely on the square root of \( \Lambda \), which scales with the magnetic field integral and correlation lengths, \( L_{int} \) and \( L_B \), according to Equation (11). As emphasized in the main body of this paper, \( \Lambda \) can only be estimated observationally from measurements of the RM distribution. We can, however, in our current experiments compare those observational estimates to the actual \( L_{int} \) of the 3D magnetic field. Figure 10 shows the 1D magnetic field power spectrum, \( P_B(k) \), inside our (1 Mpc)\(^3\) analysis box (blue line). For comparison we also show the g676 ICM kinetic energy 1D power spectrum in the same volume (red line), along with a \( k^{-5/3} \) line (black), representing the slope of a Kolmogorov spectrum. Here, \( k = 1 \) corresponds to a length of 1 Mpc.

To facilitate our discussions below we note some additional relevant properties of these power spectra. The kinetic energy power spectrum in Figure 10 is very roughly consistent with the \( k^{-5/3} \) form. However, as already pointed out, the broadly peaked magnetic power spectrum is not at all well-represented by a power law.

From the form of Equation (11) it is obvious that \( L_{int} \sim k^{-1} \). Over the full (1 Mpc)\(^3\) box, we therefore see from Figure 10 that \( L_{int} \sim 50 \text{ kpc} \). Indeed, applying the power spectrum in Figure 10 we obtain numerically \( L_{int} = 54 \text{ kpc} \). On the other hand, as mentioned previously, we generally expect \( L_{int} \) to vary with the scales of ICM structures. For instance, flux freezing during compression of a disordered magnetic field would lead to \( L_{int} \sim \ell \propto n_e^{-1/3} \propto B^{-1/2} \). The exact scaling would depend on dynamical circumstances analogous to the scaling for \( B \) itself. Nonetheless, we would expect from dynamical “similarity” arguments that \( L_{int} \) would usually increase toward the cluster outskirts (e.g., Shi et al. 2018). As discussed in the main text, we express this behavior in terms of a density scaling as a convenient proxy. We might guess a generalization of our flux-freezing example to take a form something like \( L_{int} \propto n_e^{-n/2} \). Since the scaling is weak, the detailed form should not be critical, and this has the modeling advantage of not adding free parameters to data fits. To test this expectation in the g676 ICM magnetic field we computed magnetic power spectra in nine additional cluster-centered volumes spanning scales ranging from 100 to 900 kpc. Indeed we found that \( L_{int} \) increased smoothly from \( L_{int} \approx 23 \text{ kpc} \) in the smallest box (essentially including only the cluster core) to the aforementioned \( L_{int} \approx 54 \text{ kpc} \) in the (1 Mpc)\(^3\) box. The values fit a tight correlation \( L_{int} \propto (n_e)^{-1/4} \), as well (not shown), so consistent with our guess that \( L_{int} \propto n_e^{-n/2} \). We find that the results of our RM analysis are substantially improved if we allow \( L_{int} \) to increase with distance from the cluster center using this scaling compared to keeping \( L_{int} \) fixed throughout the cluster. Such a radial dependence would be suggested by the change in the derived slope of the power-law spectrum in
A2255 as a function of distance from the cluster center (Govoni et al. 2006).

ORCID iDs

L. Rudnick https://orcid.org/0000-0001-5636-7213
P. J. Mendygral https://orcid.org/0000-0002-1638-8930

References

Ascasibar, Y., & Markevitch, M. 2006, ApJ, 650, 102
Bonafede, A., Feretti, L., Murgia, M., et al. 2010, A&A, 513, A30
Brunetti, G., & Jones, T. W. 2014, JMPD, 23, 1430007
Carilli, C. L., & Taylor, G. B. 2002, ARA&A, 40, 319
Cavaliere, A., & Fusco-Femiano, R. 1976, A&A, 49, 137
Cho, J., & Ryu, D. 2009, ApJL, 705, L90
Clarke, T. E., Kronberg, P. P., & Börhinger, H. 1999, in Diffuse Thermal and Relativistic Plasma in Galaxy Clusters, ed. H. Böhringer, L. Feretti, & P. Schuecker (Garching: MPE), 82
Dolag, K., Borgani, S., Murante, G., & Springel, V. 2009, MNRAS, 399, 497
Dolag, K., Grassi, D., Springel, V., & Tkachev, I. 2005, JCAP, 1, 9
Dolag, K., Schindler, S., Govoni, F., & Feretti, L. 2001, A&A, 378, 777D
Donnert, J., Dolag, K., & Lesch, H. 2009, MNRAS, 392, 1008
Eilek, J. A., & Owen, F. N. 2002, ApJ, 567, 202
Enßlin, T. A., & Vogt, C. 2003, A&A, 401, 835
Enßlin, T. A., Vogt, C., Clarke, T. E., & Taylor, G. B. 2003, ApJ, 597, 870
Felten, J. E. 1996, in ASP Conf. Ser. 88, Clusters, Lensing, and the Future of the Universe, ed. V. Trimble & A. Reisenegger (San Francisco, CA: ASP), 271
Feretti, L., Dallacasa, D., Giovannini, G., & Tagliani, A. 1995, A&A, 302, 680
Feretti, L., Dallacasa, D., Govoni, F., et al. 1999, A&A, 344, 472
Feretti, L., Giovannini, G., Govoni, F., & Murgia, M. 2012, A&ARv, 20, 54
Govoni, F., Murgia, M., Feretti, L., et al. 2006, A&A, 460, 425
Govoni, F., Murgia, M., Vacca, V., et al. 2017, A&A, 603, 122
Guidetti, D., Laing, R. A., Bridle, A. H., Parma, P., & Gregorini, L. 2011, MNRAS, 413, 2525
Guidetti, D., Laing, R. A., Croston, J. H., Bridle, A. H., & Parma, P. 2012, MNRAS, 423, 1355
Guidetti, D., Laing, R. A., Murgia, M., et al. 2010, A&A, 514, 50
Guidetti, D., Murgia, M., Govoni, F., et al. 2008, A&A, 483, 699
Huarte-Espinosa, M., Krause, M., & Alexander, P. 2011, MNRAS, 418, 162
Kunz, M., Schekochihin, A., Cowley, S., Binney, J., & Sanders, J. 2011, MNRAS, 410, 2446
Laing, R. A., Bridle, A. H., Parma, P., & Murgia, M. 2008, MNRAS, 391, 521
Laing, R. R. A., Bridle, A. H., Parma, P., & Murgia, M. 2008, MNRAS, 391, 521
Lawler, J. M., & Dennison, B. 1982, ApJ, 252, 81
Mendygral, P. J., Jones, T. W., & Dolag, K. 2012, ApJ, 750, 166
Miniati, F., & Beresnyak, A. 2015, Natur, 523, 59
Murgia, M., Govoni, F., Feretti, L., et al. 2004, A&A, 424, 429
Newman, W. I., Newman, A. L., & Rephaeli, Y. 2002, ApJ, 575, 755
Parrish, J. J., McCourt, M., Quataert, E., & Sharma, P. 2012, MNRAS, 422, 704
Porter, D. H., Jones, T. W., & Ryu, D. 2015, ApJ, 810, 93
Rudnick, L., & Blundell, K. M. 2003, ApJ, 588, 143
Ryu, D., Kang, H., Cho, J., & Das, S. 2008, Sci, 320, 909
Sanders, J. S., Fabian, A. C., & Smith, R. K. 2011, MNRAS, 410, 1797
Schuecker, P., Finoguenov, A., Miniati, F., Börhinger, H., & Briel, U. G. 2004, A&A, 426, 387
Shi, X., Nagai, D., & Lau, E. T. 2018, MNRAS, 481, 1075
Stasyszyn, F. A., Dolag, K., & Beck, A. M. 2013, MNRAS, 428, 138
Taylor, G. B., Gugliucci, N. E., Fabian, A. C., et al. 2006, MNRAS, 368, 1500
Taylor, G. B., & Perley, R. A. 1993, ApJ, 416, 554
Tribble, P. C. 1991, MNRAS, 250, 726
Vacca, V., Murgia, M., Govoni, F., et al. 2012, A&A, 540, A38
Vazza, F., Brunetti, G., & Brüggen, M. 2018, MNRAS, 474, 1672
Vazza, F., Jones, T. W., Brüggen, M., et al. 2017, MNRAS, 464, 210
Vikhlinin, A., Kravtsov, A., Forman, W., et al. 2006, ApJ, 640, 691
Vogt, C., Dolag, K., & Enßlin, T. 2005, MNRAS, 358, 732
Vogt, C., & Enßlin, T. A. 2003, A&A, 412, 373
Vogt, C., & Enßlin, T. A. 2005, A&A, 434, 67
Wittor, D., Vazza, F., & Brüggen, M. 2016, MNRAS, 464, 4448
Xu, H., Govoni, F., Murgia, M., et al. 2012, ApJ, 759, 40
Xu, H., Li, J., Collins, D. C., Li, S., & Norman, M. L. 2011, ApJ, 739, 77
Zhuravleva, I., Churazov, E., Sazonov, S., Sunyaev, R., & Dolag, K. 2011, AstL, 37, 141
Zuhone, J. A., Markevitch, M., & Lee, D. 2010, ApJ, 743, 16