I. INTRODUCTION

Self-assembly [1–3] is the spontaneous formation of complex ordered equilibrium structures from simpler component particles. The range of possible structures includes unusual crystals [4–7], viral capsids [8,9], and colloidal molecules [10–12]. Self-assembly is viewed as a promising alternative technology to current fabrication techniques, offering a bottom-up approach to design and manufacture [13]. For example, experimental work has used DNA to generate a backbone for potential nanofabrication [14,15], potential applications of artificial viral capsids include inert vaccines and drug delivery [16], and the potential range of self-assembled structures available through control of particle shape and interactions have been discussed extensively [2,9,10]. Self-assembly has even been considered for the potential regeneration of human organs and tissues [17]. However, a key challenge for design and control of self-assembly processes is that even in systems where the equilibrium states are known, the conditions that lead to effective assembly remain poorly understood and there remains no general theoretical approach to support experimental advances.

A major step toward such a theoretical approach has been the recognition of the importance of reversibility in the assembly process [9,18–21]. As a system evolves from an initially disordered state it makes bonds between the constituent particles. If structures form that are not typical of equilibrium configurations then they must anneal before the system arrives at equilibrium. When bonds between particles are too strong, thermal fluctuations are insufficient to break incorrect bonds before additional particles aggregate and the system becomes trapped in long-lived kinetically frustrated states. To avoid this problem and achieve effective self-assembly, bonds must be both made and broken as the system evolves in time. In this sense, self-assembling systems are typically reversible on short-time scales, even though they change macroscopically on long-time scales.

Recent work [19,20,22,23] has sought to quantify this qualitative idea by making dynamic measurements of reversibility. We will compare two approaches in a simple lattice-gas model where particles assemble into clusters and the effectiveness of self-assembly is identified by the presence (or absence) of undercoordinated particles in the clusters. First, we count individual bonding and unbonding events and compare their relative frequencies by defining dynamical flux and traffic observables [23]. Then we use measurements of responses to an applied field to probe reversibility, using fluctuation-dissipation theory combined with measurements of out-of-equilibrium correlation functions. We relate measurements of fluctuation-dissipation ratios (FDRs) to the reversibility of self-assembly, extending the analysis of Refs. [19,22]. In particular, we derive a formula for the response that elucidates its relation to microscopic reversibility and to measurements of flux and traffic [23]. We discuss how these results affect the idea [19,22] that measurements of reversibility on short-time scales might be used to predict long-time behavior.

II. MODEL

A. Definition

We use an Ising lattice gas as a simple model system for self-assembly. While this model appears much simpler than more detailed models of crystallization [4,7] or viral capsid self-assembly [9,20], previous work has shown that it mirrors many of the physical phenomena that occur in model self-assembling systems, particularly kinetic trapping [21,23,27].

At low temperatures, the equilibrium state of the lattice gas consists of large dense clusters of particles, but efficient assembly of these clusters requires reversible bonding in order to avoid kinetic trapping (particularly aggregation into ramified fractal structures [28]).

The model consists of a lattice, each site of which may contain at most one particle, and the energy of the system is

$$E = -\frac{\epsilon_b}{2} \sum_p n_p,$$  

where the summation is over all particles, $\epsilon_b$ is the interaction strength, and $n_p$ is the number of occupied nearest neighbors of particle $p$. We consider $N = 1638$ particles on a square lattice of dimension $d = 2$ and linear size $L = 128$. The relevant system parameters are the density $\rho = N/L^d \approx 0.1$ and the bond strength $\epsilon_b/T$ (or, equivalently, the reduced temperature $T/\epsilon_b$). The phase diagram is shown in Fig. 1, indicating the

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single fluid phase and the region of liquid-gas coexistence that occurs when \( \sinh^4(\epsilon_b/2T) > [1 - (2\rho - 1)^3]^{-1} \) [29]. At the density \( \rho = 0.1 \) at which we work, the binodal is located at \( T/\epsilon_b = 0.547 \) and below this temperature the system separates into high- and low-density phases at equilibrium (the critical point for the model is at \( \rho_c = 0.5 \) and \( T_c/\epsilon_b = 0.567 \)). For the analogy with self-assembly that we are considering, we are interested in behavior at fairly low densities, as in Refs. [9,10, 20–22]. All of the data that we show are taken at \( \rho = 0.1 \), but the same qualitative behavior is found for lower densities too. (At higher densities, we find that clusters of particles start to percolate through the system. We do not consider this regime since it would correspond to gelation in the self-assembling systems and this is not relevant for the systems we have in mind.)

We use a Monte Carlo (MC) scheme to simulate the diffusive motion of particles as they assemble. Our central assumptions are that (i) clusters of \( n \) particles diffuse with rates proportional to \( 1/n \) and (ii) bond making is diffusion limited (that is, bond-making rates depend weakly on the bond strength \( \epsilon_b/T \) while bond-breaking rates have an Arrhenius dependence on \( \epsilon_b/T \)). Our simulations begin from a random arrangement of particles and we simulate dynamics at fixed bond strength \( \epsilon_b/T \). The dynamics are based on the cleaving algorithm [30], which makes cluster moves in accordance with detailed balance and ensures physically realistic diffusion. We propose clusters of particles to be moved by picking a seed particle at random: Each neighbor of that particle is added to the cluster with probability \( 1 - \exp(-\lambda \epsilon_b/T) \), where \( \lambda \) is a parameter that determines the relative likelihood of cluster rearrangement and cluster motion. We set \( \lambda = 0.9 \) (see below). The cluster to be moved is built up by recursively adding neighbors of those particles in the cluster: each possible bond is tested once.

In addition, a maximum cluster size \( n_{\text{max}} \) is chosen for each proposed move and the move may be accepted only if the size of the cluster to be moved is less than \( n_{\text{max}} \). We choose \( n_{\text{max}} \) to be a real number, greater than unity, distributed as \( P(n_{\text{max}} > n) = 1/n^2 \). The result is that clusters of \( n \) particles have a diffusion constant \( D \propto 1/n \), which is consistent with Brownian dynamics; see Ref. [30] for full details.

Having generated a cluster, one of the four lattice directions is chosen at random and we attempt to move the cluster a single lattice spacing in that direction, rejecting any moves that cause particles to overlap. Finally, the proposed cluster move is accepted with probability \( P_a \), which depends on its energy change \( \Delta E \). For \( \Delta E \neq 0 \), we take a Metropolis formula \( P_a = \min(1, \exp((\lambda - 1)\Delta E/T)) \) while for \( \Delta E = 0 \) we take \( P_a = \alpha \) with \( \alpha = 0.9 \) a constant (see below). (Note that the definition of the energy in this model means that \( \Delta E/\epsilon_b \) is an integer, which ensures that \( P_a \) is monotonic in \( \Delta E \) for all values of \( \epsilon_b \) that we consider; cases where \( \Delta E/\epsilon_b \) is noninteger are discussed in Sec. III C.) Simulation time \( t \) is measured in Monte Carlo sweeps (MCSs), with one MCS corresponding to \( N \) attempted MC moves; the time associated with a single attempted MC move is therefore \( \delta t = 1/N \). We note that the constants \( \alpha \) and \( \lambda \) within this algorithm may be chosen freely between 0 and 1, while still preserving detailed balance. Our MC algorithm mimics diffusive cluster motion most closely when \( \alpha \) and \( \lambda \) are close to unity. However, when evaluating response functions (see below), the MC transition rates should be differentiable functions of any perturbing fields; this may not be the case if either \( \alpha \) or \( \lambda \) is equal to unity. We therefore set \( \alpha = 0.9 \) and \( \lambda = 0.9 \).

**B. Effectiveness of self-assembly**

In studying the lattice gas we have in mind a system self-assembling (or phase separating) into an ordered phase such as a crystal, coexisting with a dilute fluid. In the ordered phase, particles have a specific local coordination environment. In the lattice-gas model, we take this local environment to be that a particle has all neighboring sites occupied. We define the assembly yield \( n_a(t) \) as the proportion of particles with the maximal number of four bonds to neighboring particles. In Fig. 1(b) we plot the average yield at two fixed times. We observe a maximum at a reduced temperature that we denote by \( T^*/\epsilon_b \); the value of \( T^* \) depends only weakly on time \( t \). The system shows three qualitative kinds of behavior. To illustrate this, we show snapshots from our simulations in Fig. 2, at three representative bond strengths. For strong bonds, there is a decrease in yield as the system becomes trapped in kinetically frustrated states. For weak bonds, the system is already close to equilibrium at \( t = 10^6 \), but the final state has a relatively low value of \( n_a(t) \).

In the language of self-assembly, we interpret this as a poor quality product. Based on Figs. 1 and 2, we loosely identify the kinetically trapped regime as \( T^*/\epsilon_b < 0.2 \) and the regime of weak bonding and poor assembly as \( T^*/\epsilon_b \geq 0.5 \). However, we emphasize that these regimes are separated by smooth crossovers and not sharp transitions. In the following, we use \( T^*/\epsilon_b = 0.15, 0.35, \) and \( 0.5 \) as representative state points for kinetic trapping, good assembly, and poor assembly, respectively.

Thus, despite the simplicity of the lattice gas, it captures the kinetic trapping effects and the nonmonotonic yield observed in more physically realistic model systems [9,10,19–23]. We emphasize that the kinetically trapped states we find are closely related to diffusion-limited aggregates [28], while...
the assembled states are compact clusters. The changes in cluster morphology on varying bond strength are discussed further in Ref. [27], but in this paper we concentrate on the dynamical reversibility of assembly [18–20,22] and not on the structure of the clusters. We include in Fig. 1 the equilibrium behavior for systems of this size, which we obtain by running dynamical simulations starting from a fully phase-separated state in which a single large cluster contains all particles. As \( t \to \infty \), the yield must approach its equilibrium result: The laws of thermodynamics state that if we wait long enough then the highest-quality assembly (or the lowest-energy final state) will be at the lowest temperature.

III. MEASURING REVERSIBILITY

A. Flux-traffic ratio

We now turn to measurements of reversibility and their relation to kinetic trapping effects and the nonmonotonic yield shown in Fig. 1. We follow Ref. [23] in considering the net rate of energy-changing events at a microscopic level, the flux, in proportion to the total rate of energy-changing events, the traffic (see also Refs. [31–33]). We count an event (or kink) whenever a particle changes its number of neighbors; The number of times that particle \( i \) increases its number of bonds between times \( t \) and \( t + \Delta t \) is \( K^+_{i}(t, \Delta t) \), with \( K^-_{i}(t, \Delta t) \) the number of times the particle decreases its number of bonds. We then average and normalize by the time interval

\[
\tilde{k}^\pm(t, \Delta t) = \frac{1}{\Delta t} \langle K^\pm_{i}(t, \Delta t) \rangle. \tag{2}
\]

Here and throughout, averages \( \langle \cdots \rangle \) run over the stochastic dynamics of the system and over a distribution of initial configurations where particle positions are chosen independently at random. In the limit \( \Delta t \to 0 \), then the \( k^\pm(t, \Delta t) \) converge to the rates for bond-breaking (\( + \)) and bond-making (\( - \)) events, which we denote by \( k^\pm(t) \). The net flux \( f(t) = k^+(t) - k^-(t) \) is the net rate of bond-making events and the traffic is the total rate of all events \( q(t) = k^+(t) + k^-(t) \). In contrast to Ref. [23], we focus here on rates for making and breaking bonds and not on the total numbers of bonds made or broken.

We also define a flux-traffic ratio

\[
Q(t) = \frac{f(t)}{q(t)}, \tag{3}
\]

which provides a dimensionless measure of the instantaneous reversibility of a system, which is consistent with the qualitative description of reversibility due to Whitesides and Boncheva [18]. The inverse of the flux-traffic ratio \( 1/Q(t) \) is the number of energy-changing events per net-bonding event. In a system at equilibrium \( f(t) = 0 \) and so \( Q(t) = 0 \) also. For a system quenched to \( T = 0 \), our MC dynamics does not permit bonds to be broken so particles never increase their energy; hence \( k^-(t) = 0 \) and \( Q(t) = 1 \). In systems that have been quenched, we expect to see a value between these two limits, \( 0 < Q(t) < 1 \). The smaller the flux-traffic ratio the more reversible the system, while large flux-traffic ratios permit more rapid assembly.

In Fig. 3 we plot the flux \( f(t) \), traffic \( q(t) \), and their ratio \( Q(t) \) at a range of bond strengths \( 2 < \varepsilon_b/T < 10 \). In all cases, the flux decreases toward zero as the system evolves toward equilibrium. On this logarithmic scale, the temperature dependence of the flux appears quite weak, although the difference between different bond strengths may be up to an order of magnitude. Given that the integrated flux \( \int_0^{t_h} dt' f(t') \) is always less than the maximal number of possible bonds (4 in this case), it is clear that \( f(t) \ll t^{-1} \) at long times.

In contrast to the flux, the traffic shows a large variation with bond strength. Weaker bonds are more easily broken and result in more traffic. Combining flux and traffic, the ratio \( Q(t) \) is less than 0.1 throughout the good assembly regime and decreases approximately as \( t^{-1} \) while good assembly is taking place. The regime of kinetic trapping is characterized by larger values of \( Q(t) \) and weaker time dependence.

B. Flux relation between correlation and response functions

In designing and controlling self-assembly processes, it would be useful if reversibility could be measured and controlled to avoid kinetic trapping effects. While flux and traffic observables are not readily measured except in simple simulation models, we now show how correlation and response functions can be used to reveal similar information (see also...
Ref. [19,22]). These functions can be measured without the requirement to identify specific bonding and unbounding events; in some cases correlations and responses can also be calculated experimentally [34–36].

At equilibrium, fluctuation dissipation theorems (FDTs) [37] allow the response of a system to an external perturbation to be calculated from correlation functions measured in the absence of the perturbation. Away from equilibrium, such comparisons may be used for a classification in the dynamical approach to top and in (c) from top to bottom, as indicated in the legend. The flux varies little across the temperature range considered while the traffic is switched off. The system evolves with unperturbed dynamics up to a final time T, the ratio \( \overline{Q} \) remains relatively large throughout the trajectories, while the good assembly and high-temperature regimes are associated with reversible evolution and small values of \( \overline{Q} \).

We first consider the general case for measurement of a response function in a system with MC dynamics. Figure 4 illustrates our notation and the procedure used. The system is initialized at \( t = 0 \) and evolves up to time \( w \), when a perturbing field of strength \( h \) is switched on. A single MC move is attempted with the field in place, after which the field is switched off. The system evolves with unperturbed dynamics up to a final time \( t \) at which an observable \( A \) is measured. In general, the response function depends on the two times \( t \) and \( w \) and gives the change in the average value of \( A \) at time \( t \) is

\[
\langle A(t) \rangle = \sum_{\gamma \mu \nu} \frac{\delta A(t) \delta B(w)}{\langle \delta A(t) \delta B(w) \rangle}.
\]

We assume that the system obeys detailed balance with respect to an energy function \( E = E^0 - h B \), where \( E^0 \) is the energy of the unperturbed system and \( B \) is the conjugate observable to the field \( h \). It is convenient to define a connected correlation function

\[
C(t,w) = \langle \delta A(t) \delta B(w) \rangle
\]

where \( A(\gamma) \) is the value of \( A \) in configuration \( \gamma \) and we introduce \( u = t - w - \delta t \) for compactness of notation. [Since we use Greek letters to represent configurations, there should be no confusion between \( A(\gamma) \) and \( \langle A(t) \rangle \), the former being a property of configuration \( \gamma \) and the latter a time-dependent average.] The definition of the (impulse) response is

\[
R(t,w) = T \frac{\partial \langle A(t) \rangle}{\partial h}.
\]

where the derivative is evaluated at \( h = 0 \). Hence

\[
R(t,w) = T \left( \sum_{\gamma \nu} A(\gamma) G^0_{\nu}(\nu \leftarrow \nu) \frac{\partial}{\partial h} W^h(\nu \leftarrow \mu) \rho_w(\mu) \right)
\]

The MC scheme may be specified through the probabilities \( P^0(v \leftarrow \mu) \) for transitions from configuration \( \mu \) to \( v \) in a single attempted move. However, it is convenient to work with transition rates \( \dot{W}^0(v \leftarrow \mu) = (\delta t)^{-1} P^0(v \leftarrow \mu) \). We also define \( \rho_w(\mu) \) as the probability that the system is in the specific configuration \( \mu \) at time \( w \) and the propagator \( G^0_{\nu}(v \leftarrow \mu) \) as the probability that a system initially in state \( \mu \) will evolve to state \( v \) over a specified time period \( t \). The superscript 0 indicates that no perturbation is being applied to the system; we use a superscript \( h \) if the field is applied. [It follows that \( \rho_w(\mu) = \sum_\kappa G^h_{\nu}(\mu \leftarrow \kappa) \rho_0(\kappa) \), where \( \rho_0(\kappa) \) is the probability of initial condition \( \kappa \) in the dynamical simulations.]

With these definitions, the average value of the observable \( A \) at time \( t \) is

\[
\langle A(t) \rangle = \sum_{\gamma \mu \nu} \frac{A(\gamma) G^0_{\nu}(\nu \leftarrow v) P^h(v \leftarrow \mu) \rho_w(\mu),}{\langle \delta A(t) \delta B(w) \rangle}
\]

FIG. 3. (Color online) Plots of the flux \( f(t) \) and traffic \( \tau(t) \) and their ratio \( Q(t) \). For the early-time data in (a) and in the inset to (a), temperatures increase from top to bottom. At later times the behavior is nonmonotonic in temperature. In (b) temperatures increase from bottom to top and in (c) from top to bottom, as indicated in the legend. The flux varies little across the temperature range considered while the traffic depends much more strongly on the temperature. In the kinetic trapping regime \( T/\epsilon_b \ll 0.2 \), the ratio \( Q \) remains relatively large throughout the trajectories, while the good assembly and high-temperature regimes are associated with reversible evolution and small values of \( Q \).
(here and throughout we use the notation $\delta O = O - \langle O \rangle$ for all observables $O$). We also define

$$S(t, w) \equiv \frac{\partial}{\partial w} C(t, w).$$

(8)

(The MC dynamics evolves in discrete time steps; the interpretation of the time derivative is discussed in the Appendix.) At equilibrium, one has the FDT

$$R^{eq}(t, w) = S^{eq}(t, w).$$

(9)

Out of equilibrium, we define

$$\Delta(t, w) \equiv S(t, w) - R(t, w)$$

as a deviation between the response of the actual system and the response of an equilibrated system with the same correlation functions.

Various out-of-equilibrium expressions for response functions may be derived; see, for example, Refs. [33,38,39], where different representations are defined and compared with each other. In the Appendix we give a proof of Eq. (9) and we derive the general relation

$$\Delta(t, w) = \sum_{\nu} \left[ A(\nu \leftarrow \nu) \right] G_\nu(\nu \leftarrow \nu) F(v, \mu) J_\nu^0(v, \mu),$$

(11)

where

$$F(v, \mu) = T \frac{\partial}{\partial h} \ln W^h(\mu \leftarrow v)$$

(12)

and

$$J_\nu^0(\mu, v) = W_\nu^0(\mu \leftarrow v) \rho_\mu(\mu) - W_\nu^0(\mu \leftarrow v) \rho_v(v)$$

(13)

is the probability current between configurations $\mu$ and $v$ at time $w$. We will see below that the probability current $J^0$ plays a central role in quantifying irreversibility and deviations from equilibrium.

The formula (11) is general for discrete time Markov processes obeying detailed balance (the generalization to Markov jump processes in continuous time is straightforward). We emphasize that Eq. (11) is a representation of Eq. (6), valid whenever detailed balance holds. As such, it is mathematically equivalent to several other representations that have been derived elsewhere: For example, the asymmetry function discussed by Lippiello and co-workers [39,40] plays a role similar to $\Delta(t, w)$ in the case of Langevin processes. The purpose of Eq. (11) is to make contact between reversibility and deviations from FDT in self-assembling systems, as we now discuss. Further details of the relation between Eq. (11) and previous analyses [33,38–42] are discussed in Sec. III G below.

C. Energy correlation and response functions in the lattice gas

We now turn to response functions in the lattice gas. We consider how a single particle in the lattice gas responds if the strength of its bonds is increased. To this end, we write the energy in the presence of perturbing fields $h_p$ as $E^h = \sum_p \left[ \frac{1}{2} e_p - h_p \right] n_p$, where the sum runs over all particles, as in Eq. (1). We measure the response of $\langle n_p \rangle$ to the field $h_p$ by taking the observables $A$ and $B$ of the preceding section both equal to the number of bonds $n_p$ for a specific particle $p$. Thus $R(t, w) = \sum_p \frac{\partial \langle n_p \rangle}{\partial h_p}$. Given the energy function $E^h$, there is still considerable freedom to choose the MC rates $W^h$ while preserving detailed balance.

In Refs. [33,39,40], it was mathematically convenient to set $W^h(\nu \leftarrow \mu) = W_0^h(\nu \leftarrow \mu) e^{\lambda [P_h(\mu) - P_h(\nu)]}/2$. Here our dynamics are motivated by the central assumptions (i) and (ii) of Sec. II A, which indicate that rates for bond making and cluster diffusion should only weakly on perturbations to the particle bond strengths. We therefore include all $h$ dependence in the probability $P_h^D$ of accepting a MC move, as follows. If the change in the unperturbed energy for a MC move is $\Delta E^0$ and the change in the perturbation is $\Delta V = -\sum_p h_p \Delta n_p$, then for $\Delta E^0 \neq 0$ we set $P_h^D = \min(1, e^{\Delta E^0/\Delta V})$ while for $\Delta E^0 = 0$ we set $P_h^D = 2a/(1 + e^{\Delta V/\Delta V})$. It is easily verified that this choice is compatible with detailed balance and reduces to the unperturbed probabilities $P_h$ if $\Delta V = 0$. Further, coupling the field $h$ only to the acceptance probability ensures that the perturbation affects the rates for bond breaking and bond making but does not affect the rates for diffusion of whole clusters.

We use a straightforward generalization of the no-field method [41,42] to allow efficient estimation of the response; see the Appendix for details. To attain good statistics, we consider responses in which the perturbing field $h$ acts not just for one MC move but for a time interval $\Delta t = 10$ MC sweeps. Since we are working at leading order in $h$, the response to such a perturbation is simply

$$R_M(t, w) \equiv \frac{\delta t}{\Delta t} \sum_j R(t, w + j\delta t).$$

(14)

The relevant correlation function in this case is

$$\tilde{S}_{\Delta t}(t, w) = \langle \delta n_p(t) \delta n_p(w) \rangle,$$

where $\delta n_p(t) = n_p(t) - \langle n_p(t) \rangle$ as usual. It is convenient to define normalized correlation and response functions

$$\tilde{S}_{\Delta t}(t, w) = \frac{1}{\hat{N}_{\Delta t}(w)} [C(t, w + \Delta t) - C(t, w)],$$

(15)

$$\tilde{R}_{\Delta t}(t, w) = \frac{1}{\hat{N}_{\Delta t}(w)} R_{\Delta t}(t, w),$$

(16)

with $\hat{N}_{\Delta t}(w) = C(w + \Delta t, w + \Delta t) - C(w, w + \Delta t)$ so that $\tilde{S}_{\Delta t}(w + \Delta t, w) = 1$. We also define

$$\tilde{A}_{\Delta t}(t, w) = \tilde{S}_{\Delta t}(t, w) - \tilde{R}_{\Delta t}(t, w).$$

(17)

In all cases, the subscript $\Delta t$ indicates that these functions generalize the $R(t, w)$, $S(t, w)$, and $\Delta(t, w)$ of the preceding section, while the tilde indicates normalization. For small $\Delta t$, the dependence of these functions on $\Delta t$ is weak, but numerical calculation of the response is easier for larger $\Delta t$.

In Fig. 5 we show numerical results for correlation and response functions. At $T/\epsilon_p = 0.5$, clusters of particles are growing in the system, which is far from global equilibrium. However, as found in Refs. [19,22] for several other self-assembling systems, the deviations from FDT are small because the time evolution is nearly reversible (recall Fig. 3). In contrast, at $T/\epsilon_p = 0.15$, the particles are aggregating in disordered clusters and the time evolution is far from reversible with unbinding events being rare.
The potential utility of this result was discussed in Refs. [19,22]: It means that straightforward measurements on short-time scales can be used to predict long-time assembly yield, by exploiting links between correlation-response measurements and reversibility. In what follows, we explore in more detail how the deviation function \( \Delta \) couples to microscopic irreversibility during assembly.

D. Interpretation of \( \Delta(w + \delta t, w) \) as a flux

In this section we concentrate on quantities that can be written in the form

\[
Z(t) = \sum_{\nu, \mu} z(\nu, \mu) J^0_w(\nu, \mu),
\]

where \( J^0_w \) is the probability current, defined in Eq. (13). In contrast to straightforward one-time observables such as \( (A(i)) = \sum_{\mu} A(\mu) \rho_i(\mu) \), we will show that observables such as \( Z(t) \) are currents, time derivatives, and fluxes: They measure deviations from instantaneously reversible behavior, measured at time \( t \).

To this end, we generalize the analysis of kinks in Sec. III A by defining \( \langle K^{\nu \nu}(t, \Delta t) \rangle \) as the average number of MC transitions from state \( \mu \) to state \( \nu \) between times \( t \) and \( t + \Delta t \). The associated kink rate is \( k^{\nu \nu}(t) = (\Delta t)^{-1} \langle K^{\nu \nu}(t, \Delta t) \rangle \). It follows from the definition of \( k \) that

\[
k^{\nu \nu}(t) - k^{\mu \mu}(t) = J^0_w(\nu, \mu).
\]

Thus the probability current \( J^0_w \) gives the difference between the probabilities of forward and reverse MC transitions between states \( \mu \) and \( \nu \), which clarifies its connection to irreversibility. At equilibrium, one has \( J^0_w(\nu, \mu) = 0 \) for all states \( \mu \) and \( \nu \); thus \( J^0_w \) measures deviations from equilibrium. However, measuring (or even representing) \( J^0_w \) is a near-impossible task in a system where the number of possible states \( \mu \) is exponentially large in the system size.

Instead, one makes a specific choice of the \( z(\nu, \mu) \), in which case Eq. (18) shows that \( Z(t) \) is a projection of the current \( J^0_w \) onto the observables of the system. For example, if \( z(\nu, \mu) = A(\nu) \), then \( Z(t) = \frac{\partial}{\partial t}(A(t)) \) is a time derivative that clearly vanishes at equilibrium (or in any steady state). Other choices for \( z(\nu, \mu) \) become relevant in out-of-equilibrium settings. For example, the flux \( f(t) \) in Sec. III A is obtained by setting \( z(\nu, \mu) = 1 \) if the transition \( \mu \to \nu \) involves an increase in the number of bonds of particle \( p \), with \( z(\nu, \mu) = 0 \) for all other \( \mu, \nu \). [The structure of Eq. (18) together with this choice of \( z \) ensures that \( f(t) \) acquires negative contributions from MC transitions where particle \( p \) experiences a decrease in its number of bonds, as required.] Similarly, the particle current in exclusion processes is obtained by setting \( z(\nu, \mu) = 1 \) for transitions \( \mu \to \nu \) where a particle hops to the right and zero for all other pairs. Clearly, these observables monitor the breaking of time-reversal symmetry, as is the case for all observables \( Z(t) \).

To relate deviations from FDT to irreversible events, we first consider the case \( t = w + \delta t \), so that the perturbation \( h \) is applied for just one MC move and then the response is measured immediately. In this case Eq. (11) reduces to

\[
\Delta(w + \delta t, w) = \sum_{\nu, \mu} z_\Delta(\nu, \mu) J^0_w(\nu, \mu)
\]

as in Eq. (18), with

\[
z_\Delta(\nu, \mu) = [A(\nu) - \langle A(w + \delta t) \rangle] F(\nu, \mu).
\]

The function \( F(\nu, \mu) \) was defined in Eq. (12) and measures the effect of the perturbation \( h \) on the transition rate from \( \nu \) to \( \mu \). The factor \( A(\nu) - \langle A(w + \delta t) \rangle \) indicates whether configuration \( \nu \) has a high or a low value of the observable \( A \) compared to the average of \( A \) at the measurement time \( w + \delta t \).

For the case considered in Sec. III C and Fig. 5, where the perturbation is coupled to the energy of particle \( p \), one has \( A(\nu) = n_p(\nu) \) and

\[
F(\nu, \mu) = [n_p(\nu) - n_p(\mu)] \Theta(E^0(\mu) - E^0(\nu)),
\]

where \( \Theta(x) \) is the step function, defined such that \( \Theta(0) = \frac{1}{2} \). The step function appears because if a proposed MC move results in a decrease in the bare energy, the acceptance probability does not depend on the perturbing field \( h_p \). Clearly, \( F(\nu, \mu) \) is finite only if particle \( p \) changes its energy between states \( \mu \) and \( \nu \); the same is true for \( z_\Delta(\mu, \nu) \).
Thus Eq. (20) indicates that the deviation $\Delta(t,w)$ measured at $t = w + \delta t$ reflects the imbalance between rates of bond making and bond breaking for particle $p$. In this respect it is similar to the flux $f(w)$; however, the weights given to different bond-making and bond-breaking processes differ between $\Delta(w + \delta t, w)$ and $f(w)$ due to the different forms of $z(v,\mu)$ in the two cases. Thus, while these quantities reflect similar physics, the details of their behavior are different. To see the relationship between the immediate response behavior are different. To see the relationship between the immediate response and therefore reveal similar information about the (ir)reversibility of the dynamical bonding and unbonding processes.

\begin{equation}
\tilde{\Delta}_{\Delta t}(t + \Delta t, t)
\end{equation}

![Graph](image)

**FIG. 6.** (Color online) (a) Deviation from FDT measured immediately after perturbation, $\tilde{\Delta}_{\Delta t}(t + \Delta t, t)$, with $\Delta t = 10$; temperature increases from top to bottom. At temperatures above $T^\ast$, $\tilde{\Delta}_{\Delta t}(t + \Delta t, t)$ decreases quickly with $t$, while at low temperatures it remains significant throughout the early times. In the construction of $\Delta(t,w)$ we take the observables to be $A = B = n_p$, as discussed in the main text. (b) Parametric plot of $\tilde{\Delta}_{\Delta t}(t + \Delta t, t)$ against the flux-traffic ratio $Q(t)$ for $0 < t < 1500$. The relationship between these two quantities depends very weakly on temperature, indicating the close relation between them.

**E. Time intervals for reversibility**

We now consider the deviation $\Delta(t,w)$ for $t > w + \delta t$. From Eq. (11) we see that $\Delta(t,w)$ can be written in the form of Eq. (18) if we set

\begin{equation}
z(v,\mu) = \left[ P_u^A(v) - \langle A(t) \rangle \right] F(v,\mu),
\end{equation}

where

\begin{equation}
P_u^A(v) = \sum_\gamma A(\gamma) G_0(\gamma \leftarrow v)
\end{equation}

is the propensity [43] of observable $A$ for configuration $v$ after a time $u$. That is, $P_u^A(v)$ is the value of $A$ that is obtained by an average over the dynamics of the system for a fixed initial condition $v$ and a fixed time $u$. (Recall that $u = t - w - \delta t$.)

Comparing Eq. (23) with Eq. (21), the difference is the replacement of $A(v)$ by the propensity. At equilibrium, one expects $P_u^A(v) - \langle A(t) \rangle$ to decay toward zero as $u$ increases since the system forgets its memory of the configuration $v$ and regresses back to the mean. (That is, the propensities for different fixed initial conditions all converge to the same average value at long times.) Out of equilibrium, this may not be the case: If a configuration $v$ has a more ordered structure than another configuration $v'$, then these configurations may have different propensities for order even at much later times. Loosely, a system in $v$ has a head start along the route to assembly and the system does not forget the memory of this head start as assembly proceeds.

Figure 5 shows that for the correlation and response functions considered here, the deviation $\Delta(t,w)$ decays only very slowly with time $t$. The $t$ dependence of this function comes through a weighted sum of propensities; it is therefore clear that these propensities do not regress quickly to the mean. In contrast, one may write the correlation as

\begin{equation}
S(t,w) = \sum_{\nu\mu} \left[ P_u^A(v) - \langle A(t) \rangle \right] [n_\nu(v) - n_\nu(\mu)] \\
\times W_0(v \leftarrow \mu)p_\nu(\mu),
\end{equation}

whose $t$ dependence also comes from the propensity. Figure 5 shows that this function does decay quite quickly with $t$, although it does not reach zero on the time scales considered here.

Our conclusion is the following. For configurations $v$ that are typical at time $w$, the propensity $P_u^A(v)$ has a fast contribution that decays quickly with time $u$, as well as a slow contribution that depends weakly on $u$ and reflects the head start of $v$ along the route to the assembled product. The fast contribution reflects rapid bond-making and bond-breaking events that do not lead directly to assembly while the slow contribution is a nonequilibrium effect that measures assembly progress and also dominates the deviation $\Delta(t,w)$ that we have defined here. By contrast, $S(t,w)$ picks up contributions from both slow and fast contributions. The utility of the FDR is that for suitable $w$, it couples to the irreversible bond-making behavior that is most relevant for self-assembly.
To illustrate this balance of fast and slow degrees of freedom and also to make contact with previous studies [19, 22, 26] of FDRs, we define an integrated (and normalized) response
\[ \tilde{\chi}(t, w) = \frac{1}{C(t, t)} \int_0^t d w' R(t, w') \]
and we also normalize the correlation as
\[ \tilde{C}(t, w) = \frac{C(t, w)}{C(t, t)}. \]

It is conventional to display these correlation functions in a parametric form [26]. Specifically, making a parametric plot of \( \tilde{\chi}(t, w) \) against \( \tilde{C}(t, w) \) for fixed \( t \) and varying \( w \), the gradient is \(-X(t, w)\). It has been emphasized several times [44–46] that plotting data for fixed \( w \) and varying \( t \) is not in general equivalent to this procedure and may give misleading results, especially if \( C(t, t) \) has a significant time dependence, as it does in these systems. Plotting data at fixed \( t \) as in Fig. 7, it is apparent that the high-temperature systems are the most reversible, with \( X(t, w) \approx 1 \). To understand the connection with Fig. 5 note that the time interval \( t - w \) increases from right to left in Fig. 7: When \( t - w \) is large the deviation \( \Delta(t, w) \) has a larger fractional contribution to \( S(t, w) \), so that the curves are steepest (most reversible) when \( t - w \) is small and least steep (least reversible) when \( t - w \) is large. We find that parametric plots depend weakly on the fixed values of \( t \) used, but we do not show this data, for brevity; see, for example, Ref. [22] for an analysis of \( t \) dependence in a system of crystallizing particles.

Physically, we attribute the \( w \) dependence of \( X(t, w) \) to the fact that fast degrees of freedom (such as dimer formation and breakage in the vapor phase) quickly relax to a quasiequilibrium state [27] where probability currents \( J^0 \) associated with this motion are small. It is only much slower degrees of freedom (such as the gradual growth or assembly of large clusters) for which probability currents remain significant on long-time scales and lead to long-time contributions to \( \Delta(t, w) \).

Compared to the simple flux-traffic ratio \( Q(t) \) considered in Sec. III A, the normalized deviation \( 1 - X(t, w) = \Delta(t, w)/S(t, w) \) has the same effect of comparing a generalized flux \( \Delta(t, w) \) with a normalization \( S(t, w) \) that reflects the traffic in the system. However, the effect of the time difference \( t - w \) is that while \( \Delta(t, w) \) decays slowly with time \( t \), \( S(t, w) \) decays quite quickly. Thus, as \( t \) increases, the deviation \( 1 - X(t, w) \) becomes increasingly sensitive to the deviations from reversibility since the fluctuations of fast (reversible) degrees of freedom regress to the mean while the slow (irreversible) ones remain significant. For this reason, the FDR \( X(t, w) \) is a more sensitive measure of irreversibility than the flux-traffic ratio discussed in Sec. III A.

F. Responses to different perturbations

We remark that the observables used in measuring correlation and response functions may strongly affect the results. Most relevant is the extent to which the response function couples to the slow, irreversible degrees of freedom and the fast, reversible ones. To illustrate this, we compare the results of the preceding section with a different response function. We set both observables \( A \) and \( B \) of Sec. III B equal to the occupancy \( \rho_i = 0.1 \) of a specific site \( i \) on the lattice. The effect of the perturbation on the MC dynamics is the same as that described in Sec. III C, except that the perturbed energy \( E^a = E^0 + V \), with \( V = \sum h_i \rho_i \), where the sum runs over sites of the lattice.

Unlike the bond numbers \( n_p \), which change only when bonds are made and broken, the site occupancies \( \rho_i \) also change as clusters diffuse through the system. When bonds are strong, the site correlation functions decay faster than the bond correlation functions considered above. Therefore, since the site response is coupling to fast degrees of freedom, we expect to see more reversible behavior: this is borne out by Fig. 8, where the FDRs are much closer to the equilibrium value (unity) than the FDRs shown in Fig. 7. This is consistent with recent results of Russo and Sciortino [46], who observed a similar effect, and with earlier studies of the observable dependence of FDRs [44, 45, 47] (see, however, Ref. [48], in which some fast degrees of freedom do not relax to quasiequilibrium). Similar correlation and response measurements in an Ising lattice gas were also considered by Krzakała [49], but the dynamical scheme used in that work means that clusters of particles do not diffuse: In this case the site degrees of freedom \( \rho_i \) are not able to reach quasiequilibrium and one does observe significant deviations from FDT for this observable.

We end this section by briefly considering these correlation and response functions for large \( t \). In this limit, standard arguments [26] indicate that both correlation and response functions have contributions from well-separated fast and slow sectors, with the contributions from the fastest sector obeying FDT. In coarsening systems like this one then, there is no
an approach that is different from ours. Full details are given in the Appendix, Sec. 3, but in our notation, their central result is

\[ R(t, w) = \frac{1}{2} \left[ \frac{\partial}{\partial w} \langle A(t) B(w) \rangle + \langle A(t) T(w) \rangle \right], \tag{28} \]

where \( T(w) \) measures the \( h \) dependence of the amount of dynamical activity (traffic) between times \( w \) and \( w + \delta t \). The second term is therefore associated with traffic while the first term is a correlation between \( A(t) \) and the excess entropy production at time \( w \); it is therefore related to a flux. However, neither of the correlation functions in Eq. (28) may be written in the form of Eq. (18), so they do not vanish at equilibrium. The key point is that in equilibrated (reversible) systems, the terms in Eq. (28) are both nonzero and equal to each other. So while Eq. (28) does involve a separation of terms symmetric and antisymmetric under time reversal, the resulting correlation functions are not fluxes in the form of Eq. (18) and do not provide the same direct measurement of the deviation from reversibility given by \( \Delta(t, w) \) in Eq. (11).

In another relation to irreversible behavior, the analysis of Refs. [39,40] identifies an asymmetry measurement, which in our notation is

\[ A(t, w) = R(t, w) - \frac{1}{2} [S(1) + S(w, t)]. \tag{29} \]

It follows from the fluctuation-dissipation theory that this quantity vanishes in a system with time-reversal symmetry. A comparison with Eq. (10) also illustrates the similarity between \( A(t, w) \) and the deviation \( \Delta(t, w) \). However, \( A(t, w) \) differs from \( \Delta(t, w) \) since the deviation \( \Delta(t, w) \) is a projection of \( J^0_w \) and hence vanishes if the system is behaving reversibly at time \( w \), regardless of what happens at later times. In contrast, \( A(t, w) \) vanishes only if all trajectories between times \( w \) and \( t \) are equiprobable with their time-reversed counterparts. That is, time-reversal symmetry must hold throughout the interval between \( w \) and \( t \) and not just at time \( w \).

**IV. CONCLUSION**

We have analyzed the extent to which ideas of reversible bond making are applicable to a phase-separating lattice gas, which we used as a simplified model for a self-assembling system. The data of Figs. 1, 3, and 7 further support the idea that this system is relevant for studying self-assembly, since similar results have been presented for more detailed models of self-assembling systems [9,19,20,22,23].

The flux-traffic measurements of Sec. IIIA provide an intuitive picture of reversibility, while the picture based on the correlation-response formalism is more subtle. However, the central result (11) demonstrates an explicit link between measurements of response functions and irreversibility of bonding. Also, Figs. 5 and 7 do indicate that information about both short-time reversible and long-time irreversible behavior can be obtained by considering the behavior of correlation and response functions.

It has been argued that correlation-response measurements on short-time scales might be used to predict long-term assembly yield and even to control assembly processes. The work presented here places this objective on a firmer theoretical footing and highlights the importance of using the...
right observable when probing irreversibility (compare Figs. 7
and 8); the interplay between fast quasiequilibrated degrees
of freedom and slow irreversible bonding [27] also highlights
the importance of choosing the right time scale for measuring
these functions and predicting long-time assembly quality.

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APPENDIX: DERIVATIONS OF REPRESENTATIONS
FOR RESPONSE FUNCTIONS

1. Deviation function Δ(t, w)

In this section we derive Eq. (11), which gives the deviation
between the response R(t, w) and the correlation function
S(t, w). Starting from Eq. (6), conservation of probability for
the MC transition probabilities implies that

\[ P^b(t, \mu \leftarrow \mu) = 1 - \sum_{v \neq \mu} P^b(v \leftarrow \mu), \]

with a similar relation for the rates W^b. Using this in Eq. (6) gives

\[ R(t, w) = T \sum_{\gamma, v \neq \mu} A(\gamma)G^\rho_{\mu}(\gamma \leftarrow v) \frac{\partial}{\partial h} J^b_{w}(\mu, v), \quad (A1) \]

with

\[ J^b_{w}(\mu, v) = W^b(v \leftarrow \mu) \rho_w(\mu) - W^b(\mu \leftarrow v) \rho_w(v), \quad (A2) \]

which has the structure of a flux in probability from configuration
μ to configuration v. This is related [38] to an expression
for the response originally due to Agarwal [50].

We now assume that the system obeys detailed balance:

\[ W^b(\mu \leftarrow v)e^{-E^b(\mu)/T} = W^b(v \leftarrow \mu)e^{-E^b(\mu)/T}, \quad (A3) \]

where E^b(\mu) = E^0(\mu) - hB(\mu), as defined in Sec. III B.

Hence one has \( \frac{\partial}{\partial h} \ln W^b(\mu \leftarrow v) = \Delta B(\mu, \mu)/T + \frac{\partial}{\partial h} \ln W^b(\mu \leftarrow v), \) with

\[ \Delta B(\mu, \mu) = B(\mu) - B(\mu), \]

so that

\[ T \frac{\partial}{\partial h} J^b_{w}(\mu, v) = \Delta B(\mu, \mu)W^0(\mu \leftarrow v) \rho_w(\mu) \]

\[ + T J^b_{w}(\mu, v) \frac{\partial}{\partial h} \ln W^b(\mu \leftarrow v). \quad (A4) \]

Inserting Eq. (A4) into Eq. (A1) gives two contributions
to the response. The first contribution is

\[ \frac{1}{T}(A(t)[B(w + \delta t) - B(w)]). \]

In the notation of the main text, this is equal to

\[ S(t, w) + (A(t) \frac{\partial}{\partial h} B(w)), \]

where the time derivative is interpreted as a change between w
and w + \delta t, normalized by the time \delta t for a single MC move.
(Recall that we took time to be discrete in these MC models.)

At equilibrium, \( J^b = 0 \) and \( \frac{\partial}{\partial h} B(w) = 0 \), so
the only contribution to the response is \( S(t, w) \) and the FDT (9) holds.
The nonequilibrium contributions to \( R(t, w) \) can be collected
together to obtain Eq. (11) if one notes that (i) \( \sum_{\mu} G^\rho_{\mu}(\gamma \leftarrow v) = 1 \), which follows from conservation of probability and
the definition of \( G^0 \), and (ii)

\[ \sum_{\mu, v} F(v, \mu)J^b_{w}(v, \mu) = \frac{\partial}{\partial w}(B(w)), \quad (A5) \]

which follows from the detailed balance property of the \( W^b \).

2. No-field formula for the response

Another useful representation of the response function expresses
it in terms of directly observable quantities in unperturbed simulations
[41,42] (this is the path weight representation of Ref. [38]).
The original no-field method requires an extension for our system since
the cluster MC algorithm we use means that the same transition
\( v \leftarrow \mu \) may take place via several possible computational routes.
(For example, two MC moves may start from different seed
particles but result in the same cluster being moved in the
same direction.) We use \( P(v \leftarrow \mu) \) to represent the probability
of a MC move from \( \mu \) to \( v \) by some route \( C \). For \( v \neq \mu \), the
route \( C \) is the choice of seed particle and sequence of steps
by which the moving cluster is generated. If \( v = \mu \) the route
\( C \) may involve the proposal of a move that is then rejected or
the proposal of a move where the cluster size exceeds \( n_{\text{max}} \)
or would result in multiple occupancy of a site.

Starting from Eq. (6), one writes

\[ R(t, w) = T \sum_{\gamma, v, C, \mu} A(\gamma)G^{\rho}_{w}(\gamma \leftarrow v)R(v, C, \mu) \]

\[ \times P^c(v \leftarrow \mu)\rho_w(\mu), \quad (A6) \]

where \( R(v, C, \mu) = \frac{1}{T} \frac{\partial}{\partial h} \ln W^b(v \leftarrow \mu). \) This result may be
written as a simple expectation value

\[ R(t, w) = T(A(t)\bar{R}(w)), \quad (A7) \]

where the key point is that the observable \( \bar{R}(v, C, \mu) \) may be
evaluated for any attempted MC move so that the correlation
function in Eq. (A7) may be calculated directly from a MC
simulation with \( h = 0 \). [This is in contrast to the current
\( J^b_{w}(v, \mu) \), which depends on the whole ensemble of evolving
systems, so that the formula (A1) cannot be evaluated directly
by MC simulation.]

For the dynamical method described in this paper, the observable
\( \bar{R}(v, C, \mu) \) is given by \( \bar{R} = [n_p(v) - n_p(\mu)]\theta(E^0(\mu) - E^0(v)) \)
if the move is accepted. We take \( \theta(0) = \frac{1}{2} \), as in the main text. If the move is rejected because
either the cluster size \( n \) exceeds \( n_{\text{max}} \) or the move
would have resulted in overlapping particles \( R = 0 \). If, however,
the move is legal but rejected when testing the move acceptance
probability \( P_a \), then \( R = -[n_p(v) - n_p(\mu)]\theta(E^0(\mu) - E^0(v))P_a/(1 - P_a) \).

3. Alternative formula for the response

In this section we clarify the connections between
our results and those of Baiesi et al. [33]. The central
identity in the derivation of Sec. 1 of the Appendix is
\[ \frac{\Delta}{\partial h} \ln W(\nu \leftrightarrow \mu) = \Delta B(\nu, \mu)/T + \frac{\Delta}{\partial h} \ln W(\mu \leftrightarrow \nu). \]

Rearranging, one may write
\[ \frac{\Delta}{\partial h} \ln W(\nu \leftrightarrow \mu) = \Delta B(\nu, \mu)/2T + \frac{\Delta}{\partial h} \ln [W(\mu \leftrightarrow \nu)W(\nu \leftrightarrow \mu)] \] (A8) and substitution into Eq. (6) leads to
\[ R(t, w) \equiv \frac{1}{2} \sum_{\gamma, \nu} A(\gamma)G_{\alpha}(\gamma \leftrightarrow \nu) \times [\Delta B(\nu, \mu) + T(\nu, \mu)]W(\nu \leftrightarrow \mu) \rho_{\alpha}(\mu). \] (A9)

with
\[ T(\nu, \mu) = T \frac{\Delta}{\partial h} \ln [W(\mu \leftrightarrow \nu)W(\nu \leftrightarrow \mu)]. \]

Equation (A9) is the translation of the central result of Biaesi et al. [33] into our notation.

The key point here is that the two terms in square brackets in Eq. (A9) have different symmetry properties. By definition of the first term, \( \Delta B(\nu, \mu) = -\Delta B(\mu, \nu) \). Time reversal of the trajectory flips the roles of \( \nu \) and \( \mu \) so that \( \Delta B \) is antisymmetric under time reversal and is related to the entropy current [33]. This means that the average of \( \Delta B \) is zero in a time-reversal symmetric system. (Note, however, that \( \Delta B \) itself is a fluctuating quantity and not zero in general; this is different from the probability current \( J^0_w \), which is an ensemble-averaged property without fluctuations and is strictly equal to zero if time reversal holds.) In contrast \( T(\nu, \mu) = T(\mu, \nu) \) so \( T \) is symmetric under time reversal and measures dynamical activity, or traffic.

Despite the links between Eq. (A9) and time-reversal symmetries, neither of the terms in Eq. (A9) may be written as a flux in the form of Eq. (18). Noting that \( S(t, w) \) is a connected correlation function, one may write
\[ R(t, w) = \frac{1}{2} \left[ S(t, w) + \langle A(t)T(w) \rangle + \langle A(t) \frac{\partial}{\partial w} \langle B(w) \rangle \right], \] (A10)

where the final term vanishes in the nonequilibrium steady states considered in Ref. [33] while \( T(w) \) is given by \( (\delta t)^{-1} T(\nu, \mu) \), evaluated with \( \nu \) and \( \mu \) being the configurations at times \( w + \delta t \) and \( w \), respectively. The key point is that the first two terms on the right-hand side of Eq. (A10) are equal to each other and both are nonzero at equilibrium. Hence neither of them is directly identifiable as a flux that is sensitive only to deviations from reversibility. Combining the first and third terms on the right-hand side of Eq. (A10) yields Eq. (28) of the main text.