

# Analytical model for non-thermal pressure in galaxy clusters II: Comparison with cosmological hydrodynamics simulation

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19 August 2014

## ABSTRACT

Turbulent gas motion inside galaxy clusters provides a non-negligible non-thermal pressure support to the intracluster gas. If not corrected, it leads to a systematic bias in the estimation of cluster masses from X-ray and Sunyaev-Zel’dovich (SZ) observations assuming hydrostatic equilibrium, and affects interpretation of measurements of the SZ power spectrum and observations of cluster outskirts from ongoing and upcoming large cluster surveys. Recently, Shi & Komatsu (2014) developed an analytical model for predicting the radius, mass, and redshift dependence of the non-thermal pressure contributed by the kinetic random motions of intracluster gas sourced by the cluster mass growth. In this paper, we compare the predictions of this analytical model to a state-of-the-art cosmological hydrodynamics simulation. As different mass growth histories result in different non-thermal pressure, we perform the comparison on 65 simulated galaxy clusters on a cluster-by-cluster basis. We find an excellent agreement between the modeled and simulated non-thermal pressure profiles. Our results open up the possibility of using the analytical model to correct the systematic bias in the mass estimation of galaxy clusters. We also discuss tests of the physical picture underlying the evolution of intracluster turbulence, as well as a way to further improve the analytical modeling, which may help achieve a unified understanding of non-thermal phenomena in galaxy clusters.

**Key words:** galaxies: clusters: general – galaxies: clusters: intracluster medium – cosmology: observations – methods: analytical – methods: numerical

## 1 INTRODUCTION

Precise mass determinations of galaxy clusters are crucial for their cosmological applications. We usually assume hydrostatic equilibrium between the pressure gradient and the gravitational force on the intracluster gas when determining masses from X-ray and SZ observations. These observations, however, typically measure only the thermal pressure of the gas. Non-thermal pressure, if neglected, introduces a bias in the hydrostatic mass estimation (HSE mass bias). This would, in turn, bias the cosmological constraint from the cluster mass function and the SZ power spectrum, and affect the interpretation of observations of cluster outskirts from ongoing and upcoming large cluster surveys.

Observationally, the HSE mass bias manifests itself as a systematic difference between the X-ray (or SZ) derived mass and the lensing mass of up to 30% (Allen 1998; Mahdavi et al.

2008; Richard et al. 2010; Zhang et al. 2010; von der Linden et al. 2014, but see also non-detections, e.g., Israel et al. 2014). Hydrodynamics numerical simulations of intracluster gas using both grid-based (Iapichino & Niemeyer 2008; Iapichino et al. 2011; Vazza et al. 2009; Lau, Kravtsov & Nagai 2009; Nelson et al. 2014; Nelson, Lau & Nagai 2014) and particle-based (Dolag et al. 2005; Vazza et al. 2006; Battaglia et al. 2012) methods have found that turbulence<sup>1</sup> inside clusters generated in the structure formation process contributes significantly to the non-thermal pressure. This alone leads to a HSE mass bias comparable to that found from observations (Rasia et al. 2006, 2012; Nagai, Vikhlinin & Kravtsov 2007; Piffaretti & Valdarnini 2008; Lau, Kravtsov & Nagai 2009; Meneghetti et al. 2010; Nelson et al. 2012). In addition to the structure formation process, turbulent gas motions can be gen-

<sup>1</sup> Following Shi & Komatsu (2014), we refer to the non-thermal random motion in the intracluster gas as ‘turbulence’ or ‘turbulent gas motions’ without distinguishing it from isotropic bulk motions.

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erated in the cluster outskirts by the magnetothermal instability (Parrish et al. 2012; McCourt, Quataert & Parrish 2013), and in the cluster core by energy injection from black holes and stars. Magnetic fields and cosmic rays can also potentially contribute to the non-thermal pressure. We refer to, e.g., Shi & Komatsu (2014) for a discussion of these sources, and focus, in the following, on the pressure support from the turbulent motion in the intracluster gas.

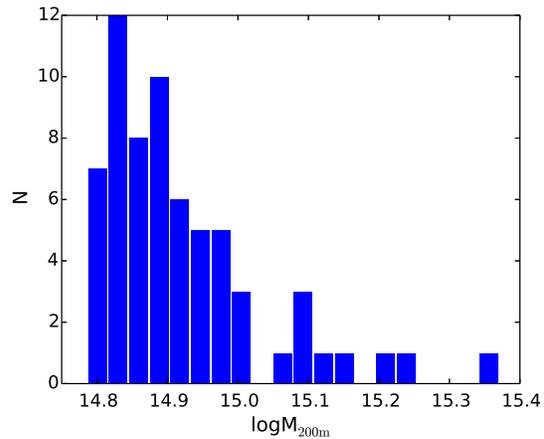
Several observations have provided indirect evidence for the intracluster gas motion: measurements of the magnetic field fluctuations in diffuse cluster radio sources (Murgia et al. 2004; Vogt & Enßlin 2005; Bonafede et al. 2010; Vacca et al. 2010), X-ray surface brightness fluctuations or pressure fluctuations inferred from X-ray maps (Schuecker et al. 2004; Churazov et al. 2012; Simionescu et al. 2012), and the non-detection of resonant scattering effects in the X-ray spectra (Churazov et al. 2004). Future observations of the X-ray emission lines are considered as the most promising method to measure turbulence velocities directly (Sunyaev, Norman & Bryan 2003; Zhuravleva et al. 2012; Shang & Oh 2012). Whereas these observations greatly contribute to our understanding of the non-thermal phenomena in the intracluster gas, it is hard to use them to estimate the turbulence pressure accurately. Moreover, these observations are mostly limited to nearby clusters or the inner regions with high surface brightness (see Nagai et al. 2013 for an estimation of the detectability of intracluster gas motions by the upcoming Astro-H mission).

On the other hand, the mass estimates require an accurate determination of the non-thermal pressure in the outskirts of clusters where most of the mass resides. Therefore, the amplitude of intracluster turbulence pressure in the outskirts has to be derived theoretically from the existing knowledge of the injection and dissipation of intracluster turbulence.

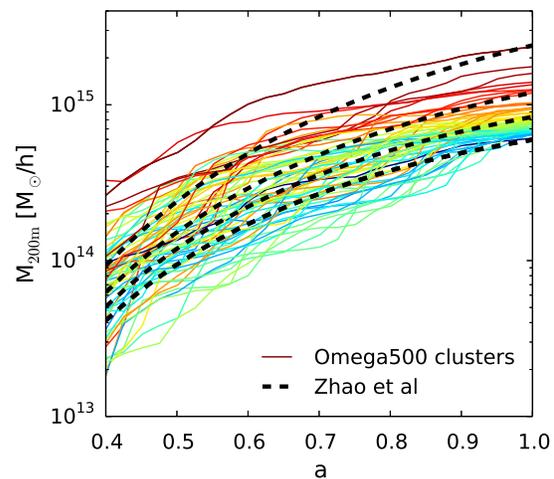
One way to estimate the turbulence pressure is to measure it from cosmological hydrodynamics simulations. However, since a large, high-precision light-cone hydrodynamics simulation is still too expensive to carry out, it is desirable to have an analytical model that can predict the turbulence pressure, either alone or combined with dark matter only N-body simulations. More importantly, an analytical model is based on physical understandings. Thus, by comparing the predictions drawn from an analytical model to simulations and observations, the physical understandings can be tested and improved, forming a healthy feedback loop.

To this end, Shi & Komatsu (2014) (SK14) developed an analytical model for computing the time evolution of the intracluster turbulence pressure. The model is based on a physical picture of turbulence injection during hierarchical cluster mass assembly, and turbulence dissipation with a time-scale determined by the turnover time of the largest turbulence eddies. In this paper, we shall compare the turbulence pressure predicted by this analytical model to that measured in a state-of-art cosmological hydrodynamics simulation. This comparison will test the validity of the analytical model as well as some aspects of the underlying physical picture.

The rest of the paper is organized as follows. In Sect. 2, we introduce the simulation and the cluster sample used for the comparison. In Sect. 3, we demonstrate how to apply the analytical model to the simulation data. In Sect. 4, we present and discuss the results. The underlying physical picture of turbulence injection and dissipation, as well as how to test them more thoroughly, are discussed in Sect. 5. We conclude in Sect. 6.



**Figure 1.** Distribution of cluster masses at  $z = 0$  in the mass-limited sample of simulated galaxy clusters.



**Figure 2.** Mass accretion histories of the mass-limited sample of 65 clusters from the Omega500 simulation. Each solid line shows the mass accretion history of one simulated cluster, color-coded according to its mass at  $z = 0$  ( $a = 1$  where  $a$  is the scale factor). We also show the mean halo mass accretion histories of four different halo masses computed using the Zhao et al. (2009) method (black dashed lines).

## 2 SIMULATION AND CLUSTER SAMPLE

We compare the SK14 model with the outputs of the Omega500 simulation (Nelson et al. 2014), a large cosmological Eulerian simulation performed with the Adaptive Refinement Tree (ART) N-body+gas-dynamics code (Kravtsov 1999; Kravtsov, Klypin & Hoffman 2002; Rudd, Zentner & Kravtsov 2008). In order to achieve the dynamic ranges necessary to resolve the cores of halos, adaptive refinement in space and time and non-adaptive refinement in mass (Klypin et al. 2001) are used. The simulation has a comoving box length of  $500 h^{-1} \text{Mpc}$  and a maximum comoving spatial resolution of  $3.8 h^{-1} \text{kpc}$ , and is performed in a flat  $\Lambda \text{CDM}$  model with the WMAP five-year cosmological parameters (Komatsu et al. 2009). For consistency with the physics included in the analytical model, the simulation we use does not include radiative cooling or feedback. See Nelson, Lau & Nagai (2014) for the implications of neglecting these additional physics in simulations.

We select a mass-limited sample of 65 galaxy clusters at  $z = 0$  from the simulation. Its mass distribution is shown in Fig. 1. We measure one-dimensional profiles of various quantities such as the density and pressure at 25 snapshots between  $z = 0$  and  $z = 1.5$ . See Nelson et al. (2014) and Nelson, Lau & Nagai (2014) (NLN14) for more information on the simulation and the cluster sample.

Fig. 2 shows the mass accretion histories of the cluster sample. Each of 65 clusters is assigned a color depending on their final mass at  $z = 0$ . The mass,  $M_{200m}$ , is defined as the mass enclosed within the radius,  $r_{200m}$ , within which the average matter density equals 200 times the mean mass density of the universe. The dashed lines in Fig. 2 show the analytical mean halo mass accretion histories of Zhao et al. (2009) for four representative halo masses. We find that the mass accretion histories of the simulated clusters largely agree with that predicted by Zhao et al. (2009), despite that the most massive clusters in the sample show slightly slower mass accretion histories than the prediction of Zhao et al. (2009). This suggests that the few most massive clusters in the simulated cluster sample can be slightly more relaxed than the cosmic average. We do not expect this to affect generality of our results.

### 3 ANALYTICAL MODEL OF NON-THERMAL PRESSURE

#### 3.1 The model

The SK14 model uses a first-order differential equation

$$\frac{d\sigma_{\text{nth}}^2}{dt} = -\frac{\sigma_{\text{nth}}^2}{t_d} + \eta \frac{d\sigma_{\text{tot}}^2}{dt}, \quad (1)$$

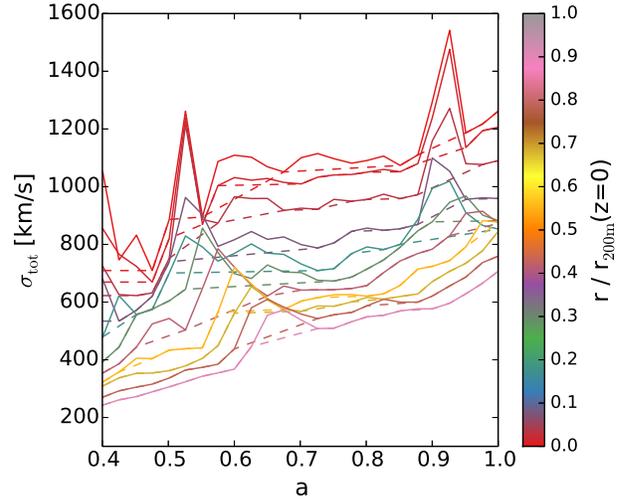
to describe the time evolution of turbulence velocity dispersion squared,  $\sigma_{\text{nth}}^2$ , which is also the turbulence pressure  $P_{\text{nth}}$  per unit density, i.e.,  $\sigma_{\text{nth}}^2 \equiv P_{\text{nth}}/\rho_{\text{gas}}$ . The evolution of  $\sigma_{\text{nth}}^2$  is sourced by that of the total velocity dispersion squared,  $\sigma_{\text{tot}}^2$ , which is the sum of turbulence ('nth', non-thermal) and thermal ('th') velocity dispersion squared:

$$\sigma_{\text{tot}}^2 \equiv \frac{P_{\text{th}}}{\rho_{\text{gas}}} + \sigma_{\text{nth}}^2 = \frac{P_{\text{tot}}}{\rho_{\text{gas}}}, \quad (2)$$

with  $P_{\text{tot}} \equiv P_{\text{th}} + P_{\text{nth}}$ . The turbulence dissipation time scale,  $t_d$ , is taken to be proportional to the dynamical time of the intra-cluster gas,  $t_d = \beta t_{\text{dyn}}/2$ . It can be derived from the accumulated total mass profile,  $M(< r)$ , as the dynamical time is defined by  $t_{\text{dyn}} \equiv 2\pi \sqrt{r^3/[GM(< r)]}$ . In general,  $\sigma_{\text{tot}}^2$ ,  $t_d$ , and hence  $\sigma_{\text{nth}}^2$ , are all functions of radius, mass, and redshift of a cluster. The two parameters in the model,  $\eta$  and  $\beta$ , are taken to be constants by assumption.

We need  $\sigma_{\text{tot}}^2$  and  $t_d$  to solve equation (1). These quantities are, to first order, dictated by the gravitational potential. The essential input knowledge is then how the gravitational potential deepens with time, or simply the mass accretion history. Different clusters have different mass accretion histories; thus, to compare the model predictions with the simulated clusters on a cluster-by-cluster basis, we take  $\sigma_{\text{tot}}^2$  and  $t_d$  directly from the simulation outputs of individual clusters.

We measure the turbulence velocity dispersion,  $\sigma_{\text{nth}}$ , in each radial shell as the r.m.s. velocity after subtracting the mean velocity of the shell with respect to the center-of-mass velocity of the total mass interior to this radial shell (NLN14). In order to remove the kinetic energy associated with sub-structures which does not contribute to the pressure of the global intracluster gas, we also exclude the contribution from gas that lies in the high-density tail in the probability distribution of gas densities accord-



**Figure 3.** Growth of  $\sigma_{\text{tot}}$  as a function of the scale factor of the universe in one representative cluster with a typical mass and accretion history. Each solid line shows  $\sigma_{\text{tot}}$  measured in the simulation at a certain Eulerian radius indicated by the color bar. The dashed lines are the smoothed  $\sigma_{\text{tot}}$  growth curves used in the modeling.

ing to the procedure presented in Zhuravleva et al. (2013). In addition, we smooth the profiles with the Savitzky-Golay filter used in Lau, Kravtsov & Nagai (2009). We then compute the total velocity dispersion squared,  $\sigma_{\text{tot}}^2$ , according to equation (2).<sup>2</sup> We then compute the non-thermal pressure fraction,  $f_{\text{nth}}$ , as their ratio, i.e.,  $f_{\text{nth}} \equiv \sigma_{\text{nth}}^2/\sigma_{\text{tot}}^2$ .

#### 3.2 Smoothing the source term

As a cluster grows in mass, its  $\sigma_{\text{tot}}$  generally increases, suggesting a positive source term in the right hand side of equation (1). For the simulated clusters, however, the  $\sigma_{\text{tot}}$  at each Eulerian radius may also decrease due to local inhomogeneities. As an example, in Fig. 3 we show  $\sigma_{\text{tot}}$  at a few radial bins of one cluster as a function of the scale factor of the universe. The selected cluster has a mass of  $8.9 \times 10^{14} h^{-1} M_{\odot}$  and an accretion history proxy  $\Gamma_{200m} = 2.3$  (see Sect. 4.3), both close to the median values of the mass-limited cluster sample. Some wiggles exist in  $\sigma_{\text{tot}}$ , which propagate from small to large radii. They likely correspond to outwardly moving merger shocks with Mach numbers around 1.5 and sweep across the cluster in a time of 1–2 Gyr.

The analytical model does not intend to capture such transient phenomena, but rather their long-term effect on the intracluster medium. Therefore we smooth these wiggles to reduce their numerical effect. We do so by choosing the points from each simulated  $\sigma_{\text{tot}}(a)$  curve which have a smaller value than all the points to

<sup>2</sup> Alternatively, one may compute  $\sigma_{\text{tot}}^2$  from  $P_{\text{tot}}$  which by itself is computed using the hydrostatic equilibrium equation, and then derive  $\sigma_{\text{nth}}^2$  as the difference of  $\sigma_{\text{tot}}^2$  and  $P_{\text{th}}/\rho_{\text{gas}}$ . Since simulated galaxy clusters are not spherically symmetric nor fully relaxed, this alternative method yields slightly different  $\sigma_{\text{tot}}^2$ . While the  $\sigma_{\text{tot}}^2$  profiles of the cluster sample computed with the two methods are very similar in the virial region of the clusters, the  $\sigma_{\text{nth}}^2$  profiles are significantly different because the alternative method computes  $\sigma_{\text{nth}}^2$  as the difference of two large quantities. Since  $\sigma_{\text{nth}}^2$  computed this way is more prone to numerical errors, we choose the method described in the main text to compute  $\sigma_{\text{nth}}^2$  and  $\sigma_{\text{tot}}^2$ .

the right of this curve (at a larger  $a$ ), and fit linearly between these chosen points. We then use the resulting monotonously increasing  $\sigma_{\text{tot}}(a)$  (dashed lines in Fig. 3) in the modeling.

### 3.3 Initial condition

In SK14 we have argued that, as long as the initial time is chosen to be early enough, the choice of the value of  $f_{\text{nth}}$  at the initial time does not affect the final value of  $f_{\text{nth}}$ . In the inner region of the cluster, this is because the short turbulence dissipation time drives  $f_{\text{nth}}$  quickly to its limiting value determined by the ratio of  $t_d$  and the cluster mass growth time scale (see Sect. 3.2 of SK14). In the cluster outskirts, the turbulence pressure does accumulate throughout time, but the growth is significant after the region enters the virial radius of the cluster, which occurs only at late times.

In this paper, the initial time is chosen at  $z = 1.5$ , which is early enough for the above arguments to hold to a high degree of accuracy for studying cluster profiles at  $z = 0$ . Thus, for convenience and consistency, we choose the initial condition to be  $f_{\text{nth}} = 0$  for all clusters at  $z = 1.5$ . Another option, namely using the values of  $f_{\text{nth}}$  measured from the simulation at  $z = 1.5$ , can provide a more precise initial condition, but only for regions inside clusters which are dynamically relaxed at that time. We have compared the  $f_{\text{nth}}$  values at  $z = 0$  using this initial condition with those using the default initial condition. The difference is negligible inside  $r_{200m}$ .

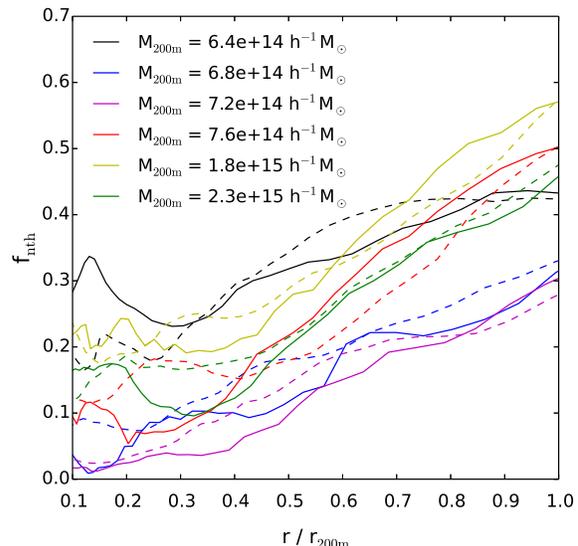
## 4 RESULTS: MODEL VS SIMULATION

We shall limit the comparison between the model predictions and the simulation outputs to  $(0.1 - 1)r_{200m}$ . We avoid the cluster core region ( $r < 0.1r_{200m}$ ) because of both theoretical and numerical difficulties there, such as the uncertainty on the feedback effect of the central AGNs, the disagreement of numerical methods on gas thermodynamical quantities in the core region, and the ambiguity in the choice of the cluster center and its consequence on the projected one-dimensional profiles. We restrict the study to  $r < r_{200m}$  (about  $1.3r_{\text{vir}}$  at  $z = 0$  and  $r_{\text{vir}}$  at  $z = 1$  in a standard  $\Lambda$ CDM cosmology for cluster-mass objects), and avoid the infall region in which the inward acceleration of gas introduces a significant additional source of deviation from the hydrostatic equilibrium (Lau, Nagai & Nelson 2013; Suto et al. 2013; Nelson et al. 2014). We choose  $\beta = 1$  and  $\eta = 0.7$  as the preferred value (SK14). Effects of varying  $\beta$  and  $\eta$  will be presented in Sect. 4.2. All comparison will be performed on the cluster sample at  $z = 0$ .

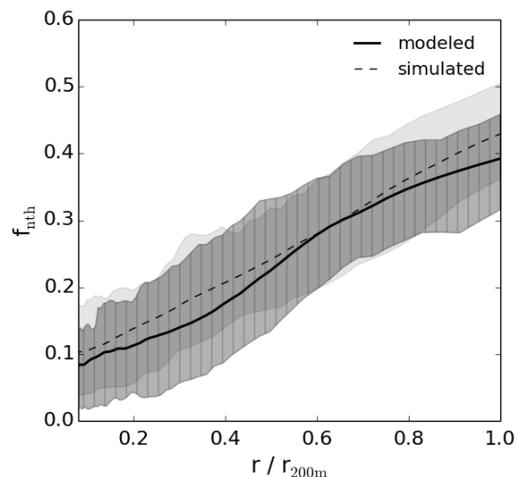
### 4.1 Non-thermal pressure fraction

We show the comparison of the modeled and simulated non-thermal fraction profiles of 6 clusters in Fig. 4. The clusters are selected such that their masses spread over the full range. For all clusters shown, there is a clear trend of  $f_{\text{nth}}$  increasing with radius in the simulated profiles. This trend is a natural consequence of an increasing turbulence dissipation time at larger radii, and is well reproduced by the modeled profiles. On the other hand, the values of the non-thermal fraction at the same radius scaled by  $r_{200m}$  vary by a factor of a few among the clusters. This distinctive difference in the  $f_{\text{nth}}$  values is also well reproduced by the modeled profiles.

The mean  $f_{\text{nth}}$  profiles of the whole sample are shown in Fig. 5. The solid and the dashed lines are the modeled and simulated profiles, respectively. Not only does the mean agree, but also the magnitude of the scatter (shown by the shaded regions) agrees.



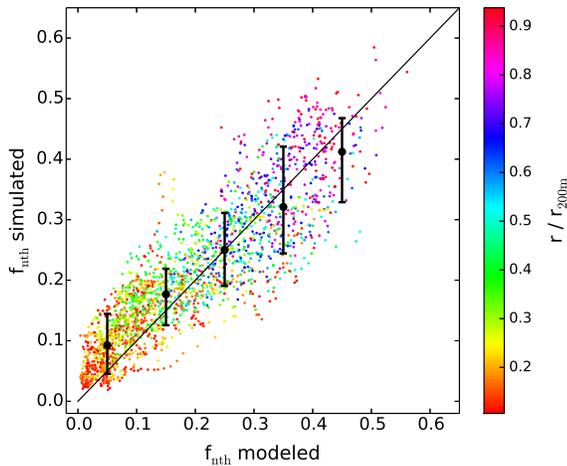
**Figure 4.** Comparison of modeled (solid lines) and simulated (dashed lines)  $f_{\text{nth}}$  profiles of individual clusters. Profiles of 6 typical clusters with a spectrum of different masses at  $z = 0$  are shown.



**Figure 5.** Non-thermal fraction profile of the mass-limited sample. The solid line and the hatched shaded region are the mean and the 16/84 percentile of the modeled profiles; the dashed line and the unhatched shaded region are those of the simulated profiles.

Fig. 6 shows a more quantitative comparison of the modeled and simulated  $f_{\text{nth}}$  values. Each data point here shows the modeled versus simulated  $f_{\text{nth}}$  values in one logarithmic radial bin of one cluster in the sample. Larger  $f_{\text{nth}}$  values are found at larger radii, as shown by the color-coding. To guide the eye, we group the data points into bins according to their modeled  $f_{\text{nth}}$  values, and mark the median simulated  $f_{\text{nth}}$  value of each group with a black point whose x-position indicates the center of the bin. The associated error bar shows the  $1\sigma$  scatter of the simulated  $f_{\text{nth}}$  distribution. We find an excellent agreement between the modeled and simulated  $f_{\text{nth}}$ .

Looking closer, the slight deviation of the black points from the one-to-one relation (the diagonal line) at large  $f_{\text{nth}}$  values can be explained by the selection effect that only data points between  $0.1$  and  $1 r_{200m}$  are shown. The same selection effect does not seem sufficient to explain the deviation at small  $f_{\text{nth}}$  values, and this may



**Figure 6.** Comparison of the modeled and simulated non-thermal fraction,  $f_{\text{nth}}$ , of the mass-limited sample. Each point on the scatter plot shows one radial bin of one cluster in the sample, and is color-coded according to the central radius of the bin relative to  $r_{200m}$ . Only radial bins between 0.1 and  $1 r_{200m}$  are shown. The black points with error bars show the median and 16/84 percentile of the distribution of  $f_{\text{nth}}$  measured from the simulation in bins of modeled  $f_{\text{nth}}$  values. The diagonal line shows the one-to-one correspondence.

suggest a systematic tendency of a smaller modeled than simulated non-thermal fraction at  $r < 0.25r_{200m}$ . Although the statistical significance is only  $1\sigma$ , we offer a possible explanation of this deviation in Sect. 5.

#### 4.2 Effect of varying model parameters

The two parameters in the analytical model,  $\eta$  and  $\beta$ , are physical parameters related to turbulence injection and dissipation, respectively. However, their values are not yet well-constrained from theory. In SK14, we find that  $\eta\beta \approx 0.7$  and  $\beta \approx 1$  provide an excellent agreement between the model predictions and the fitting formulae derived from the existing observations (Arnaud et al. 2010; Planck Collaboration et al. 2013) and numerical simulations (Shaw et al. 2010; Battaglia et al. 2012). The same values reproduce the simulation outputs used in this paper.

To examine how sensitive the comparison results are to the exact values of  $\eta$  and  $\beta$ , we show the effects of varying them in Fig. 7. Each panel in Fig. 7 uses different values of  $\beta$  and  $\eta$  as shown, and the central panel with  $\eta = 0.7$  and  $\beta = 1$  is identical to Fig. 6.

When the cluster mass growth is fast, i.e., when  $\sigma_{\text{tot}}^2$  increases with a timescale  $t_{\text{growth}}$  shorter than the turbulence dissipation time scale  $t_d$ , the non-thermal fraction approaches  $\eta$  (Sect. 3.2 in SK14). In the opposite case, the non-thermal fraction approaches  $\eta t_d / t_{\text{growth}} \propto \eta\beta$ . At  $z \approx 0$ ,  $t_d \ll t_{\text{growth}}$  in the inner region of a galaxy cluster, whereas  $t_d / \beta$  is comparable to  $t_{\text{growth}}$  in the outskirts. This suggests that  $f_{\text{nth}}$  is roughly proportional to  $\eta\beta$  when  $\beta < 1$ , and the shape of the radial dependence of  $f_{\text{nth}}$  is mainly given by the increase of the dynamical time with radius. For larger values of  $\beta$ , the radial dependence of  $f_{\text{nth}}$  should flatten towards large radii due to the saturation of  $f_{\text{nth}}$  to the value of  $\eta$  in the fast growth regime.

These features are clearly visible in Fig. 7: the slope of the modeled versus simulated  $f_{\text{nth}}$  relation is primarily determined by  $\eta\beta$ , and the curvature of the relation by  $\beta$ . As far as the slope is

concerned, the three panels on the diagonal from bottom left to top right with  $0.5 \leq \eta\beta \leq 1$  provide a good match between the modeled and simulated values. From the curvature of the relation, the central panel with the default parameter values give the best agreement, in the sense that the scatter of the data points at each radius (with each color) is most symmetric around the one-to-one relation.

#### 4.3 Dynamical state

SK14 used the analytical mean mass accretion history of Zhao et al. (2009) to show that the average non-thermal pressure fraction increases with the cluster mass and redshift. This feature is hard to test directly with the simulated cluster sample described in Sect. 2 due to the limited range of masses and redshifts for which the profiles of the clusters are well-resolved. Also, as discovered by NLN14, the redshift and mass dependencies are greatly reduced when the cluster radius is scaled by  $r_{200m}$ .

Still, we can divide the simulated cluster sample by their accretion histories, and test whether the model and the simulation yield the same difference on  $f_{\text{nth}}$  between the sub-samples. Since the model attributes the origin of the mass and redshift dependencies of  $f_{\text{nth}}$  to the dependence on the recent mass accretion history, this provides a more direct test of the model prediction than comparing the average non-thermal pressure fraction of cluster samples at different redshift or with different masses.

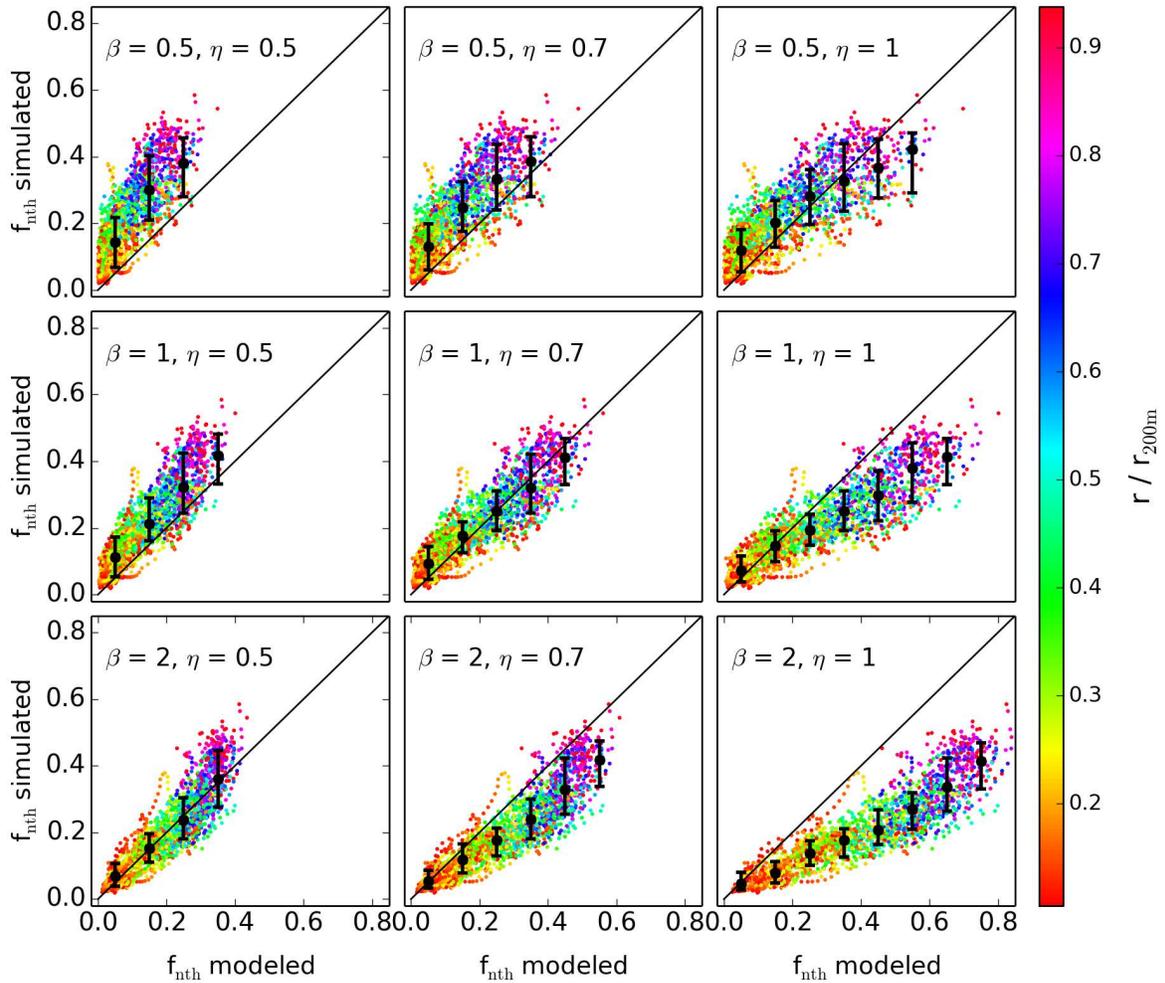
We adopt a simple quantification of the recent accretion history as introduced by NLN14 and Diemer & Kravtsov (2014),

$$\Gamma_{200m} \equiv \frac{\log_{10}[M_{200m}(z=0)] - \log_{10}[M_{200m}(z=0.5)]}{\log_{10}[a(z=0)] - \log_{10}[a(z=0.5)]}. \quad (3)$$

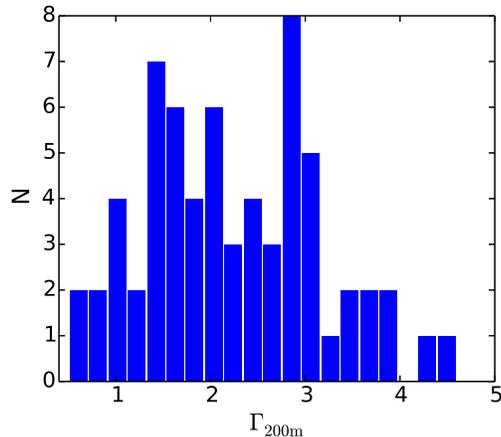
A larger  $\Gamma_{200m}$  value indicates more mass growth since  $z = 0.5$ . The value of  $\Gamma_{200m}$  is also an indicator of the dynamical state, as there is a strong correlation between the recent mass growth and the dynamical state of a galaxy cluster,

The distribution of  $\Gamma_{200m}$  in the mass-limited cluster sample is shown in Fig. 8. We select two sub-samples of the simulated clusters with  $\Gamma_{200m} < 1.8$  and  $\Gamma_{200m} > 2.7$ , respectively. Both sub-samples contain 23 galaxy clusters. We apply the analytical model to each cluster in the sub-sample and compare the mean  $f_{\text{nth}}$  profile of each sub-sample with the corresponding simulated one. As shown in Fig. 9, the sub-sample with higher recent mass growth has a significantly higher non-thermal pressure fraction at all radii. This is consistent with the previous numerical studies (e.g., Nagai, Vikhlinin & Kravtsov 2007; Piffaretti & Valdarnini 2008; Nelson et al. 2012) which consistently find a larger hydrostatic mass bias for less relaxed, recently merged clusters. This difference in the average non-thermal pressure fraction is remarkably well reproduced by the analytical model. This result reinforces the basic underlying physical picture that intracluster turbulence is triggered during the cluster mass assembly, and that the kinetic energy in the intracluster turbulence is derived ultimately from the gravitational energy released during the structure growth.

In Fig. 10 we compare the modeled and simulated non-thermal fractions in each radial bin of each cluster in the two sub-samples. It is clear that, for the early accretion ( $\Gamma_{200m} < 1.8$ ) sub-sample which consists of more dynamically relaxed clusters at the time of comparison ( $z = 0$ ), the scatter between the modeled and simulated  $f_{\text{nth}}$  values is smaller. This is in accord with the expectation that the analytical model works better for dynamically relaxed clusters. Nevertheless, a clear correlation exists also for the more disturbed clusters ( $\Gamma_{200m} > 2.7$ ), suggesting that the analytical



**Figure 7.** Effect of varying the parameters  $\beta$  and  $\eta$ . In each panel the symbols are the same as those in Fig. 6. The central panel with  $\beta = 1$  and  $\eta = 0.7$  is identical to Fig. 6.



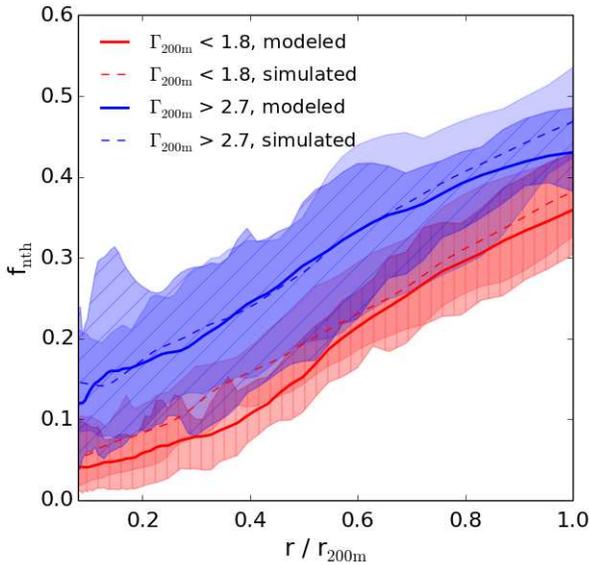
**Figure 8.** Distribution of the proxy of the accretion history and dynamical state,  $\Gamma_{200m}$ , computed from the mass-limited sample of simulated galaxy clusters.

model is also applicable to these systems in estimating the turbulence pressure, though with greater noise.

We note that the  $\Gamma_{200m}$  parameter used in this paper is not optimal as a proxy for the dynamical state, since it is defined with the mass increase between two snapshots. By definition, the dynamical state of a cluster can be determined by its temporary state. A dynamical state proxy defined at a single snapshot based on the dynamical properties of the halo particles would be more convenient to use, and at the same time provide a more direct characterization of the deviation from hydrostatic equilibrium. For future studies of assigning the non-thermal pressure profile to dark matter halos extracted from the dark-matter only N-body simulation, such more advanced dynamical state proxy may be preferred.

## 5 DISCUSSION: TEST OF THE PHYSICAL PARADIGM

Evolution of the intracluster turbulence is a problem involving a vast range of spatial and time scales. The relevant physical processes include the cluster mass assembly in a cosmological context, the merger and accretion shocks which convert the bulk kinetic energy into the turbulence kinetic energy and heat, and the detailed intracluster gas dynamics associated with the development and cascade of turbulence. Simulating all of them with sufficient numerical precision is beyond the reach of a single set of numerical simula-



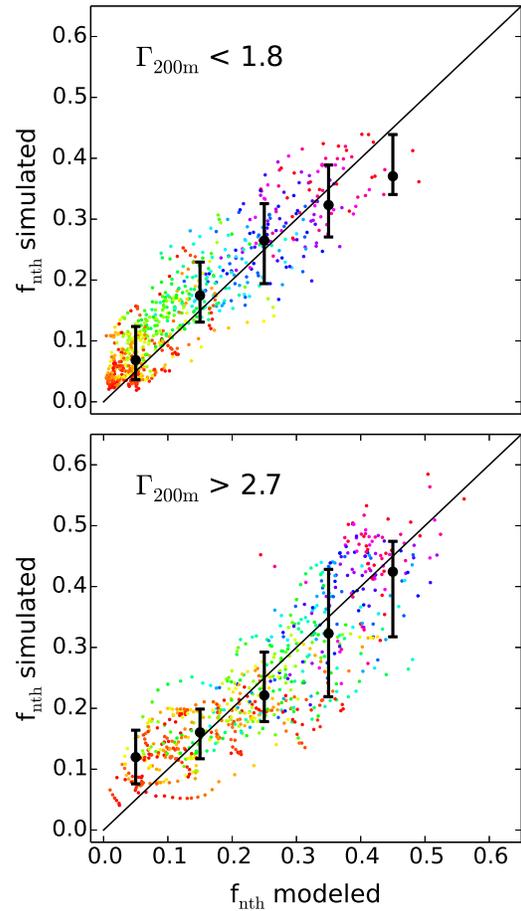
**Figure 9.** Modeled and simulated  $f_{\text{nth}}$  profiles of an early growth sub-sample ( $\Gamma_{200\text{m}} < 1.8$ , blue lines) and a late growth sub-sample ( $\Gamma_{200\text{m}} > 2.7$ , red lines). The lines and the shaded regions are the mean profile of the sample and the 16/84 percentiles, respectively.

tions. Simulations dedicated to certain physical processes would be needed for testing them in greater detail.

In this respect, the large-size cosmological simulation used in this paper is ideal for testing the relation of turbulence growth with cluster mass assembly in a cosmological context, for which the picture underlying the analytical model has been verified by the positive results presented in Sect. 4. On the other hand, cosmological simulations of a single cluster (e.g., Vazza et al. 2009, 2011; Paul et al. 2011; Miniati 2014) are better suited for studying mechanisms of turbulence injection, and high resolution simulations performed on a fixed grid (Gaspari & Churazov 2013; Gaspari et al. 2014) are better suited for studying the turbulence cascade process. The insights gained from these dedicated simulations can be easily incorporated into the analytical model.

For a precise assessment of the amplitude of intracluster turbulence pressure, it is important to know the effective thermalization ratio at turbulence injection, and the turbulence dissipation time scale. In the framework of the SK14 analytical model, this suggests the need to determine the values of the model parameters  $\eta$  and  $\beta$  and investigate their possible dependence on radius, redshift, and cluster mass. These can in principle be realized by dedicated numerical simulations, when numerical effects in these simulations are well-understood and controlled. Recently, using the moving-mesh numerical scheme, Schaal & Springel (2014) reported a higher energy dissipation fraction contributed by shocks in the warm hot intergalactic medium, and correspondingly a higher average Mach number of shocks at which the bulk of energy dissipates, than previous studies performed with the Adaptive Mesh Refinement technique (e.g., Ryu et al. 2003; Pfrommer et al. 2006; Kang et al. 2007; Vazza et al. 2011; Planelles & Quilis 2013). This, if confirmed, would suggest a higher thermalization ratio, and that a radius and redshift dependence of  $\eta$  would be determined by the relative importance of the high Mach number accretion shocks and the low Mach number internal shocks.

We note that the SK14 analytical model is not consistent



**Figure 10.** Comparison of modeled and simulated  $f_{\text{nth}}$  of the early growth sub-sample (upper panel) and the late growth sub-sample (lower panel). In each panel the symbols are the same as those in Fig. 6.

with the long-term power law decay behavior expected for the turbulence kinetic energy (Landau & Lifshitz 1959; Frisch 1995; Subramanian, Shukurov & Haugen 2006). This inconsistency is due to our assumption of a one-to-one relation between the cluster radius and the turbulence dissipation time scale, which is the ratio of the size and velocity of the largest eddies, at that radius (i.e.,  $t_d \propto t_{\text{dyn}}$ ). The consequence of this assumption is most visible in the regions where turbulence dissipates much faster than it grows, and may have contributed to the possible systematic difference between the modeled and simulated  $f_{\text{nth}}$  at small radii. To correct for this, one may need to include a spectral dimension to the model, that is to keep track of the power spectrum of turbulence velocity field at each radius as a function of time. This, in turn, will allow for an easier link to intracluster magnetic fields and cosmic rays.

## 6 CONCLUSION AND PERSPECTIVES

We have compared the SK14 analytical model for the turbulence pressure inside galaxy clusters to a state-of-the-art hydrodynamics numerical simulation. The analytical model and the simulation outputs show excellent agreement on the non-thermal pressure

fraction on a cluster-by-cluster basis - both its radial profile and its dependence on the cluster mass accretion history.

This demonstrates that the SK14 model in its current form can already be used to predict the amplitude of intracluster turbulence pressure with a precision comparable to that of the state-of-art cosmological hydrodynamics simulations. This opens up an exciting possibility that we may be able to use the analytical model to correct the systematic bias in the mass estimation of galaxy clusters due to the turbulence pressure. The analytical model, in turn, would also provide a convenient and efficient way to interpret the SZ power spectrum and observations of cluster outskirts from ongoing and upcoming large cluster surveys.

At the same time, the comparison results show that a simple analytical model can indeed capture the basic physical processes related to the evolution of intracluster turbulence pressure. In particular, our comparison study has verified the underlying physical picture that the turbulence growth is determined by cluster mass assembly in a cosmological context. The detailed physics regarding injection and dissipation of intracluster turbulence requires further tests from comparisons with dedicated high resolution simulations of individual clusters. We point out that adding a spectral dimension to the model may lead to a better description of the long-term dissipation of the turbulence, further improve the consistency with simulations in the inner regions of clusters, and provide a framework for a unified understanding of non-thermal phenomena in galaxy clusters.

#### ACKNOWLEDGEMENTS

XS thanks Massimo Gaspari for helpful discussions. KL and DN acknowledge support from NSF grant AST-1009811, NASA ATP grant NNX11AE07G, NASA Chandra grants GO213004B and TM4-15007X, and the Research Corporation. This work was supported in part by the facilities and staff of the Yale University Faculty of Arts and Sciences High Performance Computing Center.

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