Propagating Beliefs in Spin-Glass Models

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We investigate dynamics of an inference algorithm termed the belief propagation (BP) when employed in spin glass (SG) models and show that its macroscopic behaviors can be traced by recursive updates of certain auxiliary field distributions the stationary state of which reproduces the replica symmetric solution offered by the equilibrium analysis. We further provide a compact expression for instability condition of the BP's fixed point which turns out to be identical to that of instability for breaking the replica symmetry in equilibrium when the number of couplings per spin is infinite. This correspondence is extended to the case of finite connectivity to determine the phase diagram, which is numerically validated.

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1. Introduction

Recently, there is growing interest in a similarity between researches on spin glass (SG) and information processing (IP).¹⁾ Since employment of methods from SG theory provided significant progresses for several problems related to IP such as machine learning,²⁾ error-correcting³⁻⁶⁾ and spreading codes,^{7,8)} it is natural to expect that the opposite direction might be possible.

The purpose of this paper is to show such an example. More specifically, we will show that investigating dynamics of an iterative inference algorithm termed the belief propagation (BP), which has been developed in IP research,^{9,10)} provides a new understanding about thermodynamical properties of SG when it is employed in SG models.

This paper is organized as follows. In the next section, we introduce BP to a family of SG models. This model family covers a variety of SG models actively studied in the past, which is convenient for connecting results obtained here to the existing knowledge. In §3, we investigate macroscopic behavior of BP in the SG models. We show that the replica symmetric (RS) solution known in the equilibrium analysis can be characterized as a macroscopically stationary state in BP. However, this does not imply that BP microscopically converges to a certain state. In §4, we provide a compact expression of the microscopic instability condition around the fixed point in the BP dynamics, which turns out to be identical to that of instability for breaking the replica symmetry in equilibrium termed the de Almeida-Thouless (AT) instability¹¹⁾ when the number of connectivity per spin is infinite. Efficacy of this expression for a sparsely connected SG model is also numerically confirmed. The final section is devoted to summary.

2. Belief Propagation in Spin-Glass Models

We here take up a family of Ising SG models define by Hamiltonian

$$\mathcal{H}(\boldsymbol{S} \mid \boldsymbol{J}) = -\sum_{\mu=1}^{M} J_{\mu} \prod_{l \in \mathcal{L}(\mu)} S_{l}, \qquad (1)$$

where $\mathcal{L}(\mu)$ denotes a set of indices which are connected to a quenched coupling J_{μ} . We assume that each coupling is independently generated from an identical distribution

$$P(J_{\mu}) = \frac{1 + J_0/(\sqrt{CJ})}{2} \delta\left(J_{\mu} - \frac{J}{\sqrt{C}}\right) + \frac{1 - J_0/(\sqrt{CJ})}{2} \delta\left(J_{\mu} + \frac{J}{\sqrt{C}}\right).$$
(2)

We further assume that for each μ , $\mathcal{L}(\mu)$ is composed of randomly selected $K \sim O(1)$ spin indices and each spin index *l* is concerned with *C* couplings the set of which is denoted as $\mathcal{M}(l)$. $J_0 > 0$ and J > 0 are parameters to control the mean and the standard deviation of J_{μ} , respectively, which naturally links the current system (1) to the Sherrington–Kirkpatrick (SK) model¹²⁾ in the case of K =2 and $C \sim O(N)$, and to sparsely connected SG models^{13–15)} in general.

A major goal of statistical mechanics in the current system is to calculate the microscopic spin average $m_l =$ $\operatorname{Tr}_S S_l \exp[-\beta \mathcal{H}(S|J)]/\operatorname{Tr}_S \exp[-\beta \mathcal{H}(S|J)]$ from given Hamiltonian (1). This is formally identical to an inference problem for a posterior distribution $P(S|J) \propto \prod_{\mu=1}^{M} P(J_{\mu}|S)$ derived from a conditional probability $P(J_{\mu}|S) =$ $\exp[\beta J_{\mu} \prod_{l \in \mathcal{L}(\mu)} S_l]/\sum_{J_{\mu} = \pm J/\sqrt{C}} \exp[\beta J_{\mu} \prod_{l \in \mathcal{L}(\mu)} S_l]$ and a uniform prior, which can be expressed in a bipartite graph as Fig. 1(a). In this expression, spins and couplings are denoted as two different types of nodes and are linked by



Fig. 1. (a) Graphical expression of SG models in the case of K = 3 and C = 4. In this expression, each spin S_l denoted as \bigcirc is linked to C = 4 couplings J_{μ} (\Box), each of which is connected to K = 3 spins. $\mathcal{L}(\mu)$ and $\mathcal{M}(l)$ represent sets of indices of spins and couplings that are related to J_{μ} and S_l , respectively. In the figure, $\mathcal{L}(\mu) = \{l_1, l_2, l_3\}$ and $\mathcal{M}(l) = \{\mu_1, \mu_2, \mu_3, \mu_4\}$. (b) Cycles in a graph. A cycle is composed of multiple paths to link an identical pair of nodes. It is shown that BP can provide the exact spin averages in a practical time scale if a given graph is free from cycles.⁹

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edges when they are directly connected, which is useful to explicitly represent statistical dependences between estimation variables (spins) and observed data (couplings).

BP is an iterative algorithm defined over the bipartite graph to calculate the spin average for a given set of couplings $J = (J_{\mu})^{.9,10)}$ In the current system, this is performed by passing *beliefs* (or messages) between the two types of nodes via edges at each update as

$$\hat{m}_{\mu l}^{t+1} = \tanh\left(\beta J_{\mu}\right) \prod_{k \in \mathcal{L}(\mu) \setminus l} m_{\mu k}^{t}, \tag{3}$$

$$m_{\mu l}^{t} = \tanh\left(\sum_{\nu \in \mathcal{M}(l) \setminus \mu} \tanh^{-1} \hat{m}_{\nu l}^{t}\right), \qquad (4)$$

where beliefs $m_{\mu l}^{t}$ and $\hat{m}_{\mu l}^{t}$ are parameters to represent auxiliary distributions at *t*th update as $P(S_{l}|\{J_{\nu \neq \mu}\}) = (1 + m_{\mu l}^{t}S_{l})/2$ and $P(J_{\mu}|S_{l}, \{J_{\nu \neq \mu}\}) = \operatorname{Tr}_{S_{k\neq l}} P(J_{\mu}|S)P(S|\{J_{\nu \neq \mu}\}) \propto (1 + \hat{m}_{\mu l}^{t}S_{l})/2$, respectively. $\mathcal{L}(\mu) \setminus l$ stands for a set of spin indices which belong to $\mathcal{L}(\mu)$ other than *l* and similarly for $\mathcal{M}(l) \setminus \mu$. Calculating $\hat{m}_{\mu l}$ iteratively, the estimate of the spin average at *t*th update is provided as

$$m_l^t = \tanh\left(\sum_{\mu \in \mathcal{M}(l)} \tanh^{-1} \hat{m}_{\mu l}^t\right).$$
 (5)

It is known that BP provides the exact spin averages by the convergent solution when the bipartite graph is free from cycles [Fig. 1(b)]. Actually, BP is a very similar scheme to the transfer matrix method (TMM) and the Bethe approximation^{16,17)} which are frequently used in physics, and the current statement can be regarded as a generalization of a known property of TMM that offers the exact results for a one dimensional lattice or a tree. However, BP still has a possibility to introduce something new into physics since it is explicitly expressed as an *algorithm* and such view point has been rare in the research on matters. This strongly motivates us to examine its dynamical properties, which we will focus on hereafter.

3. Macroscopic Behavior and the Replica Symmetric Solution

Let us first discuss the macroscopic behavior of the BP dynamics (3) and (4). Although the current randomly constructed system is not free from cycles, it can be shown that the typical length of the cycles grows as $O(\ln N)$ with respect to the system size N as long as C is O(1),¹⁸⁾ which implies that the self-interaction from the past state is presumably negligible in the thermodynamic limit. On the other hand, the self-interaction is also expected as sufficiently small even if C is large since the strength of the coupling becomes weak as $O(C^{-1/2})$. This and eqs. (3) and (4) imply that the time evolution of the macroscopic distributions of beliefs $\pi^t(x) \equiv (1/NC) \sum_{l=1}^N \sum_{\mu \in \mathcal{M}(l)} \delta(x - m_{\mu l}^t)$ is likely to be well captured by recursive equations

$$\hat{\pi}^{t+1}(\hat{x}) = \int \prod_{l=1}^{K-1} \mathrm{d}x_l \pi^t(x_l) \left\langle \delta \left(\hat{x} - \tanh(\beta \mathcal{J}) \prod_{l=1}^{K-1} x_l \right) \right\rangle_{\mathcal{J}}, \quad (6)$$

$$\pi^{t}(x) = \int \prod_{\mu=1}^{C-1} d\hat{x}_{\mu} \hat{\pi}^{t}(\hat{x}_{\mu}) \delta\left(x - \tanh\left(\sum_{\mu=1}^{C-1} \tanh^{-1} \hat{x}_{\mu}\right)\right), (7)$$

where $\langle \cdots \rangle_{\mathcal{J}}$ represents the average with respect to \mathcal{J} following distribution (2).

The validity of the current argument and its link to the replica symmetric (RS) ansatz in the equilibrium analysis have been shown already for finite C.^{19,20)} Here, we further show that these can be extended to the case of infinite C even if the AT stability of the RS solution is broken in equilibrium.

When *C* becomes infinite, it is more convenient to deal with an auxiliary field of finite strength $h_{\mu l}^{t} \equiv \sum_{v \in \mathcal{M}(l) \setminus \mu} \tanh^{-1} \hat{m}_{vl}^{t} \simeq \sum_{v \in \mathcal{M}(l) \setminus \mu} \hat{m}_{vl}^{t}$ rather than \hat{m}_{vl}^{t} since $\hat{m}_{\mu l}^{t}$ becomes infinitesimal. Due to the central limit theorem, the distribution of the auxiliary field $\rho^{t}(h) \equiv (1/NC)$ $\sum_{l=1}^{N} \sum_{\mu \in \mathcal{M}(l)} \delta(h - h_{\mu l}^{t})$ can be regarded as a Gaussian

$$\rho^{t}(h) = \int \prod_{\mu=1}^{C-1} d\hat{x}_{\mu} \hat{\pi}^{t}(\hat{x}_{\mu}) \delta \left(h - \sum_{\mu=1}^{C-1} \tanh^{-1} \hat{x}_{\mu} \right)$$

$$\simeq \frac{1}{\sqrt{2\pi F^{t}}} \exp\left[-\frac{(h-E^{t})^{2}}{2F^{t}} \right],$$
(8)

where E^t and F^t are the average and the variance to parameterize the Gaussian distribution $\rho^t(h)$, respectively. The expression in the middle implies $\pi^t(x) = \int dh \rho^t(h) \delta$ $(x - \tanh(h))$. Plugging this into eq. (6) and recursively employing eq. (8), we obtain a compact expression for the update of E^t and F^t as

$$E^{t+1} = \beta J_0 (M^t)^{K-1},$$

$$F^{t+1} = \beta^2 J^2 (Q^t)^{K-1},$$
(9)

$$M^{t} = \int Dz \tanh(\sqrt{F^{t}}z + E^{t}),$$

$$Q^{t} = \int Dz \tanh^{2}(\sqrt{F^{t}}z + E^{t}),$$
(10)

where $Dz \equiv \exp[-z^2/2]/\sqrt{2\pi}$ and M^t and Q^t can be expressed as $M^t \simeq (1/N) \sum_{l=1}^N m_{\mu l}^t \simeq (1/N) \sum_{l=1}^N m_l^t$ and $Q^t \simeq (1/N) \sum_{l=1}^N (m_{\mu l}^t)^2 \simeq (1/N) \sum_{l=1}^N (m_l^t)^2$, respectively, due to the law of large numbers. Equations (9) and (10) serve as alternatives of eqs. (6) and (7).

It should be noticed here that these equations can be regarded as the forward iteration of the saddle point equations to obtain the RS solution in the replica analysis of the multi-spin interaction infinite connectivity SG models¹⁾ and, in particular, of the SK model for K = 2.¹²⁾ In order to confirm the validity of the above argument, we compared the time evolution of the belief update (3) and (4) (**BP**) with that of eqs. (9) and (10) (**RS**) for the SK (K = 2) model, which is shown in Figs. 2(a) and 2(b). We also compared them with the trajectory of the naive iteration of the BP's fixed point condition

$$m_{l} = \tanh\left(\sum_{\mu \in \mathcal{M}(l)} \beta J_{\mu} \prod_{k \in \mathcal{L}(\mu) \setminus l} m_{k} - \sum_{\mu \in \mathcal{M}(l)} (\beta J_{\mu})^{2} \times \sum_{j \in \mathcal{L}(\mu) \setminus l} \left(\prod_{k \in \mathcal{L}(\mu) \setminus l, j} m_{k}\right)^{2} (1 - m_{j}^{2}) m_{l}\right),$$
(11)

(**TAP**) which can be obtained inserting $m_{\mu l} \simeq m_l - (1 - m_l^2)\hat{m}_{\mu l}$ to the fixed point of eqs. (3) and (4) $m_{\mu l}^t = m_{\mu l}$, $\hat{m}_{\mu l}^t = \hat{m}_{\mu l}$ and $m_l^t = m_l$. This becomes identical to the



Fig. 2. Time evolution of macroscopic variables $M^t = (1/N) \sum_{l=1}^{N} m_l^t$ and $Q^{t} = (1/N) \sum_{l=1}^{N} (m_{l}^{t})^{2}$ in the SK model for the BP updates (3) and (4) (BP: \bigcirc), the reduced dynamics (9) and (10) (RS: lines) and the naive iteration of the TAP equation (11) (**TAP**: +) for (a) $J_0 = 1.5$ and (b) $J_0 = 0.5$ keeping J = 1 and T = 0.5. TAP is plotted only for Q^t in the case of $J_0 = 0.5$ in order to save space. Each marker is obtained from 100 experiments for N = 1000 systems. The AT stability is satisfied for $J_0 = 1.5$ but broken for $J_0 = 0.5$. Irrespectively of the AT stability, the behavior of the macroscopic variables in the BP dynamics can be well captured by the reduced dynamics while the naive iteration of the TAP equation does not exhibit any convergence even in the macroscopic scale. Insets: Squared deviation of spin averages between the successive updates $D^{t} = (1/N) \sum_{l=1}^{N} (m_{l}^{t} - m_{l}^{t-1})^{2}$ is plotted for the BP dynamics. The deviation vanishes to zero indicating convergence to a fixed point solution for $J_0 = 1.5$ while remains finite signaling instability of the fixed point for $J_0 = 0.5$. It may be worthwhile to mention that the microscopic trajectory of the BP dynamics for $J_0 = 0.5$ exhibits not simple oscillatory but chaotic behavior although D^t converges to a constant value. In experiments, such chaotic motion was observed even when control parameters were set much closer to the onset of the AT instability. The origin of this non-trivial behavior may be due to the the (continuous) semi-circular eigenvalue distribution of the interaction matrix of the SK model for which many modes of fluctuations simultaneously become unstable at the critical point.

famous Thouless–Anderson–Palmer (TAP) equation of the SK model, in particular, for K = 2.²¹⁾

The experiments were performed for $J_0 = 1.5, 0.5$ keeping J = 1 and T = 0.5, where the AT stability of the RS solution in equilibrium is satisfied for $J_0 = 1.5$ but broken for $J_0 = 0.5$.¹¹⁾ Figures 2(a) and 2(b) show that **BP** and **RS** exhibit excellent consistency with respect to the macroscopic variables irrespectively of whether the AT stability is satisfied or not. This strongly validates the reduction from BP (3) and (4) to the macroscopic dynamics (9) and (10). On the other hand, TAP is considerably different from the others. In a sense, this may be natural because naively iterating eq. (11) is just one of procedures for obtaining a solution and its trajectory in dynamics does not necessarily have any consistency with BP or RS while the BP's fixed point is correctly characterized by the TAP equation (11) which does have a certain relation to the RS solution in equilibrium as shown in ref. 22. These figures also imply that the dynamics of BP cannot be traced by a closed set of equations with respect to singly indexed variables m_l^t even for $C \rightarrow \infty$ while the fixed point condition in this limit is provided as coupled equations of m_l (11). Such necessity of keeping extra variables for tracing the trajectory of m_l^t in the BP dynamics is also observed in a similar system of infinite connectivity.⁸⁾

4. Microscopic Stability and the AT Condition

Although Figs. 2(a) and 2(b) show that the macroscopic variables rapidly converge to those of the RS solution in BP, this does not imply that BP microscopically converges to a certain solution. In order to probe this microscopic con-

vergence, we numerically examined the squared difference of the spin averages between successive updates $D^t \equiv (1/N) \sum_{l=1}^{N} (m_l^t - m_l^{t-1})^2$, the time evolution of which is shown in the insets of Figs. 2(a) and 2(b). These illustrate that the (microscopic) local stability of the BP's fixed point can be broken even if the macroscopic behavior seems to converge, which cannot be detected by only examining the reduced macroscopic dynamics (9) and (10).

In order to characterize such instability, we next turn to the stability analysis of the BP updates (3) and (4). Linearizing the updates with respect to the auxiliary field $h_{\mu l} = \tanh^{-1} m_{\mu l}$ around a fixed point solution $m_{\mu l}^{t} = m_{\mu l}$, we obtain a dynamics of the auxiliary field fluctuation $\delta h_{\mu l}^{t}$ as

$$\delta h_{\mu l}^{t+1} = \sum_{\nu \in \mathcal{M}(l) \setminus \mu} \frac{\tanh(\beta J_{\nu}) \prod_{k \in \mathcal{L}(\nu) \setminus l} m_{\nu k}}{1 - \left(\tanh(\beta J_{\nu}) \prod_{k \in \mathcal{L}(\nu) \setminus l} m_{\nu k} \right)^2}$$
(12)
$$\times \sum_{j \in \mathcal{L}(\nu) \setminus l} \frac{1 - m_{\nu j}^2}{m_{\nu j}} \times \delta h_{\nu j}^t.$$

Analytically solving this linearized equation for a large graph is generally difficult. However, since the current system is randomly constructed, the self-interaction of $\delta h_{\mu l}^{t}$ from the past can be considered as small as those of beliefs are. This implies that the time evolution of the fluctuation distribution $f^{t}(y) \equiv (1/NC) \sum_{l=1}^{N} \sum_{\mu \in \mathcal{M}(l)} \delta(y - \delta h_{\mu l}^{t})$ can be provided by a functional equation

$$f^{t+1}(y) = \int \prod_{\mu=1}^{C-1} \prod_{l=1}^{K-1} dy_{\mu l} f^{t}(y_{\mu l}) \\ \times \left\langle \delta \left(y - \sum_{\mu=1}^{C-1} \frac{\tanh(\beta \mathcal{J}_{\mu}) \prod_{k=1}^{K-1} x_{\mu k}}{1 - \left(\tanh(\beta \mathcal{J}_{\mu}) \prod_{k=1}^{K-1} x_{\mu k} \right)^{2}} \right. \\ \left. \times \sum_{l=1}^{K-1} \frac{1 - x_{\mu l}^{2}}{x_{\mu l}} \times y_{\mu l} \right) \right\rangle_{\mathcal{J}_{\mu}, x_{\mu l}}$$
(13)

where $\langle \cdots \rangle_{\mathcal{J}_{\mu}, x_{\mu l}}$ denotes the average over \mathcal{J}_{μ} and $x_{\mu l}$ following eq. (2) and the stationary distribution of $\pi^{t}(x) = \pi(x)$, respectively, and the stability of the BP's fixed point can be characterized by whether the stationary solution $f^{t}(y) = f(y) = \delta(y)$ is stable or not in update (13). This formulation makes analytical investigation possible to a certain extent.

In order to connect eq. (13) to the existing analysis, let us first investigate the limit $C \to \infty$ for which much more results are known compared to the case of finite C. Due to the central limit theorem, the distribution of the field fluctuation can be assumed as a Gaussian $f^t(y) = (1/\sqrt{2\pi b^t}) \exp[-(y-a^t)^2/(2b^t)]$, where a^t and b^t are the mean and the variance of the distribution, respectively. Plugging this expression into eq. (13) offers update rules with respect to a^t and b^t as $a^{t+1} = (K-1)\beta J_0 M^{K-2}(1-Q)a^t$ and $b^{t+1} = (K-1)(\beta J)^2 Q^{K-2} \int Dz(1-\tanh^2(\sqrt{F}z+E))^2 b^t + ((K-1)(\beta J)^2 Q^{K-2} \int Dz(1-\tanh^2(\sqrt{F}z+E))^2 - (K-1)^2(\beta J_0)^2 M^{2K-4}(1-Q)^2)(a^t)^2$, where M, Q, E and F represent the convergent solutions of eqs. (9) and (10). In order to examine the stability of $f(y) = \delta(y)$, we linearize these equations around $a^t = b^t = 0$, which provides the critical condition of the instability with respect to the growth of b^t

$$(K-1)(\beta J)^2 Q^{K-2} \int Dz \left(1 - \tanh^2(\sqrt{F}z + E)\right)^2 = 1, \quad (14)$$

which becomes identical to that of the de Almeida–Thouless (AT) stability for the infinite range multi-spin interaction SG models and, in particular, for the SK model when K = 2.¹¹⁾ Furthermore, in the case of the SK model (K = 2), the critical condition with respect to a^t around the paramagnetic solution M = Q = 0 corresponds to the para-ferromagnetic transition. These mean that the two different phase transitions from the paramagnetic solution can be linked in a unified framework to the dynamical instabilities of BP by eq. (13).

When *C* is finite, one can numerically perform the stability analysis employing eq. (13), the detail of which will be reported elsewhere. In addition, analytical investigation also becomes possible for K = 2 as follows since transitions from the paramagnetic solution in this case occur due to the local instability.

For a small β , the paramagnetic solution $\pi(x) = \hat{\pi}(x) = \delta(x)$ ($m_{\mu l} = \hat{m}_{\mu l} = 0$) expresses the correct stable fixed point of the BP dynamics. Inserting this into eq. (13) does not provide a closed set of equations with respect to a finite number of parameters since $f^t(y)$ is no more a Gaussian. However, assuming $f^t(y) \simeq \delta(y)$, the stability analysis can be reduced to coupled equations with respect to the mean and the variance of $f^t(y)$ as $a^{t+1} = (C-1)\langle \tanh(\beta \mathcal{J}) \rangle_{\mathcal{J}} a^t$ and $b^{t+1} = (C-1)(\langle \tanh^2(\beta \mathcal{J}) \rangle_{\mathcal{J}} b^t + (\langle \tanh^2(\beta \mathcal{J}) \rangle_{\mathcal{J}} - \langle \tanh(\beta \mathcal{J}) \rangle_{\mathcal{J}}^2)(a^t)^2$). Linearizing these around $a^t = b^t = 0$ provides the critical conditions with respect to the growth of a^t and b^t as

$$(C-1)(\tanh(\beta \mathcal{J}))_{\sigma} = 1, \tag{15}$$

$$(C-1)\langle \tanh^2(\beta \mathcal{J}) \rangle_{\mathcal{J}} = 1, \tag{16}$$

respectively. It should be mentioned that these critical conditions around the *simple* paramagnetic solution have been already obtained in similar systems using perturbation methods.^{23–25)} However, the current scheme may be superior to the methods employed in the past since the expression (13) is compact and, therefore, can be easily applied to the stability analysis of the *non-trivial* ferromagnetic solution with the aid of numerical methods (Δ in Fig. 3) even in the case of multi-spin interaction ($K \ge 3$) while such extension in the other schemes requires higher order expansion and becomes highly complicated.

Equations (15) and (16) might correspond to the paraferromagnetic and the para-SG phase transitions, respectively, since they do in the limit $C \rightarrow \infty$. In order to examine this, we performed numerical experiments for N =2000 and C = 4. Although further investigation may be necessary to prove correctness, the data obtained from 100



Fig. 3. Phase diagram for K = 2 and C = 4 suggested by eqs. (15) and (16). P, F and SG stand for the paramagnetic, the ferromagnetic and the spin glass phases, respectively. The boundary between F and SG is a conjecture. In order to examine the validity of this diagram, 100 Monte Carlo experiments were performed for N = 2000 systems at conditions denoted by +. For each condition, frequencies of macroscopic magnetizations $M_a = (1/N) \sum_{l=1}^{N} m_l^a$ and overlaps $Q_{ab} = (1/N) \sum_{l=1}^{N} m_l^a m_l^b$ (a > b) were evaluated, where m_l^a is the average of S_l obtained from 20000 Monte Carlo steps per spin for experiments a, b = 1, 2, ..., 100(insets). For both of the two conditions in P, all of M_a and Q_{ab} fall into the first bin. On the other hand, sharp peaks indicate the order to the ferromagnetic state in F and a broad distribution of Q_{ab} signals the breaking of the replica symmetry in SG. Markers (Δ) in the ferromagnetic phase represent the critical condition with respect to the stability of the BP's fixed point numerically obtained from eq. (13), which may correspond to the AT instability in equilibrium of the non-trivial ferromagnetic solution.

experiments of 20000 Monte Carlo steps per spin exhibit good consistency with the analytical expressions (15) and (16) indicating that the correspondence between the phase transitions in equilibrium and the dynamical instabilities of BP holds for finite C as well (Fig. 3).

5. Summary

In summary, we have investigated dynamical behavior of BP when employed in SG models. We have shown that the time evolution of macroscopic variables can be well captured, even in the transient stage and even when the replica symmetry is broken in equilibrium, by recursive updates of auxiliary field distributions which becomes identical to the forward iteration of the saddle point equations under the RS ansatz in the replica analysis. We have further shown that the dynamical instability of the BP's fixed point is closely related to the AT instability of the RS solution, which has been numerically supported.

Relationship between the current scheme and an existing AT analysis for finite connectivity SG models²⁶⁾ that generally requires complicated calculation and is not frequently employed in practice is under investigation. Besides this, extending the current framework to the local stability analysis of the replica symmetry breaking solution^{27,28)} is a challenging and interesting future work.

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