

Colloid–polymer mixtures in random porous media: finite size scaling and connected versus disconnected susceptibilities

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Received 11 April 2008, in final form 9 June 2008

Published 10 September 2008

Online at stacks.iop.org/JPhysCM/20/404222

Abstract

As a generic model for liquid–vapor-type transitions in random porous media, the Asakura–Oosawa model for colloid–polymer mixtures is studied in a matrix of quenched spheres using extensive Monte Carlo (MC) simulations (in $d = 3$ spatial dimensions). Since such systems at criticality, as well as in the two-phase region, exhibit a lack of self-averaging, the analysis of MC data via finite size scaling requires special care. After presenting the necessary theoretical background and the resulting subtleties of finite size scaling in random-field Ising-type systems, we present data on the order parameter distribution (and its moments) as a function of colloid and polymer fugacities for a broad range of system sizes, and for many (thousands) realizations of the porous medium. Special attention is paid to the ‘connected’ and ‘disconnected’ susceptibilities, and their respective critical behavior. We show that both susceptibilities diverge at the critical point, and we demonstrate that this is compatible with the predicted scenario of random-field Ising universality.

(Some figures in this article are in colour only in the electronic version)

1. Introduction

Understanding the behavior of fluids that undergo a liquid–vapor phase transition in the bulk (or, equivalently, of binary mixtures undergoing bulk phase separation) is still rudimentary when one considers the *confinement* of such systems in mesoporous materials, such as porous glasses or silica gels [1]. Such amorphous materials form a highly irregular, interconnected, three-dimensionally percolating network and liquid–vapor-type transitions of fluids confined between the walls of such networks are widely observed and of practical importance [2]. However, the precise nature of the liquid–vapor critical point of such systems is still only partly understood [3–14]. While de Gennes [7] has presented a simple argument that the critical behavior of fluids in such random media can be mapped onto the random-

field Ising model (RFIM) [15–17], this prediction could be confirmed neither by experiments [3–6] nor by numerical calculations (MC simulations [8, 9, 11] or density functional theories [12–14], respectively).

On the other hand, these studies could not point out any flaw in this, more than twenty year old, argument of de Gennes [7] either. In short, the argument of de Gennes starts from the well-known phenomenon of capillary condensation [1, 18, 19]. In an infinitely long slit pore, the liquid–vapor transition is ‘shifted’ relative to the bulk, due to the attractive forces between the fluid particles and the walls. The magnitude of the shift depends on the nature of the fluid and the type of walls. In addition, if ‘drying’ rather than ‘wetting’ would occur for very thick slits, the opposite effect of ‘capillary evaporation’ may also take place [1, 18, 19]. In general, the chemical potential $\mu_{\text{coex}}(D)$ where liquid and

vapor coexist inside a slit pore differs from the bulk coexistence chemical potential $\mu_{\text{coex}}(\infty)$, and this difference depends on the width D of the slit pore. In an irregular interconnected pore network, the local pore diameter at position \vec{r} fluctuates randomly around some average value. As a result, the local chemical potential $\mu(\vec{r})$, where phase coexistence would occur, will also exhibit (quenched) fluctuations around some average value (the fluctuations are quenched because the structure of the porous network does not change over time). The analogy to the RFIM is readily seen when the fluid is described as a lattice gas, since the latter is isomorphic to the Ising ferromagnet. In terms of the Ising ferromagnet, quenched random fluctuations in the local chemical potential become isomorphic to quenched external field variables h_i , with h_i a random variable acting on the spin at the i th lattice site (which is precisely the random-field Ising model). This reasoning is also easily carried over to binary fluid mixtures [7].

In the present work, we contribute to the clarification of this problem by presenting extensive MC data for a particularly simple model, namely the Asakura–Oosawa (AO) model [20, 21] of colloid–polymer mixtures inside a quenched random porous medium. The AO model is known to capture *bulk* experimental observations very well (by bulk we mean in the absence of any porous medium), including phase separation [22, 23] and interfacial properties [24]. Computer simulations [25–27] have shown that bulk phase separation in the AO model, which occurs for sufficiently large polymers at sufficiently high polymer fugacity, belongs to the universality class of the Ising model [28–31]. In addition, the standard predictions for capillary condensation in slit pores [18, 19, 32] have been well confirmed for this model [33–36]. Following our previous work [37, 38], we now consider the more complex problem of the AO model inside a random porous medium. The porous medium is obtained using an ‘easy’ recipe: we simply distribute a set of obstacles (spheres) at random positions in the simulation box. Once the spheres have been positioned, they remain ‘fixed’, to mimic the quenched nature of the medium. Next, the AO model is inserted into the medium and its phase behavior is studied. In particular, we will focus on (appropriately constructed) ‘susceptibilities’ of the form $[\langle \cdot \rangle^2] - [\langle \cdot \rangle]^2$, with $\langle \cdot \rangle$ the conventional Gibbs–Boltzmann thermal average and $[\cdot]$ an average over many different realizations of the quenched obstacles. Of course, in the *absence* of the porous medium, any such ‘susceptibility’ is trivially zero. However, in its *presence*, the analogy to the random-field Ising model implies that such quantities will actually diverge at the critical point and will do so with a characteristic critical exponent $\bar{\nu}$.

The outline of our paper is as follows. In section 2, we recall in detail the necessary background of finite size scaling in the Ising and the random-field Ising models, and we discuss how these techniques may be carried over to fluids with quenched disorder. Next, in section 3, we define the AO model, explain how this model may be extended to also capture quenched disorder, and we describe our simulation method. The results are presented in section 4 and we end with a discussion, conclusion and summary in section 5.

2. Finite size scaling in the Ising model, the random-field Ising model and related models

2.1. Ising model

We first consider the pure Ising model, i.e. without any random field, and discuss how finite size scaling can be used to extract the critical properties of this model. To be specific, we consider a (nearest-neighbor) Ising ferromagnet on a hypercubic d -dimensional lattice of linear dimension L and periodic boundary conditions

$$\mathcal{H}_{\text{Ising}} = -J \sum_{\langle i,j \rangle} s_i s_j - H \sum_i s_i, \quad s_i = \pm 1, \quad (1)$$

with J the exchange constant and H an uniform external magnetic field. Defining the instantaneous magnetization per spin s as

$$s = \frac{1}{L^d} \sum_i s_i, \quad (2)$$

the object of interest is essentially the distribution

$$P_L(s) \equiv P_L(s|T, H), \quad (3)$$

defined as the probability to observe a magnetization per spin s , in a system of size L , at temperature T and field strength H . Basic observables of interest follow from the moments $\langle s^k \rangle \equiv \int_{-\infty}^{+\infty} s^k P_L(s) ds$ of the distribution, where $\langle \cdot \rangle$ is a conventional Gibbs–Boltzmann thermal average. For instance, the average magnetization per spin can be written as

$$m(T, H) = \langle s \rangle, \quad (4a)$$

and for the susceptibility χ we obtain

$$\chi(T, H) = \frac{\partial m}{\partial H} = L^d (\langle s^2 \rangle - \langle s \rangle^2), \quad (4b)$$

where the factor $k_B T$ has been absorbed in the definition of χ , with k_B the Boltzmann constant. We emphasize that these expressions must be used with care when $P_L(s)$ is bimodal. This happens, for example, at low temperature and $H = 0$, since then a spontaneous magnetization exists, which may be positive or negative. Consequently, $P_L(s)$ has two peaks, one at positive and one at negative values. However, blindly applying equation (4a), one finds that $m(T, 0) = 0$, irrespective of T , which is not really desirable. Since the Ising model has spin reversal symmetry, we have $P_L(-s) = P_L(+s)$, and so an easy fix is to introduce

$$m'(T, H) = \langle |s| \rangle, \quad (5a)$$

$$\chi'(T, H) = L^d (\langle s^2 \rangle - \langle |s| \rangle^2), \quad (5b)$$

which are to replace m and χ in these cases. The absolute value has the same effect as using a modified distribution $(P_L(s) + P_L(-s))/2$ with the integration domain restricted to $s > 0$. Clearly, such a modification is reasonable when $P_L(-s) = P_L(+s)$ somewhat holds. For very asymmetric distributions, a safer approach is to define m and χ in terms of peak positions and widths, respectively. This approach was

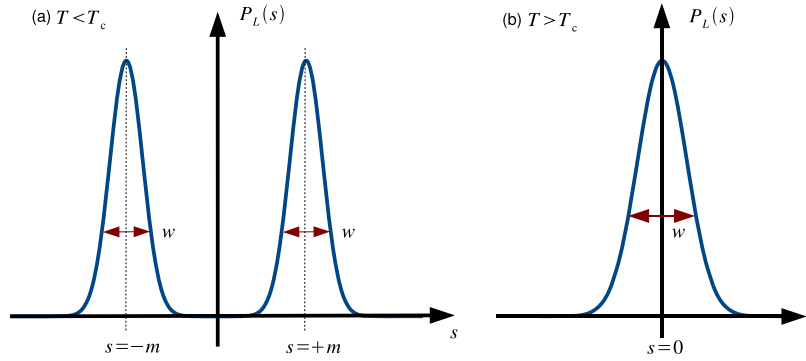


Figure 1. Schematic representations of the distribution $P_L(s)$ below the critical temperature T_c (a) and above (b). Since the susceptibility for $T \neq T_c$ is finite, the peak widths w vanish with increasing system size $\propto L^{-d/2}$, leading to a distribution featuring two δ peaks when $T < T_c$ and a single δ peak when $T > T_c$.

successfully applied to the AO model in [27] and will also be used in this work later on. Of course, in the thermodynamic limit, all definitions become equivalent, see the discussion in [39].

As is well known, for $d \geq 2$, the Ising model has a second-order phase transition from the (disordered) high-temperature paramagnetic phase to the (ordered) low-temperature ferromagnetic phase, at some critical temperature T_c . In the vicinity of T_c , we expect power law singularities [28]:

$$m(T, 0) \propto (-t)^\beta \quad (\text{order parameter}), \quad (6a)$$

$$\chi(T, 0) \propto |t|^{-\gamma}, \quad (6b)$$

$$\xi \propto |t|^{-\nu}, \quad (6c)$$

with $t \equiv T/T_c - 1$ the reduced distance from the critical point and ξ the correlation length of the magnetization fluctuations. In the above, β , γ and ν are critical exponents, which characterize the universality class.

Of course, the divergence of the correlation length cannot be captured in a finite simulation box of size L , and so the above power laws are never observed directly. The state-of-the-art is to perform several simulations, using a range of system sizes L , and to extrapolate the simulation data to $L \rightarrow \infty$ via finite size scaling. In its simplest form, finite size scaling is just the statement that, in a finite system at the critical point, $\xi \propto L$ [28]. Eliminating t from equations (6a) and (6c), and using $\xi \propto L$, one immediately derives the L dependence of the magnetization order parameter at the critical point:

$$m_L \propto L^{-\beta/\nu}. \quad (7a)$$

Similarly, for the susceptibility, one obtains

$$\chi_L \propto L^{\gamma/\nu}. \quad (7b)$$

These equations simply state that, if one performs a simulation at the critical point over a range of system sizes, the magnetization should vanish $\propto L^{-\beta/\nu}$ and the susceptibility should increase $\propto L^{\gamma/\nu}$.

The above scaling laws are quite general and should hold near any critical point where the correlation length diverges

as a power law, i.e. conforms to equation (6c). If also the hyperscaling relation is obeyed

$$\gamma + 2\beta = \nu d, \quad (8)$$

with d the spatial dimension, it follows that the entire distribution $P_L(s)$ at $H = 0$ scales with L as [40]

$$P_L(s)|_{H=0} = L^{\beta/\nu} \tilde{p}(L/\xi, sL^{\beta/\nu}), \quad (9)$$

with $P_L(s)$ the magnetization distribution of equation (3). Here, $\tilde{p}(x, x')$ is a universal scaling function, which essentially depends on the universality class, and the scaling should hold in the limits $\xi \rightarrow \infty$, $L \rightarrow \infty$, with L/ξ finite. Using equation (9), one readily obtains the moments

$$\langle |s| \rangle = \int |s| P_L(s) ds = L^{-\beta/\nu} \tilde{f}_0(L/\xi), \quad (10a)$$

$$\langle s^k \rangle = \int s^k P_L(s) ds = L^{-k\beta/\nu} \tilde{f}_k(L/\xi) \quad (k > 0), \quad (10b)$$

which also define the scaling functions \tilde{f} . For $k = 1$, one recovers the average magnetization of equation (4a) and the expected scaling law equation (7a) is correctly reproduced. For the susceptibility, however, these moments imply $\chi \propto L^{d-2\beta/\nu}$, consistent with equation (7b) only when hyperscaling holds. For the Ising model, hyperscaling indeed holds [28], and so the use of equation (9) is justified here.

Hyperscaling also implies a remarkable property concerning the *shape* of $P_L(s)$ at criticality. To see this, note first that, below T_c , there exists a spontaneous magnetization. The magnetization may be positive (+) or negative (-), and so $P_L(s)$ features two peaks centered around $s = \pm m$, see figure 1(a). Each of the peaks may be approximated by a Gaussian [40]:

$$P_L^\pm(s) \approx L^{d/2} (2\pi T \chi)^{-1/2} \exp[-(s \mp m)^2 L^d / (2T \chi)], \quad (11)$$

leading to a squared peak width $w^2 \equiv \langle s^2 \rangle - \langle s \rangle^2 = \chi T / L^d$. In the thermodynamic limit, the peaks remain at their respective positions $s \pm m$. At the same time, since the susceptibility away from T_c is finite, the peaks also become increasingly narrow, eventually converging to a distribution consisting of

two δ peaks. The behavior in the disordered region $T > T_c$ follows analogously. In this case, a spontaneous magnetization is absent and $P_L(s)$ is just a single Gaussian, centered around $s = 0$, see figure 1(b). For $T > T_c$, $P_L(s)$ thus converges to a single δ peak in the thermodynamic limit. Mathematically, the shape of the distribution in these two limiting cases can be expressed using the cumulant $U_1 = \langle s^2 \rangle / \langle |s| \rangle^2$.⁴ After some algebra, one finds that

$$\lim_{L \rightarrow \infty, T < T_c} U_1 = 1 \quad (\text{two } \delta - \text{peaks}), \quad (12a)$$

$$\lim_{L \rightarrow \infty, T > T_c} U_1 = \pi/2 \quad (\text{one } \delta - \text{peak}). \quad (12b)$$

Precisely at T_c , the behavior of the cumulant is more subtle [40]. In a finite system at T_c , the magnetization vanishes $\propto L^{-\beta/\nu}$, see equation (7a). Hence, $P_L(s)$ still exhibits two peaks, at positions $\propto \pm L^{-\beta/\nu}$. For the root-mean-square peak width, we obtain $w^2 \equiv \langle s^2 \rangle - \langle s \rangle^2 \propto L^{\gamma/\nu-d}$, where now equation (7b) was used. Comparing the *distance* between the peaks to their *widths*, we find

$$\Delta \equiv \frac{\text{peak width}}{\text{peak-to-peak distance}} \propto L^\omega, \quad (13)$$

with $\omega = (\gamma/\nu - d)/2 + \beta/\nu$. By virtue of hyperscaling one has $\omega = 0$, implying that the relative peak width Δ does not vanish in the thermodynamic limit. Consequently, $P_L(s)$ at criticality does *not* become a superposition of two δ functions but instead converges to a distribution of two overlapping peaks. The cumulant U_1^* at T_c differs therefore from the off-critical values [40]. By using equations (10a) and (10b), U_1^* can be expressed in terms of the scaling functions as $U_1^* = \tilde{f}_2(L/\xi) / \tilde{f}_0^2(L/\xi)$, which is a universal function of L/ξ , and tends to a universal finite constant. In simulations, this result is useful since plots of U_1 versus T for various system sizes L will show a common intersection point, yielding an estimate of both U_1^* and T_c (cumulant intersection method [40]).

2.2. Random-field Ising model: finite size scaling

The analysis of MC simulation [39, 41] data for systems belonging to the universality class of the random-field Ising model (RFIM) [17] has certain subtleties [42, 43] when one tries to apply finite size scaling methods [28, 40]. The source of the problem is that the standard hyperscaling relation [28] between critical exponents, which is required by equation (9) [40], does not hold for the RFIM [17, 44–47]. Since the presence of the random field breaks the spin reversal symmetry, it is necessary to consider also the ‘disconnected’ susceptibility χ_{dis} [45], in addition to the standard ‘connected’ susceptibility χ .

The RFIM Hamiltonian is

$$\mathcal{H}_{\text{RFIM}} = -J \sum_{\langle i,j \rangle} s_i s_j - H \sum_i s_i - \sum_i h_i s_i, \quad (14a)$$

$$s_i = \pm 1, \quad (14a)$$

with J and H defined as before. In addition, at each lattice site i , there acts a quenched random field h_i , which we take to be

completely uncorrelated between neighboring sites, and with an average of zero:

$$h_i = \pm h, \quad \langle h_i \rangle = 0. \quad (14b)$$

The amplitude h of the random field should be small but finite ($h/J \ll 1$), but we are not concerned with the crossover to the pure Ising model here, and hence disregard the limit $h \rightarrow 0$. Basic observables are again the average magnetization per spin m , the connected susceptibility χ and the disconnected susceptibility χ_{dis} :

$$m(T, H) = \langle \{s\} \rangle, \quad (15a)$$

$$\chi(T, H) = L^d [\langle s^2 \rangle - \langle s \rangle^2], \quad (15b)$$

$$\chi_{\text{dis}}(T, H) = L^d [\langle s \rangle^2]. \quad (15c)$$

For the same reason as before, we also introduce

$$m'(T, H) = \langle \{|s|\} \rangle, \quad \chi'(T, H) = L^d [\langle s^2 \rangle - \langle |s| \rangle^2]. \quad (16)$$

For the RFIM model, one has to perform the standard Gibbs–Boltzmann thermal average $\langle \cdot \rangle$ for one realization of the random field, followed by an average over M different random-field configurations $[\cdot]$, whereby M should be large. Note that χ_{dis} is simply the fluctuation of the average magnetization $\langle s \rangle$ between different realizations of the random field. Due to random variations in these fields, $\langle s \rangle$ will sometimes be negative and sometimes be positive. In the limit $M \rightarrow \infty$, one has $\langle \{s\} \rangle = 0$, of course, but the fluctuation $[\langle s \rangle^2] - \langle \{s\} \rangle^2$ will generally *not* be zero, which is essentially what χ_{dis} corresponds to. We shall also be interested in the distributions

$$P_{L,i}(s) \equiv P_{L,i}(s|T, H_i), \quad (i = 1, \dots, M), \quad (17)$$

defined as the probability to observe a magnetization per spin s , in a system of size L , at temperature T and external field H_i , for the i th random-field realization. Note that we allow H_i to vary between different random-field realizations. Ideally, one would like to have $M \rightarrow \infty$, but since resources are limited, simulations always deal with finite M .

Assuming that the RFIM, for small enough h , has a second-order phase transition at T_c , we expect power law singularities for m , χ and ξ as before, but with different critical exponents characteristic of the RFIM universality class [17, 45]. In addition, a power law is expected for the disconnected susceptibility

$$\chi_{\text{dis}} \propto |t|^{-\bar{\gamma}}, \quad (18)$$

with a new critical exponent $\bar{\gamma}$ [17, 45]. It has been proved rigorously that the RFIM in $d = 3$ dimensions, at low enough temperature, indeed exhibits a non-zero spontaneous magnetization [16]. It has not, however, been proved that the second-order transition assumed above actually exists (also weak first-order transitions [48] or spin glass type phases [49] have been suggested). Recent MC simulations, however, favor a second-order transition, albeit that the critical exponents are still not known very accurately [50, 51].

⁴ Of course, the cumulant is to be calculated for the full distribution. In particular, for $T < T_c$, one should write $P_L(s) = (P_L^- + P_L^+)/2$.

While for the pure Ising model we have the standard hyperscaling relation between critical exponents [28], for the RFIM, rather a different relation has been proposed [44]

$$\gamma + 2\beta = \nu(d - \theta). \quad (19)$$

Here, θ is an exponent which measures the deviation from the standard hyperscaling relation; when it is zero, standard hyperscaling is again recovered. Using the further result that $\theta = \gamma/\nu$ [45], it follows that $(\gamma + \beta)/\nu = d/2$.

We now discuss finite size scaling in the RFIM, following Eichhorn and Binder [42, 43]. Note first that the ‘derivation’ of equations (7a) and (7b) still holds. Hence, m and χ scale with L as before, albeit with different exponents. Similarly, for the scaling of the disconnected susceptibility at T_c , we expect that

$$\chi_{L,\text{dis}} \propto L^{\tilde{\gamma}/\nu}. \quad (20)$$

If we assume that each distribution $P_{L,i}(s)$ scales conforming to equation (9), it follows that $\langle |s| \rangle_i$ and $\langle s^k \rangle_i$ scale according to equations (10a) and (10b), respectively (the subscript denotes that the thermal average was taken in the i th random-field realization). Of course, the scaling functions \tilde{f} may depend on the particular random-field realization, but the leading L dependence will be the same each time. Since by definition $\langle X \rangle \equiv (1/M) \sum_{i=1}^M \langle X \rangle_i$, it follows trivially that the L dependence implied by equations (10a) and (10b) appears in the quenched average also. We thus obtain

$$\begin{aligned} \langle |s|^2 \rangle &= \hat{c}_0 L^{-2\beta/\nu}, & \langle s^2 \rangle &= \hat{c}_1 L^{-2\beta/\nu}, \\ \langle s^2 \rangle &= \hat{c}_2 L^{-2\beta/\nu}, \end{aligned} \quad (21)$$

with redefined scaling functions \hat{c} , which can be expressed in terms of the functions \tilde{f} , of course, but for our subsequent discussion the precise form does not matter. Using the definitions of χ , χ' and χ_{dis} the above equation implies

$$\begin{aligned} \chi &= (\hat{c}_2 - \hat{c}_1) L^{d-2\beta/\nu}, & \chi' &= (\hat{c}_2 - \hat{c}_0) L^{d-2\beta/\nu}, \\ \chi_{\text{dis}} &= \hat{c}_1 L^{d-2\beta/\nu}. \end{aligned} \quad (22)$$

On the other hand, finite size scaling also demands that $\chi \propto \chi' \propto L^{\gamma/\nu}$ and $\chi_{L,\text{dis}} \propto L^{\tilde{\gamma}/\nu}$. The solution of the paradox is to require that

$$\tilde{\gamma} + 2\beta = \nu d, \quad (23)$$

which correctly sets the scaling of χ_{dis} , and also that $\hat{c}_0 = \hat{c}_2$ and $\hat{c}_1 = \hat{c}_2$. Note that equation (23) is just the standard hyperscaling relation, but with γ replaced by $\tilde{\gamma}$. Hence, even though normal hyperscaling in the RFIM does not hold, equation (9) still gives a consistent description of finite size scaling, but one must accept that the connected susceptibility is not described by it, since the leading terms in χ and χ' cancel. To also describe the scaling of χ and χ' , one needs to include the leading correction to scaling. This correction can be derived by assuming that $P_{L,i}(s)$ at and below T_c is a superposition of two Gaussians. Expressing the peak at positive magnetization as $P_{L,i}^+(s) \propto \exp(-(s - m_i)^2/(2w_i^2))$, it follows that $\langle s \rangle_i = m_i$ and $\langle s^2 \rangle_i = m_i^2 + w_i^2$. Performing the quenched average, we now obtain a non-zero expression for the

connected susceptibility $\chi = (L^d/M) \sum_{i=1}^M w_i^2$. This term, consequently, is the sought-for correction; finite size scaling then implies that $w_i^2 \propto L^{\gamma/\nu-d}$ at criticality.

Since $\hat{c}_0 = \hat{c}_2$, it also follows that the cumulant at criticality $U_1^* \equiv [\langle s^2 \rangle]/[\langle |s| \rangle^2]$ in the RFIM tends to unity [42, 43]. The *shape* of the quenched-averaged distribution at T_c is therefore similar to that *below* T_c : both distributions are characterized by $U_1 = 1$ in the thermodynamic limit. Hence, also at T_c , we have a distribution featuring two δ peaks⁵. This is profoundly different from systems where hyperscaling holds, since here U_1^* tends to a non-trivial value different from the off-critical values (as explained in section 2.1). For the RFIM, plots of U_1 versus T , for various system sizes $L \rightarrow \infty$, no longer intersect. In practice, however, the system sizes feasible in simulations are still quite small, and so one is plagued by ‘crossover’ effects [52] (in this case from Ising to RFIM universality). This means that an intersection point can typically still be identified, but it occurs at a value much closer to $U_1^* = 1$ of the RFIM [37, 43, 50]⁶. The fact that the quenched-averaged distribution in the RFIM remains sharp at criticality, in contrast to overlapping, is also obvious from equation (23). Considering again the ratio Δ between peak width and peak-to-peak distance, i.e. conforming to equation (13), one finds that $\omega = (\gamma - \tilde{\gamma})/(2\nu)$. Using the result of Schwartz that $\tilde{\gamma} = 2\gamma$ [45], it immediately follows that $\omega < 0$. In other words, for the RFIM at its critical point, the relative peak width Δ vanishes, leading to a distribution featuring two δ peaks.

2.3. Random-field Ising model: sample-to-sample fluctuations

The result of Schwartz [45], namely that $\tilde{\gamma} = 2\gamma$, can be made plausible when we consider one particular realization of the random field. In a volume L^d , roughly half the lattice sites ‘feel’ a negative random field (and the other half a positive random field, obviously) but with Poissonian fluctuations. Hence, there will typically be an excess Zeeman energy of order $\pm hL^{d/2}$, which has the same physical effect as if a uniform external field of strength

$$H_c \sim \pm hL^{-d/2}, \quad (24)$$

acted on the spins in this volume (recall that h is the strength of the random field). But then we expect a non-zero magnetization $\langle s \rangle = H_c \chi \sim \chi hL^{-d/2}$ in this sample, with χ the connected susceptibility. Using near T_c the standard finite size scaling relations for $\langle s \rangle$ and χ , we obtain $L^{-\beta/\nu} \propto L^{\gamma/\nu-d/2}$ or $\beta/\nu = d/2 - \gamma/\nu$. Combining with equation (23) one finds that $\tilde{\gamma} = 2\gamma$.

It is of some interest to explore the consequences of equation (24) further and study the behavior of the magnetization for different realizations of the random field. If $H_c > 0$ and $T > T_c$, we have for $H = 0$ a positive magnetization of order $\langle s \rangle \sim \chi hL^{-d/2}$ as argued above (recall that H is the strength of the uniform external field). The field

⁵ Of course, whereas for $T = T_c$ the peak positions scale $\propto L^{-\beta/\nu}$, they saturate at finite values $\pm m$ when $T < T_c$.

⁶ Note in particular figure 7 of [50] for the RFIM.

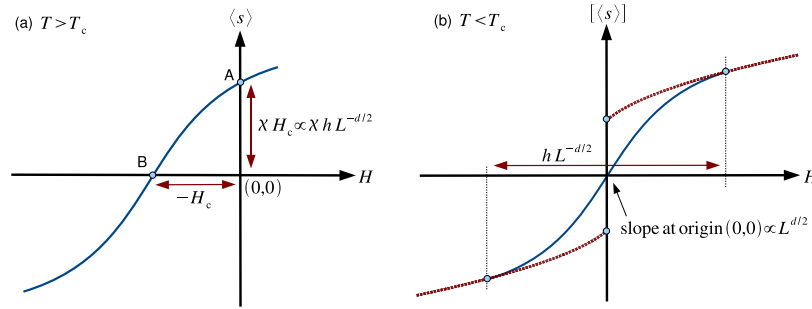


Figure 2. (a) $\langle s \rangle$ versus H at $T > T_c$, for one realization of the random field where H_c happens to be positive. At $H = 0$, we have a finite magnetization of order $\chi h L^{-d/2}$ (point A). The field at which $\langle s \rangle$ changes sign, and where $\chi \propto \partial \langle s \rangle / \partial H$ attains its maximum, occurs at $H = -H_c$ (point B). Note that the slope at B approaches the zero-field connected susceptibility $\chi|_{H=0}$ in the limit $L \rightarrow \infty$. (b) $[\langle s \rangle]$ versus H at $T < T_c$. The dashed curve shows the behavior in the thermodynamic limit; the solid curve in a finite system of size L . Note that the rounding is of order $L^{-d/2}$. This means that the slope $\partial[\langle s \rangle] / \partial H|_{H=0}$ in finite systems grows $\propto L^{d/2}$ and that the region where $[\langle s \rangle]$ deviates significantly from $L \rightarrow \infty$ behavior shrinks $\propto L^{-d/2}$.

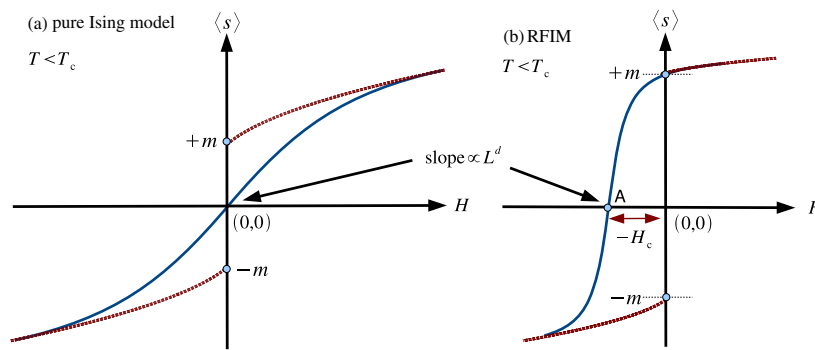


Figure 3. $\langle s \rangle$ versus H at $T < T_c$ for the pure Ising model (a) and the RFIM (b). The dashed curves show the behavior in the thermodynamic limit; solid curves for finite systems (see details in the text). Note that scenario (b) holds only in spatial dimension $d > 2$.

at which the susceptibility $\chi \propto \partial \langle s \rangle / \partial H$ is maximized is therefore not $H = 0$, but rather $H = -H_c$, where the net effect of the random field is canceled. This is sketched in figure 2(a), where $\langle s \rangle$ versus H is plotted⁷. Taking the limit $L \rightarrow \infty$, it follows that the *slope* of the curve at points A and B becomes the same, since, on the small scale of H_c , the curve may be approximated by a straight line. Note that the slope approaches the zero-field susceptibility χ , and also that the slope is independent of H_c . Plotting $\langle s \rangle$ versus H for different realizations of the random field, one thus obtains a set of parallel straight lines. In other words, χ is rather insensitive to the particular random-field configuration, which just expresses the fact that the system is self-averaging for $T > T_c$, as expected [53, 54]. This result is important since, in systems lacking spin reversal symmetry, the natural path in the (T, H) plane to follow is no longer the line $H = 0$, but rather the path along which χ assumes its maximum for each realization of the random field.

The situation is qualitatively different for $T < T_c$, of course, since we now expect a first-order transition and, consequently, two-phase coexistence. In the pure Ising model, coexistence between two states (with positive and negative spontaneous magnetization $\pm m$) occurs at $H = 0$, irrespective of the system size L . In the thermodynamic limit, $\langle s \rangle$ increases

monotonically with H and jumps from $-m$ to $+m$ at $H = 0$; see the dotted curve in figure 3(a). In a finite system, the transition is rounded and a true jump does not appear. Instead, $\langle s \rangle$ passes smoothly through the origin, but with slope $\partial \langle s \rangle / \partial H \propto L^d$ [55, 56], consistent with the formation of a jump in the thermodynamic limit; see the full curve in figure 3(a). Consequently, phase coexistence in the pure Ising model may always be studied using $H = 0$. Provided T is sufficiently below T_c , double-peaked distributions $P_L(s)$ are readily observed, i.e. conform to figure 1(a), from which the coexistence properties follow.

For the RFIM below T_c and finite system size L , the behavior is more subtle, since the random field breaks the spin reversal symmetry. We still expect a (rounded) first-order transition, but centered around the shifted field $H = -H_c$, see figure 3(b). In the thermodynamic limit, $H_c \rightarrow 0$, and so the magnetization jumps, as before, at $H = 0$ (dotted curve). In a finite system $\langle s \rangle$ increases smoothly with H (solid curve), passing through zero at $H = -H_c$ (point A), with slope $\partial \langle s \rangle / \partial H \propto L^d$. Since $H_c \propto h L^{-d/2}$ asymptotically exceeds the rounding, it follows that, in a finite system at $H = 0$, phase coexistence is unlikely. At $H = 0$, one either observes the phase with positive magnetization (as one would in figure 3(b)) or, if the random field happens to resemble $H_c < 0$, a negative magnetization. Only very rarely, when the inflection point A happens to coincide with $H = 0$, will both phases be observed simultaneously. Hence, at $H = 0$, the

⁷ Of course, the graph of $[\langle s \rangle]$ versus H is antisymmetric about the origin, since in the quenched average both signs of H_c appear equally often.

distribution $P_{L,i}(s)$ will mostly feature just one peak, located at positive or negative values. In the quenched average, one recovers $[\langle s \rangle] = 0$, of course, but the fluctuation $[\langle s \rangle^2] - [\langle s \rangle]^2$ is not zero, since this, apart from a factor L^d , is precisely the disconnected susceptibility, see equation (15c). Clearly, to study phase coexistence in simulations, it does not make sense to use $H = 0$, since one would rarely see a double-peaked distribution. Instead, it is more meaningful to obtain these properties at the inflection point A, where $\chi \propto \partial \langle s \rangle / \partial H$ attains its maximum. To be precise: one should apply an external field $H = -H_c$ ‘tailored’ for each random-field realization. In the limit $L \rightarrow \infty$, one has $H_c \rightarrow 0$ and coexistence properties obtained at the inflection point will agree with those obtained at $H = 0$. The advantage of the former method being that double-peaked distributions $P_{L,i}(s)$ will now already appear in much smaller systems. Of course, for these double-peaked distributions, $\langle s \rangle_i$ will be close to zero each time, and so it follows that a different definition for the disconnected susceptibility should be used, presumably of the form $\chi'_{\text{dis}} = L^d [\langle |s|^2 \rangle]$.

Considering now the behavior of $[\langle s \rangle]$ versus H below T_c , we expect the scenario of figure 2(b). In the thermodynamic limit, a jump in $[\langle s \rangle]$ at $H = 0$ is anticipated. In finite systems, the jump is rounded, but on a more severe scale $L^{-d/2}$, as pointed out by Kierlik *et al* [14]. Note that graphs of $[\langle s \rangle]$ versus H for finite L intersect the origin since, in the quenched average, both signs of H_c are equally likely.

Unfortunately, these arguments cannot be easily extended to $T = T_c$. As discussed in detail by Wiseman and Domany [53, 54], systems with quenched random disorder at criticality exhibit lack of self-averaging. For small enough fields H , it still holds that $\langle s \rangle$ versus H for one realization of the random field is a straight line, with slope $\propto L^{\gamma/\nu}$. The same holds for $[\langle s \rangle]$ versus H , where the slope is also $\propto L^{\gamma/\nu}$, but the prefactors differ. The ratio of these prefactors is a quantity characterizing the lack of self-averaging, in the sense of Wiseman and Domany [53, 54].

2.4. Obtaining the quenched average using a sample-dependent H_i

For the Ising model, one knows beforehand that the inflection point of $\langle s \rangle$ versus H (or $[\langle s \rangle]$ versus H in the case of the RFIM), occurs on the symmetry line $H = 0$. Hence, varying T at fixed $H = 0$, one cannot miss the critical point. In less symmetric models, the field H at the inflection point is not known beforehand. In these cases, it is clearly more convenient to follow the path $H = -H_c(T)$ in the (T, H) plane of each random-field realization. That is, for each realization of the random field i , one numerically locates the field H_i where $\partial \langle s \rangle_i / \partial H$ in that sample is maximized. Properties of interest are then collected at H_i and the process is repeated over many different random-field samples. Extrapolating the data to $L \rightarrow \infty$ is demanding in practice, but does not present any principal objections. Only the *prefactors* of the finite size scaling laws at criticality

$$\begin{aligned} \overline{\langle s \rangle} &\propto L^{-\beta/\nu}, & \overline{\langle s^2 \rangle - \langle s \rangle^2} &\propto L^{\gamma/\nu-d}, \\ \overline{\langle s \rangle^2} &\propto L^{\tilde{\gamma}/\nu-d}, \end{aligned} \quad (25)$$

will differ from those of the standard quenched average $[\cdot]$ obtained at fixed H . Here, the overbar denotes averaging at the sample-dependent H_i . Since many typical fluids (including the AO model) are asymmetric, collecting the quenched average as $\overline{\langle \cdot \rangle}$ simply becomes a necessity in these cases.

2.5. Extension to fluids

We now consider a liquid–vapor transition of a fluid confined to a quenched porous medium. We use the grand canonical (GC) ensemble, i.e. volume L^d , temperature T and chemical potential μ are fixed, but the number of particles N in the system fluctuates. Our analysis is based on the (normalized) distribution

$$P_{L,i}(N) \equiv P_{L,i}(N|T, \mu_i), \quad (i = 1, \dots, M), \quad (26)$$

defined as the probability to observe a system containing N particles, in the i th realization of the porous medium. Note the dependence on L and T , and also that we allow the chemical potential μ_i to vary between different realizations of the porous medium. For given L and T , $P_{L,i}(N)$ is sampled from $N = 0$ to N_{max} , using a biased sampling scheme [57]. This process is repeated for M different realizations of the porous medium. The sampling scheme is constructed to visit the full range $0 \leq N \leq N_{\text{max}}$, irrespective of the imposed chemical potential. Hence, we set $\mu_i = 0$ in the simulations and use histogram reweighting [58] to extrapolate to different values afterward.

Assuming that the liquid–vapor transition in a porous medium belongs to the universality class of the RFIM, we expect, in the thermodynamic limit, a critical point at temperature T_c and chemical potential μ_{cr} . Below T_c , we anticipate bimodal distributions $P_{L,i}(N)$, but only if μ_i is chosen reasonably. In finite systems, we actually expect the bimodal form to persist considerably above T_c also since, for the RFIM, $P_{L,i}(N)$ remains sharp at criticality. In this work, μ_i is tuned for each realization of the porous medium such that $\partial \langle N \rangle_i / \partial \mu_i$ for that realization is maximized, with $\langle N \rangle_i = \sum_{N=0}^{\infty} N P_{L,i}(N)$ ⁸. Loosely identifying $\langle N \rangle_i \leftrightarrow \langle s \rangle$, $\mu_i \leftrightarrow H$, our choice may be regarded as the analog of the inflection point A in figure 3(b) for the magnetic case.

Figure 4 shows a schematic of $P_{L,i}(N)$ in bimodal form. The left peak represents the vapor, the right peak the liquid, with the average $\langle N \rangle_i$ of the full distribution located in between (arrow A). Figure 4 also shows that N_{max} should be chosen well beyond the liquid peak (arrow B). If we shift $P_{L,i}(N)$ by its average, we approximately recover the Ising symmetry $P_L(s) = P_L(-s)$ of the magnetization distribution. Taking the quenched average, this requires a shift over $[\langle N \rangle] = (1/M) \sum_{i=1}^M \langle N \rangle_i$. Therefore

$$(N - [\langle N \rangle]) / L^d, \quad (27)$$

in a fluid with quenched disorder, is the analog of s in a magnetic system, where the factor L^d is needed because s is

⁸ Note that other choices are conceivable also, such as the ‘equal-area-rule’ [56, 59] or the generalized k -locus defined in [60]; all become identical in the limit $L \rightarrow \infty$, of course.

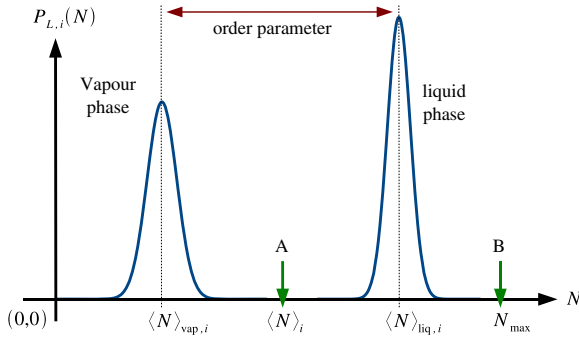


Figure 4. Schematic representation of the expected bimodal form of $P_{L,i}(N)$ obtained in a single realization of the porous medium at and below T_c . The average of the full distribution $\langle N \rangle_i$ separates the vapor from the liquid peak. The distance between the peaks, defined as the average number of particles in the liquid phase $\langle N \rangle_{liq,i}$ minus the average $\langle N \rangle_{vap,i}$ of the vapor, gives an estimate of the order parameter. The moments of the vapor and liquid peak yield the connected and disconnected susceptibilities, see details in the text.

the magnetization per spin. Replacing s in the definitions of χ and χ_{dis} by equation (27), one obtains

$$\chi = [\langle N^2 \rangle - \langle N \rangle^2] / L^d, \quad (28)$$

$$\chi_{dis} = ([\langle N \rangle^2] - [\langle N \rangle]^2) / L^d, \quad (29)$$

as the analogs of the connected and disconnected susceptibility in a fluid with quenched disorder. As stated before, χ and χ_{dis} are analyzed for the vapor and liquid phase separately, using $\langle N \rangle_i$ as a ‘cutoff’ separating the peaks in $P_{L,i}(N)$. In this way, we obtain for the vapor phase

$$\langle N^k \rangle_{vap,i} = 2 \sum_{N=0}^{\langle N \rangle_i} N^k P_{L,i}(N), \quad (30)$$

where the factor of 2 is a consequence of the normalization of $P_{L,i}(N)$. The moments $\langle N^k \rangle_{liq,i}$ of the liquid are obtained similarly, with the summation from $N = \langle N \rangle_i$ to N_{max} . The connected and disconnected susceptibilities of the vapor phase can now be written as

$$\begin{aligned} \chi_{con}^{vap} &= \frac{[\langle N^2 \rangle_{vap}] - [\langle N \rangle_{vap}^2]}{L^d}, \\ \chi_{dis}^{vap} &= \frac{[\langle N \rangle_{vap}^2] - [\langle N \rangle_{vap}]^2}{L^d}, \end{aligned} \quad (31)$$

with the quenched average $[\cdot]$ conveniently expressed in terms of equation (30) as $[\langle N^k \rangle_{vap}^l] = (1/M) \sum_{i=1}^M \langle N^k \rangle_{vap,i}^l$. Similar expressions hold for χ_{con}^{liq} and χ_{dis}^{liq} also. Note that, since the chemical potential μ_i is ‘fine-tuned’ for each realization of the porous medium, the quenched average obtained above actually corresponds to \bar{X} of equation (25), but this should be obvious from our discussion. For completeness, we remark that the quenched-averaged distance between the peaks in figure 4 may be used as the order parameter $m = [\langle N \rangle_{liq}] - [\langle N \rangle_{vap}]$, although in this work the emphasis is on the susceptibilities.

Of course, it needs to be verified in simulations if the expected bimodal form of $P_{L,i}(N)$ really occurs in practice. In

our previous work, this turned out to be the case [37]. However, GC simulations of the Lennard-Jones fluid with quenched disorder have revealed distributions with three peaks also [9]; the possibility of two fluid phase transitions occurring has also been suggested [10], although this probably does not survive in the quenched average [11].

3. Model and simulation method

3.1. AO model

We now proceed to test the concepts of the previous section in a colloid–polymer mixture with quenched disorder. Our primary aim is to measure the connected and disconnected susceptibilities, and to show that *both* diverge at criticality. To describe the mixture, we use the AO model [20, 21]. In this model, colloids (species c) and polymers (species p) are treated as spheres with respective diameters σ_c and σ_p . Hard-sphere interactions are assumed between colloid–colloid and colloid–polymer pairs, while the polymer–polymer interaction is taken to be ideal. In this work, σ_c is the unit of length, the colloid-to-polymer size ratio $q \equiv \sigma_p / \sigma_c = 1$, and the spatial dimension will be $d = 3$. The behavior of this model for $q = 1$ *without* quenched disorder has been studied before [36], and bulk phase separation, whereby the mixture ‘splits’ into a colloid-rich (polymer-poor) and colloid-poor (polymer-rich) domain, was readily observed. If one ‘identifies’ the colloid-rich phase with a liquid, and the colloid-poor phase with a vapor, the phase separation can be treated in much the same way as a liquid–vapor transition. In the GC ensemble, one then introduces the colloid chemical potential μ and, following convention, the polymer ‘chemical potential’ η_p^r . Phase separation occurs at the coexistence colloid chemical potential $\mu = \mu_{coex}$, for values of η_p^r exceeding the critical value $\eta_{p,cr}^r$ (η_p^r is therefore the analog of inverse temperature; for $q = 1$, $\eta_{p,cr}^r \approx 0.861$ has been reported [36]). The discussion and definitions of section 2.5 thus trivially ‘carry over’ to the AO model if one identifies $N \leftrightarrow$ number of colloids, $\mu \leftrightarrow$ colloid chemical potential and $T \leftrightarrow 1/\eta_p^r$.

3.2. AO model with quenched disorder

To study the AO model with quenched disorder, we introduce a third species Q of immobile (quenched) particles. These particles are also spheres, with diameter $\sigma_Q = \sigma_c$, and they are distributed in the simulation box at the start of each simulation (the simulation box, incidentally, is a cube of volume $V = L^d$ with periodic boundary conditions). The quenched particles, N_Q of them in total, are located at random positions, irrespective of overlap. Consequently, the structure of the quenched system is just that of an ideal gas. The average packing fraction of the quenched system is fixed at $\eta_Q = \pi \sigma_Q^3 N_Q / (6V) = 0.05$ but, consistent with our GC approach, we allow for Poissonian fluctuations around the average. From a computational point of view, the quenched

⁹ Strictly speaking, η_p^r is defined as the polymer reservoir packing fraction [22]. For the present case of *ideal* polymers $\eta_p^r = \pi \sigma_p^3 e^{(\mu_p/k_B T)} / 6\Lambda^3$, with μ_p the polymer chemical potential, and Λ the thermal wavelength.

system is trivial to generate: one simply draws N_Q from a Poisson distribution and generates a corresponding number of positions in the simulation box. Next, a GC simulation of the AO model is performed in the simulation box containing the quenched system, whereby the colloid and polymer positions are continuously updated, but not the positions of the quenched particles, of course. The colloids and polymers interact with the quenched particles in a simple way: colloids may *not* overlap with quenched particles, while the polymers may overlap *freely* with them. Of course, computational efficiency is the main motivation for using such simple interactions, although one could envision similar interactions in experiments also, using polymer quenched disorder. In any case, the simple approach adopted here is appealing, as previous work indicates [13, 37, 38]. An estimate $\eta_{p,cr}^r \approx 1.192$ has also already been reported [37] for the exact same parameters as considered here.

3.3. Implementation details

We now discuss some implementation details. For the i th realization of the quenched system, grand canonical MC is used to measure $P_{L,i}(N)$ of equation (26), with N the number of colloids. The distribution is obtained using the (already mentioned) biased sampling scheme [57], in conjunction with a cluster move [25, 61]. The cluster move is needed to alleviate the otherwise (too) slow equilibration of the AO model. Of course, simulations of a single-component fluid do not require the cluster move. We consider system sizes $L = 7-12$. For each system size, $P_{L,i}(N)$ is typically measured for $M \sim 2000$ (!) realizations of quenched disorder, at several values of η_p^r in the vicinity of $\eta_{p,cr}^r$. Large values of M are needed to obtain χ_{dis} accurately.

In GC simulations, particles are continuously inserted and deleted from the simulation box, and so one can define a time τ after which a given population of particles has been completely ‘updated’ by new ones. The duration of a GC simulation may therefore be expressed in units of τ . In the biased sampling scheme [57], simulation time can be conveniently allocated, since the scheme constructs $P_{L,i}(N)$ step-by-step via so-called windows. In the first window, N varies between 0 and 1, in the next window between 1 and 2, and so forth, up to N_{max} (the number of polymers N_p fluctuates freely in each window, of course). Hence, we allocate a fixed amount of simulation time, typically 5τ , to each window. It then takes roughly 12 min to obtain $P_{L,i}(N)$ for $L = 7$ and about 1 h for $L = 12$. Of course, these benchmarks depend on η_p^r , as well as on the precise computer architecture, but they suffice to give an overall impression of how much computer time was used.

A final remark concerns the implementation of histogram extrapolation [58]. As stated earlier, all simulations are performed at colloid chemical potential $\mu_i = 0$, and $P(N|\mu_i = \mu') \propto P(N|\mu_i = 0) \exp(\mu'N)$ is used to extrapolate to different values. Obviously, a similar expression holds for the polymers also, which one could use to extrapolate in η_p^r . In fact, an important ingredient of this work is precisely the latter extrapolation, and our analysis would become extremely cumbersome without it. However, this requires that

we store the full two-dimensional histogram $P_{L,i}(N, N_p)$, with N the number of colloids and N_p the number of polymers. Since we typically consider 2000 realizations of quenched disorder, storage requirements become enormous. Fortunately, storage can be reduced tremendously when one realizes that, for a fixed number of colloids N , the corresponding distribution in N_p is to a good approximation a single Gaussian peak. For $N = 0$ this is obvious, since then we have a pure polymer system, but it holds well for $N > 0$ also. Hence, to facilitate extrapolations in η_p^r , we only need to store the average and variance in N_p for each window (which costs only very little storage, at no cost in CPU time either). We have verified this approach and checked that results obtained at one value of η_p^r indeed extrapolate to those obtained at a different value (not too far away, of course). Note also that the histogram extrapolation method itself can be optimized since, for a Gaussian distribution, integrations over N_p can be performed exactly beforehand; the resulting expressions become functions of the average and variance, which can be hard-coded.

4. Results

4.1. Sample-to-sample fluctuations

The analysis of section 2.5 requires that the distributions $P_{L,i}(N)$ are somewhat bimodal, i.e. that they resemble the schematic shape of figure 4. In order to verify this, we show, in figure 5, $P_{L,i}(N)$ for a number of realizations of quenched disorder at a value of η_p^r significantly *below* the critical value $\eta_{p,cr}^r \approx 1.192$. Clearly, the bimodal shape is already present in most distributions, even for this low value of η_p^r . Of course, by making η_p^r even lower, the bimodal shape will eventually vanish for all realizations of quenched disorder, since we then enter the one-phase region where $P_{L,i}(N)$ is just a single peak, conforming to figure 1(b). In any case, figure 5 does confirm our expectation that, for random-field Ising universality, bimodal distributions persist well above T_c (recall that η_p^r is the analog of inverse temperature). Figure 5 also reveals that not all the distributions are bimodal, see, for example, the distribution in the upper left corner. In these cases, splitting the distribution in half at the average is not meaningful anymore, although numerically this can still be applied. Since, for the thousands of distributions generated in our simulations, inspecting each one visually by hand is not feasible, the (occasional) single-peaked distribution is treated in the same way as the bimodal ones. Of course, single-peaked distributions become increasingly rare upon increasing η_p^r , as figure 6 clearly indicates, where the same realizations of quenched disorder were used as in figure 5. Note that, in figure 6, all distributions now feature two peaks.

Another feature that emerges from these figures is that the vapor peak is much sharper than the liquid peak. This appears to be a non-universal feature that depends on the interaction between fluid and quenched particles. In our previous work, we have studied a different type of quenched disorder, whereby also the polymers were not allowed to overlap with the quenched species [37, 38]. In this case, a reversed trend was

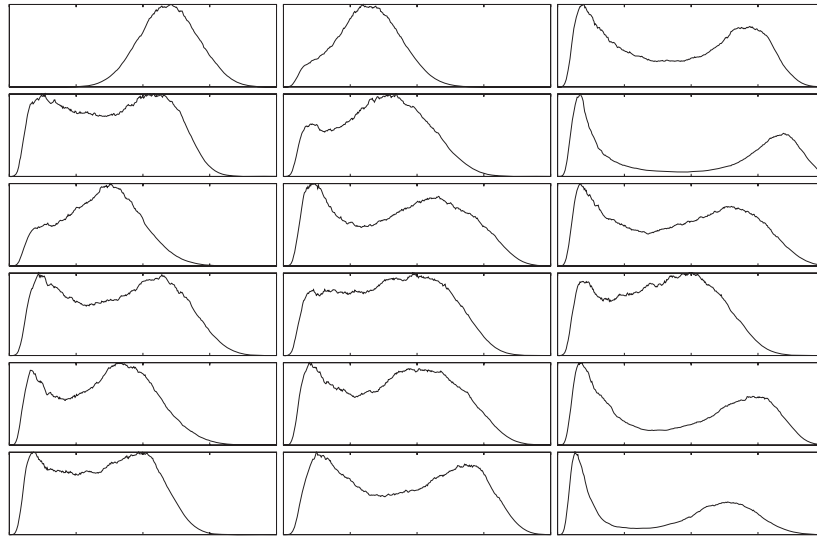


Figure 5. Distributions $P_{L,i}(N)$ for 18 different realizations of quenched disorder using $\eta_p^r = 1.05$ and $L = 10$. The horizontal axes in each of the plots show the colloid packing fraction $\eta_c \equiv \pi\sigma_c^3 N/(6V)$ from $\eta_c = 0 \rightarrow 0.2$ (left to right); the unit on the vertical axes is arbitrary.

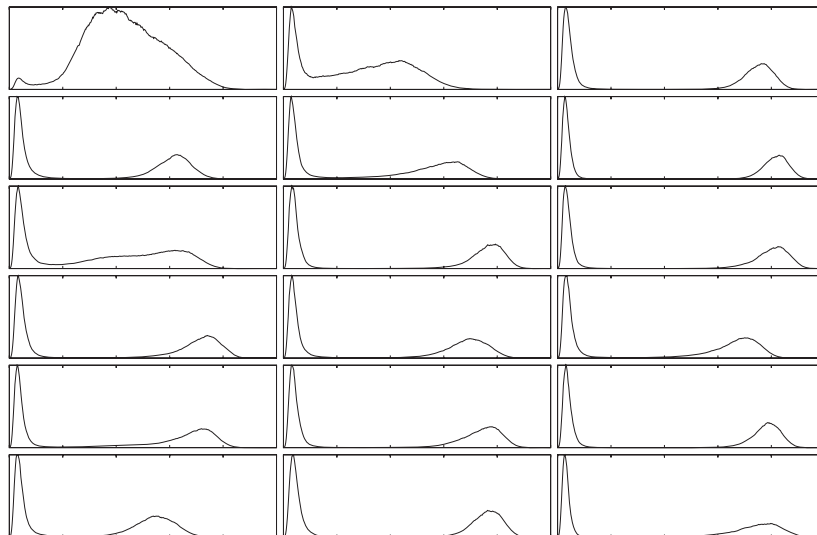


Figure 6. Same as figure 5 but for $\eta_p^r = 1.15$ and $L = 10$; the colloid packing fraction on the horizontal axes is from $\eta_c = 0 \rightarrow 0.25$.

observed, namely a sharp liquid peak ‘coexisting’ with a much broader vapor.

Having shown that our assumption of bimodal distribution shape is a reasonable one, we need to determine the number of quenched disorder realizations M typically required to obtain $\chi_{\text{con}}^{\text{liq}}$ and $\chi_{\text{dis}}^{\text{liq}}$ accurately. To this end we show, in figure 7, the variation of these quantities as a function of M , for two system sizes L , and several values of η_p^r (as indicated in the label of each subplot). The trends revealed in figure 7 are typical for other state points also. Clearly, from this figure, we conclude that M should be of the order of 1000 at least. Larger values are better still, but then we meet the limit of our computational resources.

4.2. Connected susceptibility

We now consider the connected susceptibility, first of the liquid phase. Shown in figure 8(a) is $\chi_{\text{con}}^{\text{liq}}$ versus η_p^r , for several

system sizes. Note the presence of the peak. Consistent with finite size scaling, the peak height increases with L ; the latter could now be fitted to equation (7b) to obtain γ/ν . However, a more stringent test is to plot

$$tL^{1/\nu} \text{ versus } \chi_{L,\text{con}}^{\text{liq}} L^{-\gamma/\nu}, \quad (32)$$

with $t = \eta_p^r/\eta_{p,\text{cr}}^r - 1$ the relative distance from the critical point. Although not derived in this work, finite size scaling implies that data from different system sizes, when scaled conforming to equation (32), collapse onto a single master curve, provided the correct values of $\eta_{p,\text{cr}}^r$, ν and γ are used¹⁰. In figure 8(b) the resulting scaling plot is shown, where $\eta_{p,\text{cr}}^r = 1.194$, $\nu = 1.1$ and $\gamma/\nu = 1.87$ were used. The quality of the collapse is clearly very good. However, we noticed that good collapses were obtained for different

¹⁰ The derivation is straightforward, see, for example, [41].

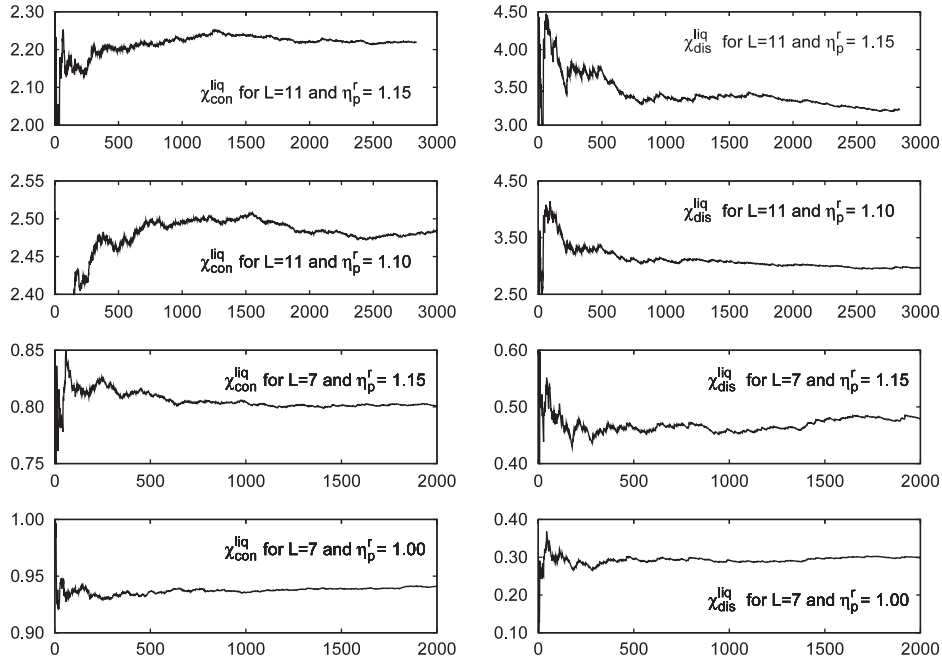


Figure 7. ‘Moving average’ of the susceptibilities $\chi_{\text{con}}^{\text{liq}}$ (left frames) and $\chi_{\text{dis}}^{\text{liq}}$ (right frames) of the liquid phase, for $L = 7$ and 11 , using several values of η_p^r as indicated in the labels. Plotted are the susceptibilities (vertical axes) versus the number of quenched disorder realizations M (horizontal axes).

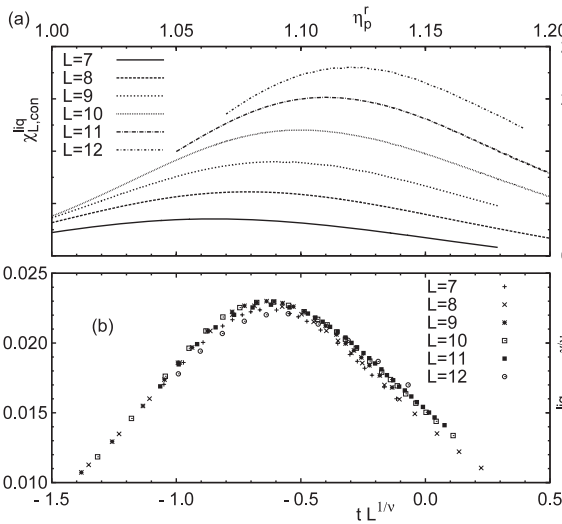


Figure 8. (a) Connected susceptibility of the liquid phase versus η_p^r for several system sizes L . Note the increase of peak height with L , and also the shift in the peak positions. (b) Finite size scaling plot, where $\eta_{p,\text{cr}}^r = 1.194$, $\nu = 1.1$ and $\gamma/\nu = 1.87$ were used (see details in the text).

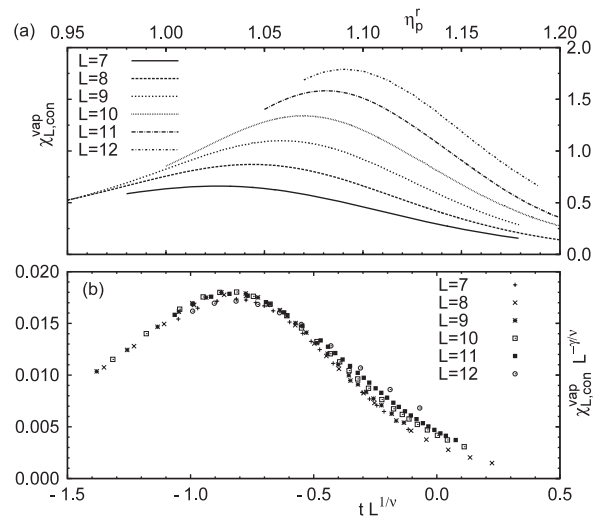


Figure 9. Same as figure 8 but for the connected susceptibility of the vapor. In the scaling plot of (b), $\eta_{p,\text{cr}}^r = 1.194$, $\nu = 1.1$ and $\gamma/\nu = 1.87$ were used.

values also, typically $\nu = 1.0\text{--}1.2$ and $\eta_{p,\text{cr}}^r = 1.19\text{--}1.22$, which gives an indication of the uncertainty. The problem is that both ν and $\eta_{p,\text{cr}}^r$ follow from the L dependence of the peak positions. Over the range of available system sizes, the shift in the peak positions is rather small and hence large uncertainties in ν and $\eta_{p,\text{cr}}^r$ are unavoidable. In contrast, γ/ν can be obtained more reliably, since the latter is set by the peak height versus L , which yields a more pronounced numerical signature. Similar conclusions are

reached for the connected susceptibility of the vapor, see figure 9.

4.3. Disconnected susceptibility

We now come to the main result of this work, namely the behavior of the disconnected susceptibility. If fluids with quenched disorder belong to the universality class of the RFIM, the analog of χ_{dis} defined in section 2.5 should diverge with critical exponent $\bar{\gamma}$. Since $\bar{\gamma} = 2\gamma$ [45], the divergence should be very pronounced, much more pronounced than that of the

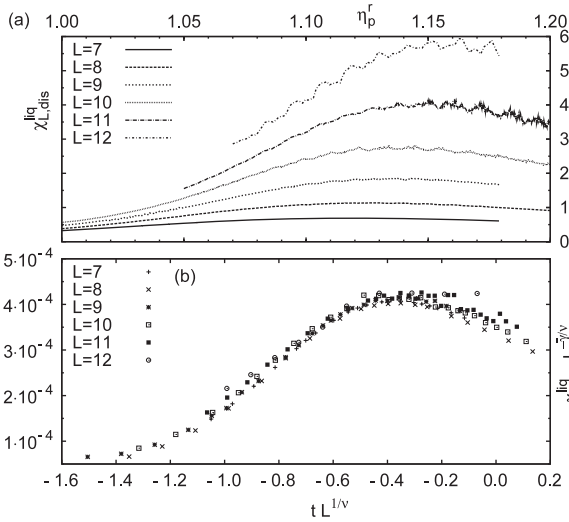


Figure 10. The main result of this paper: (a) disconnected susceptibility of the liquid phase versus η_p^r for several system sizes L and (b) the corresponding finite size scaling plot, where $\eta_{p,cr}^r = 1.194$, $\nu = 1.1$ and $\bar{\gamma}/\nu = 3.82$ were used (see details in the text).

connected susceptibility, in fact. In figure 10(a), we show χ_{dis}^{liq} of the liquid phase versus η_p^r for several system sizes. The formation of a peak is clearly visible. Note also the rapid growth of the peak height: increasing the system size from $L = 7 \rightarrow 12$, the disconnected susceptibility increases by a factor of more than six, compared to a factor of about three for the connected susceptibility. The corresponding scaling plot is shown in figure 10(b), which now involves $\bar{\gamma}$, of course. Using $\eta_{p,cr}^r = 1.194$, $\nu = 1.1$ and $\bar{\gamma}/\nu = 3.82$, the data collapse convincingly, confirming the power law divergence of χ_{dis} . For the same reason as before, the scaling plot is rather insensitive to $\eta_{p,cr}^r$ and ν , and so the uncertainty in these quantities is similar as before, but the ratio $\bar{\gamma}/\nu$ should again prove reliable. In figure 11(a) we plot the disconnected susceptibility of the vapor phase versus η_p^r , but only for $L \leq 10$. For reasons we do not yet fully understand, the statistical uncertainty in χ_{dis}^{vap} is very large. While the growth of a peak with system size is still confirmed, the data clearly do not lend themselves for measuring critical exponents, and so a scaling plot is not attempted here. One reason for the large statistical uncertainty in χ_{dis}^{vap} is the smaller number of particles in the vapor phase (compared to the liquid).

4.4. Scaling of the cumulant

In figure 12(a), we show the cumulant as a function of η_p^r for several system sizes. Recall that the cumulant is defined as $U_1 \equiv [(s^2)]/[(|s|)^2]$, with $s = (N - \langle N \rangle)/L^d$, which can be calculated straightforwardly from the distributions $P_{L,i}(N)$ ¹¹. As expected, the cumulants from different system sizes do not intersect at criticality, but instead reveal a scatter of intersection points, close to $U_1^* = 1$ of the RFIM. This behavior conforms

¹¹ In our previous work [37], we used $U_1 \equiv [(s^2)]/[(|s|)^2]$, but the reader can verify following section 2.2 that both definitions become equivalent for $L \rightarrow \infty$.

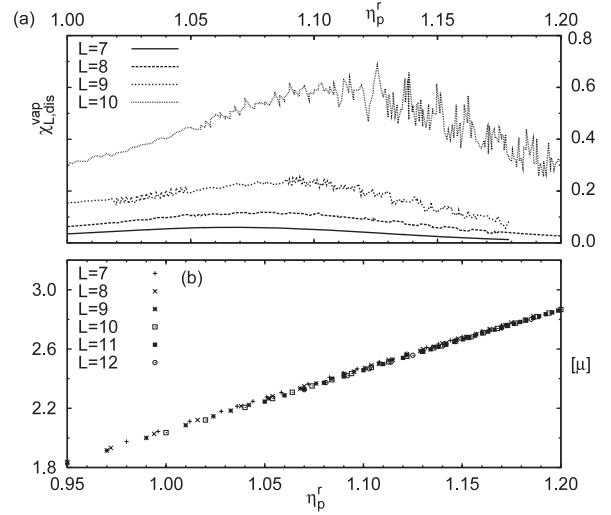


Figure 11. (a) Disconnected susceptibility of the vapor phase versus η_p^r and (b) the quenched-averaged chemical potential versus η_p^r (for several system sizes L).

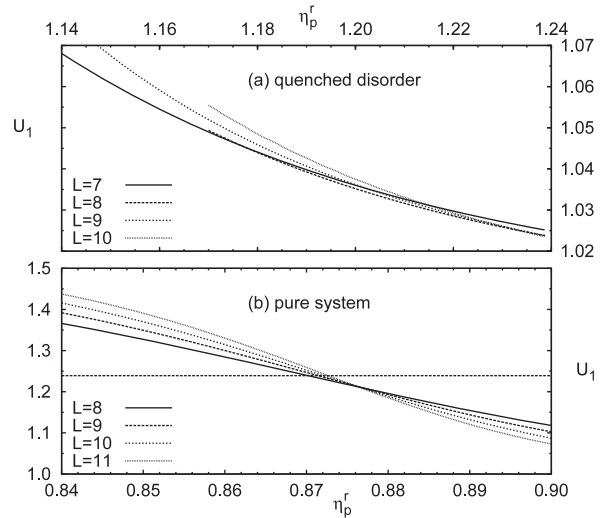


Figure 12. Plots of U_1 versus η_p^r , using several system sizes L , for the AO model with quenched disorder (a) and without (b). The horizontal line in (b) marks $U_1^* \approx 1.239$ of the Ising model in three dimensions.

to our discussion of section 2.2 and confirms that $P_{L,i}(N)$ remains *sharp* at the critical point, featuring two well-separated peaks, since hyperscaling is now violated.

For comparison, figure 12(b) shows the cumulant of the AO model in the pure system, i.e. without quenched disorder. In this case, hyperscaling is not violated and a sharp intersection point is indeed revealed, occurring at a value U_1^* *different* from the off-critical values 1 and $\pi/2$, respectively. For Ising systems in $d = 3$ dimensions, we expect that $U_1^* \approx 1.239$ [62], marked by the horizontal line in figure 12(b), and our data indeed intersect close to this value (some deviation is clearly apparent, but to account for this would require a field-mixing analysis [26, 63]). From the intersection point, we also conclude $\eta_{p,cr}^r \approx 0.876$ for

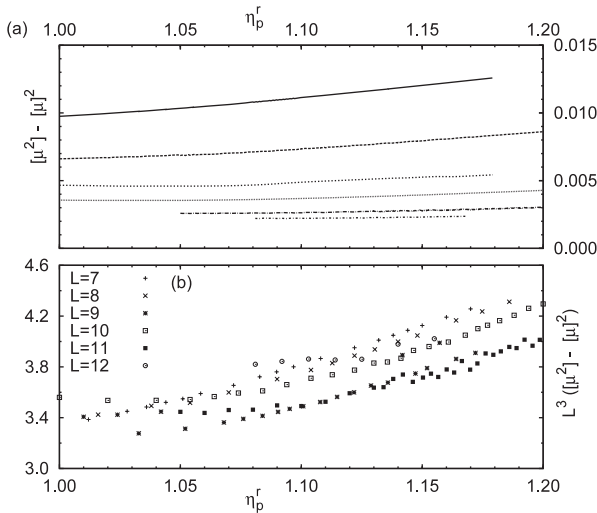


Figure 13. (a) Variance $[\mu^2] - [\mu]^2$ versus η_p^r for system sizes $L = 7 \rightarrow 12$ (from top to bottom). (b) Same as above, but with the variance scaled by L^d .

the pure system, which compares well to the estimate reported in [36].

4.5. Chemical potential

Finally, we consider the average and variance of the chemical potentials at which our data were obtained. Recall that, for each realization i of quenched disorder, we use a ‘fine-tuned’ chemical potential μ_i , chosen at the maximum of $\partial \langle N \rangle_i / \partial \mu_i$ for that realization. Hence, it is interesting to consider the quenched-averaged chemical potential $[\mu]$ and its variance $[\mu^2] - [\mu]^2$, with $[\mu^k] = (1/M) \sum_{i=1}^M \mu_i^k$. Shown in figure 11(b) is $[\mu]$ versus η_p^r for several system sizes. The data do not reveal any strong L dependence, which is similar to that observed in fluids without quenched disorder. Of more interest is the variance, which should vanish for $L \rightarrow \infty$. Shown in figure 13(a) is $[\mu^2] - [\mu]^2$ versus η_p^r and the decrease of this quantity with increasing L is clearly visible. Kierlik *et al* have shown that, below T_c in the two-phase region, the variance of the chemical potential vanishes $\propto L^{-d}$ [14]. Plotting therefore $L^d([\mu^2] - [\mu]^2)$ versus η_p^r , see figure 13(b), we observe that this prediction holds quite well for our data also.

5. Discussion and summary

We have explained finite size scaling in the random-field Ising model and shown how this technique may be applied to a fluid with quenched disorder. We have also defined the analog of the disconnected susceptibility χ_{dis} for the latter. If fluids with quenched disorder belong to the universality class of the random-field Ising model, as conjectured by de Gennes [7], χ_{dis} should diverge at criticality, and so our definition facilitates further tests of this conjecture. To perform one such test has been the topic of the present work, using the Asakura–Oosawa model of a colloid–polymer mixture confined to a random porous medium. Our data are indeed compatible with a divergence of χ_{dis} . Moreover, for the liquid

phase, we even recover $\bar{\gamma} \approx 2\gamma$, in quantitative agreement with the prediction of Schwartz for the random-field Ising model [45]. Our estimate of the correlation length exponent $\nu \approx 1.0 - 1.2$, although not very precise, is also consistent with reported random-field Ising estimates [50, 51]. Hence, the present results confirm our earlier study [37], where evidence of random-field Ising universality in fluids with quenched disorder was also presented, but based on the shape of $P_{L,i}(N)$ at criticality. We also remind the reader of the large number of porous medium realizations used in our analysis. As the ‘moving averages’ of figure 7 indicate, such numbers become a necessity, if χ and χ_{dis} are to be obtained with any meaningful accuracy.

Finally, we turn to a discussion of possible applications of our work to experiments. The prototype experimental realization of a fluid with quenched disorder is an *atomic* fluid injected into silica aerogel. This realization has the disadvantage that the coupling between the porous medium and the fluid is weak [3, 4], as manifested by the small shift of the critical temperature (compared to the system without quenched disorder). Moreover, the characteristic length over which the aerogel structure appears random is very large, compared to the size of the fluid molecules. In this respect, colloidal fluids may offer an attractive alternative. Note that investigations of critical phenomena in colloid–polymer mixtures *without* quenched disorder [64, 65] are already experimentally feasible: critical interface and density fluctuations can be visualized directly [66, 67] using confocal microscopy [68]. In principle, such confocal experiments could be extended to include quenched disorder also. The generation and synthesizing of quenched colloidal porous media has received considerable attention [69–71]. One could envision an experiment whereby a colloid–polymer mixture is injected into a rigid colloidal gel. Such gels could be formed using small nanoparticles which can grow into randomly branched networks at volume fractions of only a few percent [71]. The size of these nanoparticles can be much smaller than the typical colloid or polymer diameter, and so one can easily reach the regime where the critical correlations of the colloid–polymer mixture average over the random structure of the gel. Another feasible realization would be to use a polymer blend containing nanoparticles of suitable size, such that the diffusion of these particles in the blend is small. The structure formed by the nanoparticles will then appear to be frozen (quenched) on the timescales needed for the critical correlations of the polymer blend to equilibrate. The latter could then be measured using, for example, light scattering¹². In any case, we hope that the simulational efforts of the present work will stimulate experimental efforts also, in order to completely settle this long-standing problem.

Acknowledgments

This work was supported by the Deutsche Forschungsgemeinschaft under the SFB-TR6 (project sections A5 and D3) and the Emmy Noether program (VI 483/1-1).

¹² We are indebted to G Fytas for a stimulating discussion of this point.

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