Correction to scaling in the response function of the two-dimensional kinetic Ising model

Federico Corberi,* Eugenio Lippiello,[†] and Marco Zannetti[‡]

Istituto Nazionale di Fisica della Materia, Unità di Salerno and Dipartimento di Fisica "E. Caianiello," Università di Salerno, 84081

Baronissi (Salerno), Italy

(Received 27 October 2004; revised manuscript received 27 July 2005; published 3 November 2005)

The aging part $R_{ag}(t,s)$ of the impulsive response function of the two-dimensional ferromagnetic Ising model, quenched below the critical point, is studied numerically employing an algorithm without the imposition of the external field. We find that the simple scaling form $R_{ag}(t,s) = s^{-(1+a)}f(t/s)$, which is usually believed to hold in the aging regime, is not obeyed. We analyze the data assuming the existence of a correction to scaling. We find $a=0.273\pm0.006$, in agreement with previous numerical results obtained from the zero field cooled magnetization. We investigate in detail also the scaling function f(t/s) and we compare the results with the predictions of analytical theories. We make an ansatz for the correction to scaling, deriving an analytical expression for $R_{ag}(t,s)$. This gives a satisfactory qualitative agreement with the numerical data for $R_{ag}(t,s)$ and for the integrated response functions. With the analytical model we explore the overall behavior, extrapolating beyond the time regime accessible with the simulations. We explain why the data for the zero field cooled susceptibility are not too sensitive to the existence of the correction to scaling in $R_{ag}(t,s)$, making this quantity the most convenient for the study of the asymptotic scaling properties.

DOI: 10.1103/PhysRevE.72.056103

PACS number(s): 05.70.Ln, 75.40.Gb, 05.40.-a

I. INTRODUCTION

A powerful tool in the study of slowly relaxing systems is the extension of the fluctuation dissipation theorem (FDT) to off-equilibrium conditions. The relation between the response and the autocorrelation function has been shown to encode basic properties of the dynamics [1] and of the equilibrium state [2]. However, while the autocorrelation function C(t,s) has been studied for a long time and, in some systems, is well understood, the knowledge of the response function R(t,s) remains comparatively poor.

In the prototypical case, considered here, of nondisordered coarsening systems quenched below the critical temperature, general properties of C(t,s) and R(t,s) may be inferred from the structure of the configurations. It is well known, in fact, that in the late stage of phase ordering the interior of the growing domains is equilibrated, while their boundary is out of equilibrium. Accordingly, a distinction between bulk and interface fluctuations can be made. For the autocorrelation function this leads to the splitting [3–6]

$$C(t,s) = C_{st}(t-s) + C_{ag}(t,s)$$
 (1)

where $C_{st}(t-s)$ is the contribution from the interior of domains, behaving as the equilibrium autocorrelation function $C_{eq}(t-s)$ of the ordered state at the final temperature *T* of the quench, while $C_{ag}(t,s)$ is the off-equilibrium aging contribution, coming from the interfaces. Similarly, for the response function one has

$$R(t,s) = R_{st}(t-s) + R_{ag}(t,s)$$
⁽²⁾

and the stationary contributions $R_{st}(t-s)$ and $C_{st}(t-s)$ are related by the FDT

$$TR_{st}(t-s) = \frac{\partial C_{st}(t-s)}{\partial s}.$$
(3)

For s large, $C_{ag}(t,s)$ obeys the simple aging scaling form

$$C_{ag}(t,s) = s^{-b}g(x) \tag{4}$$

with b=0 and x=t/s [6–8]. Taking s=0 and t large, $C_{ag}(t,0)$ decays algebraically as $C_{ag}(t,0) \sim t^{-\lambda/z}$, where λ is the Fisher-Huse exponent [9] and z is the dynamic exponent regulating the growth of the characteristic size $L(t) \sim t^{1/z}$ of domains. Assuming that the same behavior holds also for s>0 [10] gives $g(x) \sim x^{-\lambda/z}$, for $x \ge 1$. These properties of C(t,s), as mentioned above, are well understood and well documented by both analytical and numerical results.

By analogy, the response function is also expected to obey, for large s, a simple aging scaling form

$$R_{ag}(t,s) = s^{-(1+a)} f(x)$$
(5)

with

$$f(x) \sim x^{-\lambda_R/z} \tag{6}$$

for $x \ge 1$. Evidence that this is the case comes from the few models tractable by analytical methods [4,11–15]. In the large majority of cases, where information can be extracted only from numerical simulations, investigating R(t,s) is considerably more difficult than C(t,s). The reason is that R(t,s) is much more noisy than C(t,s). A possible way around this difficulty is to study the less noisy integrated response functions (IRFs), such as the zero field cooled (ZFC) magnetization

^{*}Electronic address: corberi@sa.infn.it

[†]Electronic address: lippiello@sa.infn.it

[‡]Electronic address: zannetti@na.infn.it

$$\chi(t,t_w) = \int_{t_w}^t ds \ R(t,s) \tag{7}$$

or the thermoremanent magnetization (TRM)

$$\rho(t,t_w) = \int_0^{t_w} ds \ R(t,s). \tag{8}$$

However, if resorting to the IRF does indeed cut down the noise, it has turned out that recovering R(t,s) from the IRF is a delicate task which involves more than one subtlety (see Ref. [16] and Sec. II of this paper). As a result, so far, in the literature no consensus has been reached either on the exponent *a* or on the form of the scaling function f(x). Therefore, it seems that the issue of the properties of R(t,s) can be settled only by going to its direct measurement. Recently, this has become feasible after the introduction of algorithms [17–19] for the computation of the response function without applying an external perturbation. This speeds the simulation to such an extent that the direct measurement of R(t,s) has become accessible [17–20].

In this paper we present results for R(t,s) in the twodimensional Ising model obtained with our algorithm [19]. The outcome is quite rich and interesting. We find the unexpected result that in the d=2 Ising model the simple scaling form (5) is not enough to represent the data, but a large correction term is needed, up to the longest times we have reached in the simulations. The quality of the data allows us to detect and analyze this correction and, by taking it properly into account, to make statements on the exponent a and the scaling function f(x) appearing in Eq. (5).

The paper is organized as follows. In Sec. II we summarize the problem of the exponent *a*. In Sec. III we recall the basic features of the measurement of the response function without applying an external perturbation and the method we use to isolate $R_{ag}(t,s)$ in Eq. (2). In Sec. IV we present and discuss the numerical results for $R_{ag}(t,s)$. In particular, we show the existence of a large preasymptotic scaling correction, we extract the value of the exponent *a*, and we analyze the scaling function f(x). In Sec. V, we investigate the implications of the structure of $R_{ag}(t,s)$ for the properties of the IRFs and we compare our predictions with the numerical simulations of these quantities. This gives further insight into the problem of the retrieval of R(t,s) from the IRFs. Finally, in Sec. VI we draw the conclusions.

II. THE EXPONENT a

Before entering the presentation of data and results, we recall briefly what is the problem with the exponent a in Eq. (5).

(1) The straightforward substitution of Eq. (5) into Eqs. (7) and (8) yields the scaling forms of the aging parts of the IRFs:

$$\chi_{ag}(t,t_w) = t_w^{-a_\chi} F_\chi(y) \tag{9}$$

$$\rho_{ag}(t,t_w) = t_w^{-a_\rho} F_\rho(y) \tag{10}$$

with

$$a_{\chi} = a_{\rho} = a \tag{11}$$

and $y = t/t_w$.

(2) An intuitively appealing argument, originally introduced by Barrat [21], predicts that $\chi_{ag}(t,t_w)$ ought to be proportional to the density of defects at the time t_w , which goes as $L(t_w)^{-n}$, with n=1 and 2 for a scalar and vectorial order parameter, respectively. Using Eq. (11), this leads to the dimensionality independent result

$$a_{\gamma} = a_{\rho} = a = n/z \tag{12}$$

since, we recall, the exponent z in the quenches below T_C does not depend on d. For systems without conservation of the order parameter, as considered in this paper, z=2 [7].

(3) Contrary to the previous statement, *all* available analytical results, which include the exact solutions of the large N model [4] and of the d=1 Ising model [11,12], as well as approximate calculations [13–15], show that a depends linearly on dimensionality according to

$$a = \frac{n}{z} \frac{d - d_L}{d_U - d_L} \tag{13}$$

where d_L is the lower critical dimension of the statics and $d_U > d_L$ is a dimensionality whose significance becomes clear as soon as ZFC magnetization is considered. In fact, from these analytical calculations it comes out that the relation between a_{χ} and a is not as simple as in Eq. (11). Rather, it must be replaced by

$$a_{\chi} = \begin{cases} a & \text{as in Eq. (13) for } d < d_U, \\ n/z & \text{with logarithmic corrections for } d = d_U, \\ n/z & \text{for } d > d_U. \end{cases}$$
(14)

This behavior of a_{χ} is due to the existence of an irrelevant variable in $\chi_{ag}(t, t_w)$, which becomes dangerous for $d \ge d_U$ [4,16]. In the large N model the above formula holds with $(n=2, d_L=2, d_U=4)$, while in the approximate calculations with scalar order parameter of Refs. [13–15] $(n=1, d_L=1, d_U=2)$.

(4) The problem of the exponent a, then, is whether Eq. (13) is a peculiarity of just those cases where analytical results are available, the generic behavior being that of Eq. (12), or, vice versa, it is Eq. (13) that captures the generic behavior, revealing thereafter that the argument leading to Eq. (12) does miss some important feature in the mechanism of the response.

In order to answer the questions one has to investigate as many systems as possible, with the aim of putting together the generic picture. This can be done only by numerical methods. We have carried out such a program performing simulations and measuring a_{χ} in several systems with conserved and nonconserved dynamics, with both scalar and vectorial order parameter, at different dimensionalities [22]. The large body of results that we have obtained does indeed indicate, quite convincingly in our opinion, that Eq. (14) is of

and

general validity, with $d_U=3$ [23] and 4 for systems with scalar and vector order parameter, respectively.

This conclusion is challenged, mainly on the basis of the measurement of a_0 from TRM in the Ising model with d=2and 3 [24], which seems to agree with Eq. (12). In other words, the investigations of ZFC and TRM seem to produce different results. We have explained in detail elsewhere [16] that the difference is only apparent and is due to the fact that while the data for ZFC are asymptotic, those for TRM are not. To us, the interesting question remaining open is not anymore what is the value of a, but what is the physics behind Eq. (13). However, as of yet this is not a shared conclusion [25-27]. So what emerges from this brief account of the problem is that working with the IRFs is a tricky business, because it has opened the somewhat intricate problem of why ZFC and TRM seem to give different results. Therefore, in order to make progress, a fresh start is needed. In this respect, the direct numerical study of R(t,s), with these algorithms, seems well suited to the task. We shall concentrate on the d=2 Ising model, where discrimination between Eqs. (12) and (13) ought to be easier, since the difference in the predicted values of *a* is quite large

$$a = \begin{cases} 1/2 & \text{from Eq. (12),} \\ 1/4 & \text{from Eq. (13).} \end{cases}$$
(15)

III. THE ALGORITHM

A. Measurement of the response without applying a perturbation

The basic idea in the methods of [17-19] for the measurement of the response function without applying the external perturbation is to relate R(t,s) to some correlation function of the unperturbed dynamics, in much the same way as in the equilibrium FDT. In this paper we will use our method introduced in [19], since we have checked that it is numerically more efficient. Let us briefly describe it, referring to [19] for details.

We consider a spin system with Hamiltonian $\mathcal{H}[\sigma]$, where $[\sigma]$ is a generic configuration of the spin variables $\sigma_i = \pm 1$. The response function is defined by

$$R(t,s) = \lim_{\Delta s \to 0} \frac{1}{\Delta s} \left. \frac{\partial \langle \sigma_i(t) \rangle}{\partial h_i} \right|_{h=0}$$
(16)

where h_i is a magnetic field acting on the *i*th site during the time interval $[s, s + \Delta s]$, and the right-hand side does not depend on *i* due to space translational invariance. Computing $\langle \sigma_i(t) \rangle$ by means of the master equation and inserting the result into Eq. (16), one arrives at

$$TR(t,s) = \frac{1}{2} \lim_{\Delta s \to 0} \left(\frac{C(t,s+\Delta s) - C(t,s)}{\Delta s} - \langle \sigma_i(t-\Delta s)B_i(s) \rangle \right)$$
(17)

where B_i enters the evolution of the magnetization according to

$$\frac{d\langle \sigma_i(t)\rangle}{dt} = \langle B_i(t)\rangle.$$
(18)

This result holds in complete generality for generic Hamiltonian and transition rates, with or without conservation of the order parameter. In the single spin flip dynamics

$$B_i(t) = 2\sigma_i(t)w([\sigma] \to [\sigma_i]) \tag{19}$$

where $w([\sigma] \rightarrow [\sigma_i])$ is the transition rate between two configurations differing for the value of the spin on the site *i*. In the simulations time is discretized by single updates which, measuring time in Monte Carlo steps, occur on the microscopic time scale $\epsilon = 1/N$. The best numerical approximation to the limit $\Delta s \rightarrow 0$ in Eq. (17) is obtained by taking $\Delta s = \epsilon$, which gives

$$TR(t,s) = \frac{1}{2} \left(\frac{C(t,s+\epsilon) - C(t,s)}{\epsilon} - \langle \sigma_i(t-\epsilon)B_i(s) \rangle \right).$$
(20)

In addition to R(t,s), in the following we will also be interested in the general IRF defined by [16]

$$\mu(t, [\tilde{t}, t_w]) = \int_{t_w}^{\tilde{t}} R(t, s) ds$$
(21)

which corresponds to the application of the perturbation between the times t_w and $\tilde{t} \le t$. For this quantity, Eq. (20) is replaced by

$$T\mu(t, [\tilde{t}, t_w]) = \frac{1}{2} [C(t, \tilde{t}) - C(t, t_w)] - \frac{\epsilon}{2} \sum_{s=t_w}^{\tilde{t}} \langle \sigma_i(t-\epsilon) B_i(s) \rangle,$$
(22)

where the sum $\sum_{s=t_w}^{\tilde{t}}$ is over the discrete times in the interval $[t_w, \tilde{t}]$.

Despite the advantages of the method, the computation of the impulsive response R(t,s) remains a very demanding numerical task. In order to improve the signal to noise ratio, therefore, we have found it convenient to consider, instead of R(t,s), the quantity $\mu(t,[s+\delta,s])$ with $\delta/s \ll 1$. Assuming for R(t,s) the scaling form (5), to first order in δ/s one has

$$\mu(t, [s+\delta, s]) = R(t, s) \left(\delta - \frac{\delta^2}{s} f_1(x) + \cdots\right)$$
(23)

with $f_1(x) = (1/2)[(1+a)+x(d \ln f(x))/dx]$. For $\delta = 1$, $\mu(t, [s + \delta, s])$ differs from R(t, s) by a correction of order 1/s, which we have checked to be always negligible in our simulations. Therefore, in the following, whenever results for R(t, s) will be presented, it is understood that the data are obtained with this procedure, if no other specification is made.

In the following we shall use the above formulas for the d=2 Ising model with nearest neighbor interaction and evolving with Glauber dynamics.



FIG. 1. Plot of the difference $C(t,s)-C^{NB}(t,s)$ for $s=6\times10^2$, 2×10^3 , 10^4 , 5×10^4 (from bottom to top) showing the convergence toward $C_{eq}(t-s)$ (bold curve) as *s* increases.

B. The aging contribution

In this section we discuss an auxiliary dynamics, referred to as no bulk flip (NBF), which is used [14,16,22,28] to isolate the aging part $R_{ag}(t,s)$ of the response function appearing in Eq. (2).

We introduce a classification of the degrees of freedom by making the distinction between bulk and interface spins. A spin σ_i is regarded as belonging to the bulk of a domain if it is aligned with all its nearest neighbors. In the NBF algorithm bulk spins cannot flip. With this rule the interior of domains orders rapidly and all that is left is interface dynamics. Since in coarsening kinetics the aging contribution comes exclusively from the boundary of domains, measuring quantities in a simulation with the NBF rule yields the aging behavior.

In order to illustrate this idea, let us consider the autocorrelation function. Denoting with $\sigma_i^{NB}(t)$ the value of the spin evolving with NBF dynamics, the quantity

$$C^{NB}(t,s) = M^2 \langle \sigma_i^{NB}(t) \sigma_i^{NB}(s) \rangle, \qquad (24)$$

where M is the equilibrium spontaneous magnetization at the temperature T, is expected to coincide with $C_{ag}(t,s)$. The M^2 factor in front of the definition (24) is needed, recalling [4–6] that $C_{ag}(t,s)$ falls from M^2 (the Edwards-Anderson order parameter) to zero. According to the discussion of Sec. I, $C_{st}(t-s)$ coincides with the equilibrium correlation $C_{eq}(t-s)$, which decays from $1-M^2$ to zero. In Fig. 1 we have made the comparison between the stationary parts of C(t,s) computed in two different ways. In the first case we measure directly $C_{eq}(t-s)$ in the equilibrium state prepared at the temperature $T/J=2.2(M^2 \simeq 0.616)$. We recall that the critical temperature is given by $T_C/J=2.269$ 18, where J is the nearest neighbor coupling constant. The second prescription is of calculating the autocorrelation function $C^{NB}(t,s)$ in the quench from an initial disordered state, corresponding to an infinite temperature, to the same final temperature T with the NBF rule, and then subtracting it from C(t,s), computed with the full dynamics. If the NBF algorithm correctly isolates the aging part of the dynamics what is left is the sta-



FIG. 2. $R_{ag}(x,s)$ (continuous lines) obtained with the NBF algorithm and R(x,s) (squares) obtained with the full quench dynamics are plotted for s=111, 229, 537, 1085, 1577 from top to bottom. Circles represent the fit according to Eq. (46).

tionary term $C(t,s) - C^{NB}(t,s) = C_{eq}(t-s)$. However, one does not expect this to be satisfied at all times. The reason is that a sharp separation between two independent components, bulk and interface, applies only when L(t) is sufficiently large [4,29]. Actually, the larger the domains are, the larger is the average distance from a bulk spin to the nearest interface, so that they are more effectively decoupled. The data of Fig. 1 show a convergence of $C(t,s) - C^{NB}(t,s)$ toward $C_{eq}(t-s)$ as *s* increases, in agreement with the previous discussion. This implies that, for large *s*, the NBF algorithm does isolate the aging part.

One arrives at the same conclusion by considering the response function. The stationary part $R_{eq}(t-s)$ decays to zero on a characteristic microscopic time. Then, by using s and x as independent variables (which we will do from now on) and denoting with R(x,s) the response function in terms of these variables, in the limit of large s the stationary part gives a contribution only for $x \approx 1$. Therefore, for sufficiently large s and x > 1, one has $R(x,s) \simeq R_{ag}(x,s)$. If the NBF algorithm isolates the aging contribution, the response $R^{NB}(x,s)$ obtained with the NBF rule and R(x,s), measured using the usual dynamics, should coincide for large s. In order to check this, we have computed R(x,s) and $R^{NB}(x,s)$ for a system of 1000² spins, quenched from the initial disordered state, corresponding to an infinite temperature, to the final temperature $T/J=1.5(M^2 \approx 0.9732)$. We have considered several values of s ranging from s=100 to 1600 and times t > s up to 5000. More precisely, we have used (and we shall use in the following) values of s generated from $s_n = 100 + \text{Int}(1.5^n)$ with *n* ranging from 1 to 18. The range of times considered belongs to the scaling regime of the system. The results in Fig. 2 show that R(x,s) and $R^{NB}(x,s)$ coincide within the statistical errors. Small differences can be detected only for the smallest values of s. This is expected since, as anticipated, the agreement improves with increasing s.

In conclusion, the analysis of both the autocorrelation and the response function demonstrates the reliability of the NBF algorithm. Moreover, this algorithm has the advantage of speeding up considerably the simulations, since only the



FIG. 3. (Color online) Failure of collapse of the curves of Fig. 2 with a=1/4 (lower set) and 1/2 (upper set).

fraction of spins on the interfaces must be updated. All the results presented in the following have been obtained with this technique.

IV. SCALING OF $R_{ag}(x,s)$

The first task is to check the scaling properties of $R_{ag}(x,s)$. If Eq. (5) was obeyed the curves for $s^{1+a}R_{ag}(x,s)$, obtained for different values of *s* and a suitable *a*, should collapse on a single master curve. However, a rough inspection of Fig. 2 already shows that this is not the case, since the various curves cannot be superimposed by a simple vertical translation. We make this more precise in Fig. 3, where one can see clearly that there is no collapse for neither one of the two values of *a* proposed in Eq. (15). For a=1/4 the collapse is rather good for small *x*, but gets worst with increasing *x*. For a=1/2 the collapse is bad everywhere, except for $x \approx 4$.

A. The effective exponent $a_{eff}^{R}(x,s)$ and the value of a

In order to make these considerations quantitative we introduce the effective exponent defined by

$$1 + a_{eff}^{R}(x,s) = - \left. \frac{\partial \ln R_{ag}(x,s)}{\partial \ln s} \right|_{x}.$$
 (25)

Numerically, for a chosen value of x, $a_{eff}^R(x,s)$ is given by the local slope of the plot of ln ln $R_{ag}(x,s)$ against ln s in a selected interval I_s around s. This interval must be chosen small with respect to the range of s over which $a_{eff}^R(x,s)$ varies appreciably. However, the smaller the interval I_s the more noisy gets $a_{eff}^R(x,s)$. On the basis of the data available from the simulations, the best compromise between the needs of having a local quantity and of lowering the noise has been reached by taking I_s spanning over four consecutive values of s [30]. The result is displayed in Fig. 4, where symbols with error bars represent the numerical values of $a_{eff}^R(x,s)$, obtained for the three different I_s with the four values of s indicated in the legend. The continuous curves have been obtained from a fitting procedure, which will be discussed in Sec. IV C. If Eq. (5) did hold, we should have



FIG. 4. (Color online) Symbols with error bars represent $a_{eff}^{R}(x,s)$ obtained from the simulation data, as local slopes of $\ln R(x,s)$ against $\ln s$ over the four values of *s* specified in the legend. Continuous curves are the plot of $a_{eff}^{R}(x,s)$ obtained from the fitting formula (46) for the same *s* used in the simulations. The frame is magnified in Fig. 5. The inset shows $a_{eff}^{R}(x,s)$ from formula (46) over a range $s=3 \times 10^2$, 10^3 , 10^4 , 10^5 , 10^6 (from top to bottom) that cannot be reached in the simulation.

found a flat effective exponent, i.e., $a_{eff}^{R}(x,s) = a$ independent of x and s. Instead, the data for $a_{eff}^{R}(x,s)$ show the following features.

(1) *Fixed s*. For fixed *s*, there is a strong dependence on *x*, revealing that Eq. (5) is not obeyed. Furthermore, the curves display a discontinous behavior at x=1. The equal times value at x=1, ranging from $a_{eff}^R(1,s)=-0.55$ to -0.53, depending on the I_s considered, is separated by a jump from the smoothly increasing curve starting around $a_{eff}^R(1^+,s)=0.30$. By $x=1^+$ we denote the smallest value of x>1 used in the simulations. For larger values of x, $a_{eff}^R(x,s)$ keeps growing continuously. The longest set of x data, corresponding to I_s with the smallest value of s, shows the possible saturation to an asymptotic finite value $a_{eff}^R(\infty,s) > a_{eff}^R(1^+,s)$. This behavior of a_{eff}^R explains immediately why the attempts to collapse the data with a fixed value of a, as in Fig. 3, do fail except in a restricted range of x.

(2) *Fixed x*. The size of the error bars makes it difficult to detect an *s* dependence for fixed *x*, except in the region of short time difference $x \le 2$, where with the first two I_s the decrease of $a_{eff}^R(x,s)$ upon increasing *s* exceeds the error (Fig. 5). For larger values of *x*, error bars overlap and no statement can be made.

Therefore, Eq. (5) is not obeyed and we make the assumption that the deviation is due to a correction to scaling of the form

$$R_{ag}(x,s) = R_1(x,s) + R_2(x,s) = s^{-(1+a)}f(x) + s^{-(1+c)}h(x)$$
(26)

where necessarily c > a, if the new term has to be subdominant. Having made this assumption, from Eq. (25) it follows that



FIG. 5. (Color online) Magnification of the framed portion of Fig. 4, showing that the decrease of $a_{eff}^R(x,s)$ with increasing *s* (limited to the first two sets I_s) exceeds the error bars for $x \le 2$. Symbols are the same as in Fig. 4.

$$1 + a_{eff}^{R}(x,s) = (1+a) \left(\frac{1 + [(1+c)/(1+a)]s^{-(c-a)}\kappa(x)}{1 + s^{-(c-a)}\kappa(x)} \right)$$
(27)

where

$$\kappa(x) = h(x)/f(x). \tag{28}$$

Let us now pause to explore the consequences of this formula. From Eq. (27) it follows that $a_{eff}^R(x,s) \rightarrow a$ from above as $s \rightarrow \infty$, for any fixed x. As pointed out previously, for those values of x where the size of errors is small enough, the decrease of $a_{eff}^R(x,s)$ with increasing s is confirmed by the simulations (Fig. 5). Therefore, the smallest measured value of $a_{eff}^R(x,s)$ overestimates a. Excluding the value at x=1, for the reasons which will be explained in Sec. IV B, we have

$$a \le a_{eff}^R(1^+) = 0.32 \pm 0.01$$
 (29)

where for $a_{eff}^{R}(1^{+})$ we have taken the value at $x=1^{+}$, corresponding to the intermediate I_s set. Comparing with Eq. (15), we find that this result is compatible with Eq. (13) and rules out Eq. (12). This conclusion has been reached with the sole hypothesis that a correction to scaling term needs to be taken into account. We have made no assumptions either on the form of h(x) or on the value of c, except for the obvious requirement c > a. In a short while we shall refine considerably the above estimate of a, obtaining a value much more close to the 1/4 predicted by Eq. (13).

B. Analysis of $R_1(x,s)$ and $R_2(x,s)$

Let us go further with the analysis of Eq. (27), identifying the properties of the scaling functions f(x) and h(x), which must be obeyed in order to reproduce the observed features of $a_{eff}^R(x,s)$. Notice that $R_{ag}(x,s)$, being a response function, must vanish for large x. This requires that also f(x) and h(x)must vanish for large x. Therefore, the saturation to the finite asymptotic value $a_{eff}^R(\infty,s) > a_{eff}^R(1^+,s)$, which is suggested by Fig. 4, can occur only if $\kappa(x)$ diverges for large x, that is,



FIG. 6. (Color online) Shape of the crossover curve computed from Eq. (31). The frame delimits the (x, s) region explored in the simulations.

if f(x) decays faster than h(x). This has interesting consequences. First of all we get

$$a_{eff}^R(\infty, s) = c \tag{30}$$

independent of *s*. Then, the correction to scaling contribution $R_2(x,s)$ is subdominant for fixed *x* and large *s*, but becomes dominant for fixed *s* and large *x*. Which of the two contributions $R_1(x,s)$ and $R_2(x,s)$ is dominant and which is subdominant, depends on the choice of which variable is kept fixed and which is let to grow. More precisely, the condition

$$\frac{R_1(x,s)}{R_2(x,s)} = \frac{s^{(c-a)}}{\kappa(x)} = 1$$
(31)

defines the crossover curve $\bar{x}(s)$ which divides the (s,x)plane (Fig. 6) in the two regions Γ_1 and Γ_2 , below and above $\bar{x}(s)$, where either $R_1(x,s)$ or $R_2(x,s)$ is dominant. Therefore, if in the simulations one could reach values of *s* and *t* so large as to have $\bar{x}(s) \ge 1$, together with a range of *x* extending well beyond $\bar{x}(s)$, one should observe a neat crossover from $a_{eff}^R(x,s)=a$, in a wide interval $1 < x < \bar{x}(s)$ within Γ_1 , to the large-*x* behavior $a_{eff}^R(\infty,s)=c$, after entering Γ_2 . This is visualized in the inset of Fig. 4, displaying the behavior of $a_{eff}^R(x,s)$ obtained analytically from Eq. (27), with the forms of f(x) and h(x) which will be introduced in Eqs. (32) and (45).

The next step is to focus, separately, on the two regions Γ_1 and Γ_2 .

1. Γ_1 region

We make the assumption that, with the range of *s* reached in the simulations and $x \approx 1$, we are exploring the lower boundary of Γ_1 , just above the *s* axis, where $R_2(x,s)$ is negligible with respect to $R_1(x,s)$ (see Fig. 6). The consistency of this hypothesis will be checked *a posteriori*. Let us, then, concentrate on the structure of $R_1(x,s)$. On the basis of existing analytical and numerical results, as discussed in detail in Ref. [16], the scaling function in Eq. (5) is expected to be of the general form

$$f(x,t_0/s) = A \frac{x^{-\beta}}{(x-1+t_0/s)^{\alpha}}$$
(32)

where there appears the dependence on a microscopic time t_0 , regularizing $f(x,t_0/s)$ at x=1. This extra dependence on t_0/s , for s large enough is negligible if x > 1, but becomes crucial at x=1, where it is at the root of the observed discontinuity in the effective exponent. In order to see this, let us replace Eq. (27) with the more precise form

$$1 + a_{eff}^{R}(x,s) = (1+a) \left(\frac{1 + [1/(1+a)]\tilde{f}(x,v) + [(1+c)/(1+a)]s^{-(c-a)}\kappa(x,v)}{1 + s^{-(c-a)}\kappa(x,v)} \right)$$
(33)

where $v = t_0/s$, $\kappa(x, v) = h(x)/f(x, v)$ and the new term

$$\tilde{f}(x,v) = v \frac{\partial_{v} f(x,v)}{f(x,v)}$$
(34)

comes from t_0/s in Eq. (32). A simple computation yields

$$\widetilde{f}(x,v) = \begin{cases} -\alpha \left(1 - \frac{x-1}{v}\right) & \text{for } x - 1 \ll v, \\ -\alpha \frac{v}{x-1} & \text{for } x - 1 \gg v, \end{cases}$$
(35)

which shows that this term modifies the effective exponent only at $x \approx 1$, while $\kappa(x, v)$ keeps the same properties of $\kappa(x)$, being finite for any finite *x*, including x=1, and diverging for $x \rightarrow \infty$. As a result, for *s* large but finite, we have

$$a_{eff}^{R}(x,s) = \begin{cases} a - \alpha & \text{for } x = 1, \\ a & \text{for } 1 < x \ll \overline{x}(s), \\ c & \text{for } x \gg \overline{x}(s). \end{cases}$$
(36)

In the simulations, the behavior at x=1 can be studied with great precision. Considering the definition $R_{ag}(1,s)$ $=\lim_{\Delta s\to 0}R_{ag}(1+\Delta s/s,s)$, we have computed $R_{ag}(1,s)$ as $R_{ag}(1+\epsilon/s,s)$, using Eq. (20). This quantity is very easy to simulate, since good statistics can be obtained with a modest numerical effort. The results are shown in Fig. 7, where, for completeness, we have included also the case d=3. The data show a neat algebraic decay, consistently with the assumption that $R_2(1,s)$ is negligible (if the correction was present, we should have found a crossover between two different power laws). Relating the slope to the top line of Eq. (36), we find

$$1 + a - \alpha = \begin{cases} 0.473 \pm 0.001 & \text{for } d = 2, \\ 0.477 \pm 0.001 & \text{for } d = 3. \end{cases}$$
(37)

These numbers need a comment. For such a short time difference $\chi \sim R$ and the Barrat conjecture leading to Eq. (12) is correct, i.e., in the very short time regime the response function just mirrors the density of defects. Indeed, we have measured independently the density of defects $\rho(s) \sim L^{-1}(s) \sim s^{-1/z}$, finding

$$1/z = \begin{cases} 0.474 \pm 0.001 & \text{for } d = 2, \\ 0.476 \pm 0.001 & \text{for } d = 3. \end{cases}$$
(38)

Therefore, the comparison of Eqs. (37) and (38) leads to the identification

$$1/z = 1 + a - \alpha \tag{39}$$

which is, in fact, what one finds in the exact solution of the d=1 Ising model [11,12,16] and in the approximate analytical results at higher dimensionality [13–15]. Analytical theories based on local scale invariance [24], instead, yield $\alpha=a+1$, in disagreement with our data and with the aforementioned analytical results [31].

Once the negligibility of $R_2(x,s)$ in the $x \approx 1$ region is established, the road to the direct measurement of *a* is open. This is done rewriting Eqs. (5) and (32), for $(t-s)/s \ll 1$, in the form

$$s^{1+a-\alpha}R_1(x,s) = A(t-s+t_0)^{-\alpha}$$
(40)

which predicts, using Eq. (39), that in the plot of $s^{1/z}R(x,s)$ versus t-s, the curves generated for different s ought to collapse, as long as $(t-s)/s \ll 1$ is satisfied. We have made such a plot (Fig. 8) of the data from the simulations and using the value of z in Eq. (38) we have found very good collapse for



FIG. 7. Plot of $R_{ag}(1,s)$ in d=2 and 3. Straight lines are power law best fits $t^{-0.473}$ in d=2 and $t^{-0.477}$ in d=3.



FIG. 8. (Color online) Rescaled plot of R_{ag} versus t-s for s = 537, 756, 1085, 1577. Power law best fit in the framed area $(t - s)^{-0.80}$.

 $(t-s) \leq 70$. From the slope of the curve in the region where there is collapse, we obtain $\alpha = 0.800 \pm 0.005$, from which, via Eqs. (39) and (38), we get

$$a = 0.273 \pm 0.006 \tag{41}$$

which is close enough to 1/4 to lend strong support to Eq. (13). We emphasize that this result, which has been obtained from the direct measurement of $R_{ag}(t,s)$, reproduces the previous numerical results obtained for a_{χ} from ZFC [13,16,22,28]. This is a nice check on the validity of Eq. (14), which disproves the recent claim [25] that the exponent a_{χ} is unrelated to *a*. From the same set of data we have obtained also $A=0.153\pm0.002J^{-1}$.

2. Γ_2 region

Clearly, to enter with the simulations so deep into Γ_2 that $R_1(x,s)$ can be neglected, is much more difficult. Nonetheless, the following considerations are in order. Eq. (6) requires that, for a given s, $R_{ag}(x,s)$ decays as $x^{-\lambda_R/z}$ in this sector. Therefore, the negligibility of $R_1(x,s)$ leads to the remarkable consequence that this power law behavior is a property of the correction $R_2(x,s)$ and not of the leading term $R_1(x,s)$. The implication is that the large x behaviors of h(x) and f(x) are given by

$$h(x) \sim x^{-\lambda_R/z},\tag{42}$$

$$f(x) \sim x^{-(\alpha+\beta)},\tag{43}$$

with

$$\alpha + \beta > \lambda_R / z. \tag{44}$$

This cannot be accounted for by local scale invariance [24], which assumes that the scaling function f(x) of the dominant term decays like $x^{-\lambda_R/z}$ for $x \ge 1$.

C. Fitting $R_{ag}(x,s)$

Let us recapitulate what we have done so far, specifying the assumptions that we have introduced. (1) The failure of simple scaling, displayed in Fig. 3, requires one to introduce a correction term. We have *assumed* this to have the form of the correction to scaling appearing in Eq. (26), with c > a. This is enough to derive the inequality of Eq. (29), which solves the problem of the exponent *a* formulated in Sec. II.

(2) From the two requirements, $a_{eff}^{R}(1^+,s) < a_{eff}^{R}(\infty,s) < \infty$ and $\lim_{x\to\infty} R_{ag}(x,s) = 0$, follows $a_{eff}^{R}(\infty,s) = c$ and $\lim_{x\to\infty} \kappa(x) = \infty$. This, together with the *assumption* that f(x) is of the form (32), yields the value (41) of *a* and the set of Eqs. (42)–(44) regulating the large *x* behavior of the scaling functions.

In order to complete the functional form of $R_{ag}(x,s)$, we need an additional assumption on $R_2(x,s)$, since the power law behavior (42) covers only the large x behavior. However, for $R_2(x,s)$ there are no analytical results to rely on. Short of any other hint, we make the simplest possible ansatz

$$h(x) = Bx^{-\lambda_R/z} \tag{45}$$

which continues the power law behavior (42) also into the short time region $x \approx 1$. Then, we arrive at the explicit analytical expression

$$R_{ag}(x,s) = As^{-(1+a)} \frac{x^{-\beta}}{(x-1+v)^{\alpha}} + Bs^{-(1+c)} x^{-\lambda_R/z}$$
(46)

for $R_{ag}(x,s)$ in the d=2 Ising model. We do not expect this to be quantitatively exact; rather with this formula we aim to capture the gross features of the qualitative behavior. In any case, the evaluation of the performance of this formula has to be made *a posteriori*. The program is (i) to fit the remaining free parameters in Eq. (46) from the data, (ii) to check how the predicted values for $R_{ag}(x,s)$ compare with the numerical ones, and (iii) to extrapolate to the region of x and s that have not been reached with the simulations (see Fig. 6).

Our data do not allow a direct precise determination of λ_R because, as Fig. 2 shows, the slope of $R_{ag}(x,s)$ bears a weak dependence on x even for the largest values of x reached in the simulations. Therefore, we take $\lambda_R = \lambda$ and $\lambda = 1.25$ [32] which is consistent, within errors, with the data. For the other parameters t_0 , $\beta_s \beta_s c$ we have not found a reliable direct method to measure them. We have used a four-parameter fitting routine obtaining $B=0.47J^{-1}$, c=0.81, $t_0=0.01$, $\beta=1.1$. With these parameters, and (a, α, A) determined in the previous Sec. IV B, Eq. (46) can be plotted. The comparison with the numerical data (Fig. 2) is quite good in the whole range of s and x. This provides the *a posteriori* support for the validity of the procedure.

Once f(x,v) and h(x) are given, from Eq. (33) we get the analytical form of $a_{eff}^{R}(x,s)$, which can be compared with the data of Figs. 4 and 5. However, for a meaningful comparison, rather than making a straightforward plot of Eq. (33), we have extracted the effective exponent from data generated from Eq. (46), following the same procedure described in Sec. IV A; namely, we have computed $R_{ag}(x,s)$, from Eq. (46), for the same values of s and t considered in the simulations and we have computed $a_{eff}^{R}(x,s)$ as the local slope over the same intervals I_s of four values of s appearing in Fig. 4. The result is displayed in the same Figs. 4 and 5, where, although a discrete set, these values of $a_{eff}^{R}(x,s)$ have been connected by continous lines, in order to ease the reading of the figures. The comparison shows that the behavior of $a_{eff}^{R}(x,s)$ from Eq. (46) reproduces the numerical data within errors. However, the size of the error bars prevent from making claims other than qualitative. Namely, the agreement displayed in Figs. 4 and 5, between the different ways of computing $a_{eff}^{R}(x,s)$, makes us confident that with Eq. (46) we have captured the basic mechanism underlying the behavior of $R_{ag}(x,s)$ in the d=2 Ising model. For a detailed quantitative comparison, data much more precise than those presented here are needed.

With the aim of completing the qualitative scenario, we have used Eq. (33) to plot $a_{eff}^{R}(x,s)$, in the inset of Fig. 4, over ranges of *s* and *x* that cannot be accessed in the simulations. The global behavior of $a_{eff}^{R}(x,s)$, obtained in this way, displays the typical shape of an effective exponent with the crossover taking place about $\overline{x}(s)$. For *s* sufficiently large, $a_{eff}^{R}(x,s)$ remains equal to *a* in the range $1 < x < \overline{x}(s)$, which can be enlarged at will by pushing *s* to larger values. The crossover around $\overline{x}(s)$ shows the rise of $a_{eff}^{R}(x,s)$ to switch from *a* to *c*. The growth of $\overline{x}(s)$ with *s* is rather slow and a value as large as $s \simeq 10^4$ is needed in order to have a reasonably flat behavior $a_{eff}^{R}(x,s) \simeq a$ over a sizable *x* interval is out of reach in the simulations.

V. INTEGRATED RESPONSE FUNCTIONS

In Eq. (21) we have introduced the general form of the IRF, which contains the ZFC (7) as a particular case. The aim of this section is to investigate the impact on the IRF of the structure (26) of $R_{ag}(x,s)$. In particular, we shall have to understand why the existence of such a strong correction to scaling in $R_{ag}(x,s)$ does not surface at the level of ZFC, where the simple scaling form (9) accounts very well for the data [16]. We shall not discuss the behavior of TRM because, as mentioned above and explained in detail in [16,33], this quantity is the most unfavorable choice for the analysis of the scaling properties.

Inserting Eq. (26) into the definition (21) and keeping track of the dependence on t_0 , we get

$$\mu_{ag}(y,t_w) = \mu_1(y,t_w) + \mu_2(y,t_w) = t_w^{-a}F(y,v) + t_w^{-c}H(y)$$
(47)

where

$$F(y,v) = y^{-a} \int_{1/y}^{u} z^{-(1+a)} \phi(z,v/y) dz,$$
 (48)

$$H(y) = y^{-c} \int_{1/y}^{u} z^{-(1+c)} h(1/z) dz,$$
(49)

with $y=t/t_w$, $u=\tilde{t}/t$, $v=t_0/t_w$, and

$$\phi(z, v/y) = A \frac{z^{\alpha+\beta}}{(1-z+v/y)^{\alpha}}$$
(50)

comes from Eq. (32) substituting x with 1/z. For simplicity, we have omitted to indicate explicitly the dependence on the upper limit of integration u. The effective exponent

$$\begin{aligned} a_{eff}^{\mu}(y,t_w) &= - \left. \left. \frac{\partial \ln \mu_{ag}(y,t_w)}{\partial \ln t_w} \right|_y \\ &= a \left(\frac{1 + (1/a)\widetilde{F}(y,v) + (c/a)t_w^{-(c-a)}K(y,v)}{1 + t_w^{-(c-a)}K(y,v)} \right) \end{aligned}$$
(51)

has the same structure of Eq. (33), with

$$\widetilde{F}(y,v) = v \frac{\partial_v F(y,v)}{F(y,v)}$$
(52)

and

$$K(y,v) = \frac{H(y)}{F(y,v)}$$
(53)

being the analogs of $\tilde{f}(x,v)$ and $\kappa(x,v)$.

1

There are as many IRFs as there are ways of choosing \tilde{t} . Here, we restrict our attention to the two cases with $\tilde{t}/t_w = q$ and $\tilde{t}/t=p$, where q and p are fixed numbers. It will be convenient to introduce the notation ω_{ag} and π_{ag} for these two particular IRFs. The distinction between the two cases enters the above formulas in the upper limit of integration in Eqs. (48) and (49), where

$$u = \begin{cases} q/y & \text{for } \omega_{ag}, \\ p & \text{for } \pi_{ag}. \end{cases}$$
(54)

We shall now establish the gross features of $a_{eff}^{\mu}(y,t_w)$, as y is varied with fixed t_w , using the properties of $\tilde{F}(y,v)$ and K(y,v) derived in the Appendix . The behavior of these functions depends on the sign of $\alpha - 1$. With the value of $\alpha = 0.80$ obtained from the fit of Eq. (40), we shall make use only of the results for $\alpha < 1$, leaving the general discussion to the Appendix .

A. ω_{ag}

With $y \ge \tilde{t}/t_w = q$, where $q \ge 1$ is some fixed number, y takes values in $[q, \infty)$. From Eqs. (A11) and (A13) in the Appendix it follows that

$$\widetilde{F}_{\omega}(y,v) \sim \begin{cases} (v/q)^{1-\alpha} & \text{for } y = q, \\ (v/y) & \text{for } y > q, \end{cases}$$
(55)

while, from Eqs. (A14) and (A18) it follows that $K_{\omega}(y,v)$ is finite for finite y and diverges for $y \rightarrow \infty$. Hence, the effective exponent

$$a_{eff}^{\omega}(y,t_w) = \begin{cases} a & \text{for } y \simeq q, \\ c & \text{for } y \to \infty \end{cases}$$
(56)

behaves like a_{eff}^{R} in Eq. (36), apart from the absence of any discontinuity at the minimum value of y, since Eq. (55)



FIG. 9. $\alpha_{eff}^{\omega}(y, t_w)$ for t_w in the range from 200 to 400. The continuous line is the plot from Eq. (51).

shows that $\tilde{F}_{\omega}(y,v)$ vanishes for all y as $t_w \to \infty$. This is displayed in Fig. 9, illustrating the behavior of $a_{eff}^{\omega}(y,t_w)$ obtained from the simulations with t_w in the range [200,400].

B. π_{ag}

If we keep the ratio fixed $\tilde{t}/t=p$, with $p \le 1$, y takes values in $[1/p,\infty)$. In this case we obtain an IRF in which the correction to scaling does not produce the crossover in the effective exponent from a to c, as observed in the previous case and in R_{ag} . Rather, $a_{eff}^{\pi}(y,t_w)$ starts from a and ends up again at a, as y grows, with a possible discontinuity at y=1/p, as will be explained below. The difference comes from Eq. (A24) in the Appendix , showing that now $K_{\pi}(y,v)$ vanishes as $y \to \infty$. After taking into account that also $\tilde{F}_{\pi}(y,v)$ vanishes as $y \to \infty$ [see Eq. (A5) in the Appendix], this implies

$$\lim_{y \to \infty} a_{eff}^{\pi}(y, t_w) = a.$$
(57)

Around the minimum value y=1/p, the behavior is different for p < 1 and p=1. In the former case no singularity develops, since from Eq. (A22) follows that $\tilde{F}_{\pi}(y=1/p,v) \approx v$. Instead in the latter case, which corresponds to ZFC (i.e., $\pi_{ag}=\chi_{ag}$ for p=1), from Eq. (A26) we have that at y=1 that there is a discontinuity in the effective exponent with

$$a_{eff}^{\chi}(1, v, t_w) = a - \alpha \tag{58}$$

exactly as in Eq. (36) at x=1.

The overall behavior of $a_{eff}^{\chi}(y,t_w)$ is displayed in Fig. 10 obtained by plotting Eq. (51) for fixed t_w . The two curves, corresponding to $t_w=200$ and 2000, display a very fast rise from $a - \alpha = -0.53$ at y=1 followed by a very slow approach to the asymptotic value (57). The absence of the crossover is the most prominent qualitative difference with respect to the inset of Fig. 4, which explains why the correction to scaling plays a minor role in the analysis of ZFC data. This very mild increase of $a_{eff}^{\chi}(y,t_w)$ in the range $1 < y \le 10$ is observed also in the simulations (notice the vertical scale in Fig. 11).



FIG. 10. (Color online) $\alpha_{eff}^{\chi}(y,t_w)$ plotted from Eq. (51) with $t_w = 200$ (squares) and 2000 (circles). The horizontal line corresponds to a = 0.273.

Furthermore, notice that in Fig. 10 the asymptotic value is reached from above for $t_w=200$ and from below for $t_w=2000$. The two curves cross each other, due to the interplay of the relative weights of $\tilde{F}_{\chi}(y,v)$ and $K_{\chi}(y,v)$ as t_w is varied, the former being a negative quantity and the latter a positive one. Now, the important point is that this particular feature can be resolved also in the data from the simulations (Fig. 11), showing that Eq. (46) does, indeed, account for the observed behavior of ZFC.

VI. CONCLUSIONS

In this paper we have studied numerically the linear response function of the d=2 Ising model quenched below the critical temperature. The data for R_{ag} show that the simple scaling form (5), usually believed to hold in the aging regime, is not obeyed. We have attributed the deviation from simple scaling to the existence of a correction to scaling $R_2(x,s)$, which, in order to explain the observed behavior of the effective exponent, although subdominant for fixed x and



FIG. 11. (Color online) Plot of $a_{eff}^{\chi}(y, t_w)$ from the simulations for the two sets of t_w in the legend. The lines are obtained from the analytical formula.

growing *s*, must become dominant for fixed *s* and large *x*. Then, focusing on the time regime where $R_2(x,s)$ is negligible, we have been able to measure *a* with good precision obtaining $a=0.273\pm0.006$. This solves the problem of the exponent *a* formulated in Sec. II and is in agreement with previous numerical results from the ZFC. Furthermore, this value is well consistent with Eq. (13) and confirms the dependence of *a* on dimensionality, as found in all cases where analytical results are available.

Awareness of the existence of this correction to scaling is of fundamental importance when analyzing numerical data, since an interpretation which does not take into account the strong dependence of $a_{aff}^{R}(x,s)$ on x may lead to wrong conclusions. In particular, if insisting in collapsing the data according to Eq. (5), one would find that the best data collapse obtains with an exponent whose value is somewhere between a and c, depending on the range of x considered in the simulations. However, the collapse cannot be satisfactory in the whole range of x, as Fig. 3 shows. The findings of Chatelain [17], perhaps, can be interpreted along these lines. Indeed, in Ref. [17] the impulsive response function was computed and the best data collapse was obtained with a=1/2 using data for rather small values of s. The author did explicitly notice that the collapse was not perfect and hypothesized himself the existence of strong corrections to scaling.

We have also investigated the functional form of the scaling function entering $R_1(x,s)$, and we have found that Eq. (32) with $\alpha = a + 1/z$, which was proposed in [16] as a generalization of the known analytical results, compares quite well with the numerical data. Next, by making the simplest possible ansatz for $R_2(x,s)$, we have obtained an analytical expression for the full R_{ag} . This is quite useful for the investigation of the overall qualitative behavior and compares satisfactorily well with the data in the time regions reached by the simulations.

Finally, we have considered the retrieval of the properties of the impulsive response function from the IRFs. As discussed in [16], this may be an issue to treat with care when dealing with ZFC, due to the presence of t_0 , which acts as a dangerous irrelevant variable for $\alpha \ge 1$ or, equivalently, for $d \ge d_U$. However, in this paper we are not concerned with this aspect of the problem, since d=2 is below $d_U=3$ and we have $\alpha = 0.80 < 1$. Nonetheless, recovering the scaling exponent may still be complicated, due to the presence of the correction $R_2(t,s)$. In fact, we have found that the relative importance of this term strongly depends on the kind of response function considered. In particular, it is maximal in $R_{ag}(t,s)$, while it is almost negligible for $\chi_{ag}(t,s)$. This implies that (i) by weighting differently $R_1(t,s)$ and $R_2(t,s)$, different response functions may behave very differently and (ii) the best suited function in order to weaken the correction term and to access the asymptotic properties is $\chi_{ag}(t,s)$.

The rich pattern of behaviors uncovered in this paper shows that there remains much to be understood of the response function in slow relaxation phenomena, even in the relatively simple case of coarsening systems. Presently, we do not know how general is the behavior that we have found in d=2. It would be interesting to perform simulations in higher dimension to address this point, at least numerically.

ACKNOWLEDGMENT

This work has been partially supported by MURST through PRIN-2004.

APPENDIX

In order to analyze the behavior of

$$a_{eff}^{\mu}(y,t_w) = a \left(\frac{1 + (1/a)\tilde{F}(y,v) + (c/a)t_w^{-(c-a)}K(y,v)}{1 + t_w^{-(c-a)}K(y,v)} \right)$$
(A1)

it is first necessary to study the properties of $\tilde{F}(y,v)$ and K(y,v). Rewriting Eq. (48) as

$$F(y,v) = Ay^{-a}I(\alpha) \tag{A2}$$

with

$$I(\alpha) = \int_{1/y}^{u} dz \frac{z^{\beta + \alpha - (1+\alpha)}}{(1 - z + v/y)^{\alpha}}$$
(A3)

we have

$$\partial_v F(y,v) = -A\alpha y^{-(1+\alpha)} I(\alpha+1)$$
(A4)

and

$$\widetilde{F}(y,v) = -\alpha \left(\frac{v}{y}\right) \frac{I(\alpha+1)}{I(\alpha)}.$$
(A5)

Next, carrying out the integration in (49), we can write

$$K(y,v) = \mathcal{B}\frac{y^{a-c}}{I(\alpha)} [u^{\lambda_R/z-c} - y^{c-\lambda_R/z}]$$
(A6)

where $\mathcal{B}=B/A(\lambda_R/z-c)$.

The integral $I(\alpha)$ is a finite function of v/y everywhere, except at v/y=0, where a singularity develops if $\alpha \ge 1$ and $u \rightarrow 1$. In order to see this, let us separate in $I(\alpha)$ the contribution coming from the neighborhood of the upper limit of integration, by writing $I(\alpha)=I_0+I_u$, which corresponds to the partition of the domain of integration

$$\int_{1/y}^{u} = \int_{1/y}^{u-\epsilon} + \int_{u-\epsilon}^{u}$$
(A7)

where ϵ is a small number. The first contribution I_0 is always finite, while

$$I_{u} = \int_{u-\epsilon}^{u} dz \frac{z^{\beta+\alpha-(1+\alpha)}}{(1-z+v/y)^{\alpha}} \simeq \frac{u^{\beta+\alpha-(1+\alpha)}}{1-\alpha} [(1-u+\epsilon+v/y)^{1-\alpha}] - (1-u+v/y)^{1-\alpha}]$$
(A8)

is finite for u < 1, but diverges for u=1 and $v/y \rightarrow \infty$, if $\alpha \ge 1$, yielding

$$I(\alpha) \simeq \begin{cases} <\infty & \text{for } \alpha < 1, \\ -\frac{1}{\ln(v/y)} & \text{for } \alpha = 1, \\ \frac{1}{\alpha - 1} (v/y)^{1 - \alpha} & \text{for } \alpha > 1. \end{cases}$$
(A9)

Going through similar steps, one can see that also $I(\alpha+1)$ is always finite for u < 1, while for u=1 and small v/y

$$I(\alpha+1) \simeq \frac{1}{\alpha} (v/y)^{-\alpha}.$$
 (A10)

Therefore, from Eq. (A5) and for u < 1 we have

$$\widetilde{F}(y,v) \simeq -(v/y)$$
 (A11)

while for u = 1

$$\widetilde{F}(y,v) \simeq \begin{cases} -(v/y)^{1-\alpha} & \text{for } \alpha < 1, \\ \frac{1}{\ln(v/y)} & \text{for } \alpha = 1, \\ 1-\alpha & \text{for } \alpha > 1. \end{cases}$$
(A12)

The above results hold in general. In the following we continue separately the discussion of ω_{ag} and π_{ag} .

1. ω_{ag}

With $1 < q \le y$ and u = q/y, the integration domain [1/y, q/y] in Eqs. (48) and (49) includes u = 1 only when y takes the smallest possible value y = q. Therefore, from Eq. (A12) we obtain

$$\widetilde{F}_{\omega}(q,v) \simeq \begin{cases} -(v/q)^{1-\alpha} & \text{for } \alpha < 1, \\ \frac{1}{\ln(v/q)} & \text{for } \alpha = 1, \\ 1-\alpha & \text{for } \alpha > 1, \end{cases}$$
(A13)

while from Eqs. (A6) and (A9)

$$K_{\omega}(q,v) \simeq \begin{cases} <\infty & \text{for } \alpha < 1, \\ -\frac{1}{\ln(v/q)} & \text{for } \alpha = 1, \\ (v/q)^{\alpha - 1} & \text{for } \alpha > 1. \end{cases}$$
(A14)

Inserting these results into Eq. (A1), for the effective exponent at y=q we get

$$\lim_{t_w \to \infty} a_{eff}^{\omega}(q, t_w) = \begin{cases} a & \text{for } \alpha < 1, \\ a + 1 - \alpha = 1/z & \text{for } \alpha \ge 1. \end{cases}$$
(A15)

Instead, for y > q, since $\tilde{F}(y, v)$ obeys Eq. (A11) and $K_{\omega}(y, v)$ is finite, we get

$$\lim_{t_w \to \infty} a_{eff}^{\omega}(y, t_w) = a \tag{A16}$$

independently of the sign of α -1. However, care must be used when $y \rightarrow \infty$, because the domain of integration in Eq. (A3) shrinks to zero and $I(\alpha)$ vanishes like

$$I(\alpha) \sim y^{a - (\beta + \alpha)} \tag{A17}$$

yielding

$$K_{\omega}(y,v) \sim y^{\beta+\alpha-\lambda_R/z}.$$
 (A18)

Therefore, from Eq. (44) it follows that $K_{\omega}(y,v)$ diverges for large y and, in turn, this gives the noncommutativity of the limits $\lim_{y\to\infty}$ and $\lim_{t_w\to\infty}$, since keeping t_w fixed

$$\lim_{y \to \infty} a^{\omega}_{eff}(y, t_w) = c \tag{A19}$$

while with y > q fixed

$$\lim_{t_w \to \infty} a_{eff}^{\omega}(y, t_w) = a.$$
 (A20)

2. π_{ag}

The integration domain [1/y, p] in Eqs. (48) and (49) is given by [1/y, p], with $y \ge 1/p$. Therefore, the upper limit of integration u=1 is not reached if p < 1, while it is reached if p=1. Furthermore, as y hits the lowest value y=1/p, the integration domain shrinks to zero and to first order in (y-1/p)

$$I(\alpha) = p^{2+\beta+\alpha-(1+\alpha)} [1-p(1-v)]^{-\alpha} (y-1/p)$$
 (A21)

from which follows

$$\lim_{\to 1/p} \widetilde{F}_{\pi}(y,v) = -\alpha \frac{pv}{1-p+pv}.$$
 (A22)

Notice, also, that from Eq. (A6) we get

у

$$K_{\pi}(y,v) = \frac{\mathcal{B}}{I(\alpha)} \left[p^{\lambda_R/z-c} y^{a-c} - y^{a-\lambda_R/z} \right]$$
(A23)

where both the exponents of y in the right hand side are negative. Hence, for fixed t_w

$$\lim_{y \to \infty} K_{\pi}(y, v) = 0, \qquad (A24)$$

contrary to what happens in the case of ω_{ag} , where the limit for $y \rightarrow \infty$ produces a divergence. Let us proceed by considering separately p < 1 and p=1.

a. *p* < 1

In this case u < 1 for all y and $I(\alpha)$ does not develop any singularity. From Eqs. (A11), (A22), and (A23) it then follows that

$$\lim_{t_w \to \infty} a_{eff}^{\pi}(y, t_w) = a \tag{A25}$$

for all y.

b.
$$p = 1$$

From Eq. (A22), at y=1 we have

$$\tilde{F}_{\pi}(1,v) = -\alpha \tag{A26}$$

while for y > 1 Eq. (A12) applies. Switching to the behavior of $K_{\pi}(y,v)$, notice that by inserting Eq. (A9) into Eq. (A23) we obtain

$$K_{\pi}(y,v) \simeq \begin{cases} y^{a-c} - y^{a-\lambda_{R}/z} & \text{for } \alpha < 1, \\ -\frac{y^{a-c} - y^{a-\lambda_{R}/z}}{\ln(v/q)} & \text{for } \alpha = 1, \\ (y^{a-c} - y^{a-\lambda_{R}/z})(v/q)^{\alpha-1} & \text{for } \alpha > 1. \end{cases}$$
(A27)

Using these results, for fixed y we get

 $\lim_{t_w \to \infty} a_{eff}^{\pi}(y, t_w) = \begin{cases} a - \alpha = 1/z - 1 & \text{for } y = 1, \\ a & \text{for } y > 1 \text{ and } \alpha < 1, \\ a + 1 - \alpha = 1/z & \text{for } y > 1 \text{ and } \alpha \ge 1, \end{cases}$ (A28)

while for fixed t_w

$$\lim_{y \to \infty} a_{eff}^{\pi}(y, t_w) = \begin{cases} a & \text{for } \alpha < 1, \\ a + 1 - \alpha = 1/z & \text{for } \alpha \ge 1. \end{cases}$$
(A29)

Finally, let us recall that from Eqs. (13) and (39) we have

$$\alpha - 1 = \frac{1}{z} \frac{d - d_U}{d_U - d_L} \tag{A30}$$

which shows how the sign of $\alpha - 1$ changes with the dimensionality.

- L. F. Cugliandolo, in *Trends in Theoretical Physics II*, edited by H. Falomir, R. E. Gamboa Saraví, and F. A. Schaposnik, AIP Conf. Proc. No. 484 (AIP, Woodbury, NY, 1999); G. Parisi, F. Ricci-Terzenghi, and J. J. Ruiz-Lorenzo, Eur. Phys. J. B **11**, 317 (1999).
- [2] S. Franz, M. Mézard, G. Parisi, and L. Peliti, Phys. Rev. Lett.
 81, 1758 (1998); J. Stat. Phys. 97, 459 (1999).
- [3] S. Franz and M. A. Virasoro, J. Phys. A 33, 891 (2000).
- [4] F. Corberi, E. Lippiello, and M. Zannetti, Phys. Rev. E 65, 046136 (2002).
- [5] J. P. Bouchaud, L. F. Cugliandolo, J. Kurchan, and M. Mézard in *Spin Glasses and Random Fields*, edited by A. P. Young (World Scientific, Singapore, 1997).
- [6] L. F. Cugliandolo, in *Slow Relaxations and Nonequilibrium Dynamics in Condensed Matter*, edited by J.-L. Barrat, M. V. Feigelman, J. Kurchan, and J. Dalibard, Les Houches Proceedings of the Summer School of Theoretical Physics LXXVII (Springer-Verlag, Berlin, 2004).
- [7] For a review see A. J. Bray, Adv. Phys. 43, 357 (1994).
- [8] F. Corberi, E. Lippiello, and M. Zannetti, J. Stat. Mech.: Theory Exp. (2004) P12007.
- [9] D. S. Fisher and D. A. Huse, Phys. Rev. B 38, 373 (1988).
- [10] This is no longer correct in the case of a conserved order parameter. C. Yeung, M. Rao, and R. C. Desai, Phys. Rev. E 53, 3073 (1996); C. Godreche, F. Krzakala, and F. Ricci-Tersenghi, J. Stat. Mech.: Theory Exp. (2004) P04007; C. Sire, Phys. Rev. Lett. 93, 130602 (2004); P. Calabrese and A. Gambassi, J. Phys. A 38, R133 (2005).
- [11] E. Lippiello and M. Zannetti, Phys. Rev. E 61, 3369 (2000).
- [12] C. Godrèche and J. M. Luck, J. Phys. A 33, 1151 (2000).
- [13] F. Corberi, E. Lippiello, and M. Zannetti, Eur. Phys. J. B 24, 359 (2001).
- [14] L. Berthier, J. L. Barrat, and J. Kurchan, Eur. Phys. J. B 11, 635 (1999).

- [15] G. F. Mazenko, Phys. Rev. E 69, 016114 (2004).
- [16] F. Corberi, E. Lippiello, and M. Zannetti, Phys. Rev. E 68, 046131 (2003).
- [17] C. Chatelain, J. Phys. A 36, 10739 (2003).
- [18] F. Ricci-Tersenghi, Phys. Rev. E 68, 065104(R) (2003).
- [19] E. Lippiello, F. Corberi, and M. Zannetti, e-print cond-mat/ 0407523.
- [20] S. Abriet and D. Karevsky, Eur. Phys. J. B 37, 47 (2004); 41, 79 (2004).
- [21] A. Barrat, Phys. Rev. E 57, 3629 (1998).
- [22] F. Corberi, C. Castellano, E. Lippiello, and M. Zannetti, Phys. Rev. E 70, 017103 (2004).
- [23] $d_U=2$ in the approximate calculations [14–16] for systems with scalar order parameter shows that there are quantitative features of $R_{ag}(t,s)$ beyond the reach of the best analytical methods presently available.
- [24] M. Henkel, M. Pleimling, C. Godrèche and J. M. Luck, Phys. Rev. Lett. 87, 265701 (2001); M. Henkel and M. Pleimling, *ibid.* 90, 099602 (2003).
- [25] M. Henkel, M. Paessens, and M. Pleimling, Phys. Rev. E 69, 056109 (2004).
- [26] F. Corberi, E. Lippiello, and M. Zannetti, e-print cond-mat/ 0506139.
- [27] M. Henkel and M. Pleimling, Phys. Rev. E 72, 028104 (2005).
- [28] F. Corberi, E. Lippiello, and M. Zannetti, Phys. Rev. E 63, 061506 (2001).
- [29] G. F. Mazenko, O. T. Valls, and M. Zannetti, Phys. Rev. B 38, 520 (1988).
- [30] In the first set we have omitted s=102, 103, 104 since these values are too close to s=101.
- [31] For the modifications in the above argument needed in the large N model see Sec. II A of Ref. [17].
- [32] F. Liu and G. F. Mazenko, Phys. Rev. B 44, 9185 (1991).
- [33] F. Corberi, E. Lippiello, and M. Zannetti, Phys. Rev. Lett. 90, 099601 (2003).