We consider reversible and irreversible stochastic growth processes in the simplest limit, ignoring spatial degrees of freedom and resolving only the numbers of particles in the system. By ‘reversible’ we mean simply that particles may enter and leave the system, and we intentionally do not require that rates are derived from the principle of detailed balance. We consider growth of a system composed of two types of particle, labeled ‘red’ and ‘blue’. The state of the system is defined at any instant by the number of red particles $r$ and blue particles $b$ it contains, or equivalently by the system’s ‘size’ $N \equiv b + r$ and ‘magnetization’ $m \equiv (b - r)/(b + r)$. We add blue particles to the system with rate $\lambda_b$, and red particles with rate $\lambda_r$. We remove blue and red particles from the system with respective rates $\gamma_b$ and $\gamma_r$. For an irreversible process these latter two rates are zero. We allow rates to depend on the instantaneous magnetization of the system but not (directly) its size. We impose this requirement in order to model a notional growth process in which rates of particle addition and removal to a structure scale with the size of the interface between the structure and its environment. We then assume the limit of a large structure whose interfacial area does not change appreciably during the growth process, and we divide addition and removal rates by the (constant) surface area in order to obtain the rates stated above.

We studied this class of growth processes using a continuous-time Monte Carlo protocol \cite{9}. To interpret these simulations we derived a set of analytic expressions for averages over dynamic trajectories, in the limit of

\[ N \equiv b + r \quad \text{and} \quad m \equiv (b - r)/(b + r). \]
vanishing particle-number fluctuations (see Appendices 1-3). Under these conditions the net rates of addition of blue and red particles are \( \Gamma_b(m) = \lambda_b - \gamma_b \) and \( \Gamma_r(m) = \lambda_r - \gamma_r \). The time evolution of system size is \( \dot{N} = \Gamma_b + \Gamma_r \). The requirement for equilibrium, by which we mean a zero-growth-rate stationary solution, is \( \Gamma_b(m_0) = 0 = \Gamma_r(m_0) \), where \( m_0 \) is the magnetization of the system in equilibrium. These relations can be satisfied by a reversible process but not an irreversible one, except in the trivial limit of zero addition rate. Thus only a reversible process has an equilibrium for which \( \dot{N} = 0 = \dot{m} \). However, both types of process admit a steady-state growth regime for which \( \dot{N} = 0 \) and \( \dot{m} = 0 \). The time evolution of magnetization is \( \dot{m} = N^{-1} [\Gamma_b - \Gamma_r - m(\Gamma_b + \Gamma_r)] \), which vanishes for \( m = m_0 \) such that

\[
\dot{m}_0 = \frac{\Gamma_b(m_0) - \Gamma_r(m_0)}{\Gamma_b(m_0) + \Gamma_r(m_0)}.
\]

Thus there exist nullclines, at which \( \dot{m} = 0 \), in the space of dynamic growth trajectories. The existence of such nullclines requires only that net rates of particle addition are positive, whether or not removal rates vanish, and so can be displayed by reversible and irreversible processes. We shall show that these nullclines can be attractors, stable with respect to perturbations, and so constitute fixed lines to which dynamic trajectories flow. Furthermore, these attractors undergo nonequilibrium phase transitions as model parameters are varied.

We now specialize the discussion to two models that might be regarded as reversibly- and irreversibly growing versions of the mean-field Ising model. The irreversible stochastic process we consider is a space-independent version of the magnetic Eden model \([1, 2, 3]\), closely related to a model studied in Ref. \([10]\). Addition of red and blue particles occurs with rates that are Arrhenius-like in the energy of interaction between a single particle and the system, \( \lambda_b = \frac{1}{2} e^{-mJ-h} \) and \( \lambda_r = \frac{1}{2} e^{mJ+h} \). Here \( J \) is the interparticle coupling and \( h \) a magnetic field (we set \( m = 0 \) when \( N = 0 \)). We allow no particle removals, setting \( \gamma_b = \gamma_r = 0 \). There is therefore no equilibrium. The analytic evolution equations, averaged over trajectories, read \( \dot{N} = \cos(mJ + h) \) and

\[
\dot{m} = N^{-1} [\sinh(mJ + h) - m \cosh(mJ + h)],
\]

and admit the nullcline

\[
m_0 = \tanh(m_0 J + h).
\]

This equation is equivalent to the well-known mean-field expression for Ising model thermodynamics \([11, 12]\). For \( h = 0 \), Equations (2) and (3) indicate that the nullcline \( \dot{m}_0 = 0 \) is an attractor for \( J \leq J_c = 1 \) and a repeller for \( J > 1 \). For \( J > 1 \) two symmetric attractors emerge as \( m_0 \pm \sim \pm (J - 1)^{1/2} \). In other words, these equations describe a continuous phase transition of dynamic trajectories that are ‘unmagnetized’ for \( J < 1 \) and ‘magnetized’ for \( J > 1 \), via a critical point at \( J = 1 \). Thus, at mean-field level, nonequilibrium trajectories of the magnetic Eden model possess phase behavior identical to that of the equilibrium Ising model \([13]\). This result provides an analytic connection between models supporting the findings of Refs \([7, 8]\), which demonstrated a numerical equivalence between phase transitions, in distinct spatial dimensions, of Eden and Ising models (see also \([2]\)). This result also appears to be consistent with general arguments suggesting that probabilistic irreversible cellular automata with Ising-like symmetry lie in the universality class of the equilibrium Ising model \([14]\) (see also \([6, 15]\)).

The reversible model we consider is the stochastic process whose fluctuation-free limit is described in Refs. \([16, 17]\). We assume constant rates of particle addition, \( \lambda_b = pc \) and \( \lambda_r = (1-p)c \), where \( c \) is a notional ‘solution’ concentration and \( p \) is the fraction of particles in solution that are blue. To make contact with Ising model nomenclature we introduce the magnetic field \( h \) via \( p = e^h/(2 \cosh h) \). Unbinding rates are Arrhenius-like, appropriate for particles escaping from a structure via thermal fluctuations, and are \( \gamma_b = \frac{1}{2} e^{-mJ}(1 + m) \) and \( \gamma_r = \frac{1}{2} e^{mJ}(1 + m) \) (supplemented by the restriction that particle numbers cannot be negative). Note that these rates are intentionally not designed to satisfy detailed balance with respect to a particular energy function; however, the process still possesses an equilibrium. The fluctuation-free evolution equations are \( \dot{N} = c - \cosh(mJ) + m \sinh(mJ) \) and

\[
\dot{m} = N^{-1} [(1 - m^2) \sinh(mJ) - c (m - \tanh h)].
\]

Equilibrium is achieved when \( c^2_0 = (1 - m^2_0) \cosh^2 h \), with

\[
m_0 = \tanh(m_0 J + h).
\]
show that dynamic trajectories feel the influence of the dynamic attractors predicted analytically. In the magnetized region trajectories ‘flow’ to one of the two stable magnetized attractors, while at the critical point the stable attractor is unmagnetized. Individual trajectories fluctuate increasingly slowly in \( m \)-space as \( N \) increases (because, for large \( N \), fluctuations of particle number change magnetization by an amount \( \propto N^{-1} \)), and so e.g. the likelihood of a magnetized trajectory spontaneously reversing its magnetization becomes vanishingly small (see Ref. [10]). However, ensembles of trajectories show behavior characteristic of a phase transition. In Fig. 3(a,b) we show averages of \( |m(t)| \) over \( 10^5 \) dynamic trajectories generated at several different values of \( J \). We define averages of an observable \( A(t) \) as \( \langle A(t) \rangle \equiv K^{-1} \sum_{i=1}^{K} A_i(t) \), where \( A_i(t) \) is the value of \( A(t) \) for the \( i \)th of \( K \) trajectories. Trajectory averages evolve as \( t \) increases toward the attractor. This evolution is in general slow, because the mobility of individual trajectories is low: ignoring fluctuations we expect small departures \( \langle \delta m(t) \rangle \) from the attractor to decay–above, at, and just below the critical point–as \( t^{-\gamma} \) (in \( t^{-1/2} \), and \( t^2 \), respectively, where \( q = 1 - J \) for the irreversible process and \( q = (c - J)/(c - 1) \) for the reversible one.

Trajectory-to-trajectory fluctuations, which are neglected by our analytic expressions, also show behavior characteristic of a phase transition. In Fig. 3(c,d) we show the trajectory-to-trajectory fluctuations of magnetization, \( \chi \equiv \langle (N(t))((m^2(t)) - \langle m(t) \rangle^2) \rangle \) (the quantity \( \text{var}(M)/\langle N \rangle \), where \( M = mN \), behaves similarly). For both models \( \chi \) displays at the critical point a peak that increases in height with average system size as \( \langle N(t) \rangle^{0.82(1)} \) over the time interval shown (see inset to Fig. 3(b)). While individual trajectories flow to stable attractors as time increases, thereby causing \( \text{var}(m) \) to decrease with time, the same trajectories also acquire more particles, and the combination \( \langle N \rangle \text{var}(m) \) increases with time over the interval simulated. Such ‘sharpening’ of a phase transition with increasing time is reminiscent of behavior seen in glassy models that display phase transitions in space-time [18]. In the asymptotic limit (when \( N \to \infty \) and \( m = m_* \) is constant) we expect the evolution of \( M = mN \) to resemble that of a random walker, and so \( \text{var}(M) \propto t \). Thus ensembles of trajectories feel the effect of dynamic attractors, but can in certain regimes of phase space take considerable time to reach them. It would be valuable to have deeper analytic understanding of such non-stationary behavior.

In some limits the two types of process can be clearly distinguished. All growth processes must eventually end. A bacterial colony will run out of food, and a self-assembled structure will come to equilibrium with its environment. In this limit the difference between reversible and irreversible processes becomes apparent. In Fig. 4 we show dynamic simulations of the reversible model carried out in the presence of an additional rule that forbids any addition that would cause the system to contain more...
FIG. 3. Trajectory-to-trajectory averages (a,b) and variance (c,d) of magnetization taken over ensembles of $10^5$ trajectories, at various values of $J$, for (a,c) reversible ($c = 2$) and (b,d) irreversible growth. The two types of process display similar phase transitions. Numerical simulations are overlaid on the analytic results (6) (a) and (3) (b). In panel (a) we also show the results of simulations done in the presence of a system size constraint, overlaid on (5) (see Fig. 4).

than $10^3$ particles. During the growth phase dynamic trajectories fall under the influence of the dynamic attractor, but when the system size limit is reached trajectories evolve to an attractor similar to that of the equilibrium one; see also the line labeled ‘constrained’ in Fig. 3(a). Trajectories of the irreversible model, in the presence of a size constraint, simply cease to evolve. The behavior of the reversible model gives insight into the behavior of the lattice models of growth of Refs. [16, 17]. These models obey detailed balance, and so must eventually evolve to equilibrium, but during a period of growth they display nonequilibrium behavior consistent with that of a persistently-growing process. The present results indicate that one can consider dynamic trajectories of a reversible growth process to feel the effect of both nonequilibrium and equilibrium attractors, the relative influence of which varies over the lifetime of the trajectory.

### III. CONCLUSIONS

We have shown at mean-field level that reversible and irreversible growth processes are similar in that both admit attractors in the space of dynamical trajectories. At these attractors growth proceeds without limit, but averaged intensive properties of the system are time-invariant. Attractors of both types of process can undergo similar nonequilibrium phase transitions. We have also established a connection at mean-field level between an irreversible model of growth (the magnetic Eden model) and the equilibrium Ising model, supporting the findings made by other authors using numerical simulations. There is sustained interest in nucleation and growth pathways of molecular [19], active [20] and living [1] matter. Our results indicate that certain qualitative properties of nonequilibrium growth trajectories are insensitive to details of the microscopic transition rates that produce them, suggesting a unified way of describing growth processes of distinct microscopic entities.
FIG. 4. Evolution of $\langle N \rangle$ (a) and $\langle |m| \rangle$ (b) for size-limited reversible growth shows that trajectories fall under the influence of the dynamic attractor while growth persists, and evolve to the equilibrium attractor when the size constraint is reached. Lines denote averages over 500 trajectories. Parameters: $J = 1.5$, $c = 2$. The line labeled ‘constrained’ in Fig. 3(a) shows the results of similarly constrained simulations for several values of $J$.

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APPENDIX 1. STOCHASTIC MODELS OF GROWTH

The stochastic processes described in the main text can be described using a master equation. Consider the probability $P(r, b; t)$ that a system at time $t$ contains $r \geq 0$ red and $b \geq 0$ blue particles. For brevity we will write this probability
as \( P(r, b) \), with the time-dependence of the function being implicit. We add red particles to the system with rate \( \lambda_r(r, b) \) and blue particles with rate \( \lambda_b(r, b) \), and remove red and blue particles with respective rates \( \gamma_r(r, b) \) and \( \gamma_b(r, b) \). The master equation for this set of processes is

\[
\partial_t P(r, b) = \lambda_b(r, b - 1)P(r, b - 1) - \lambda_b(r, b)P(r, b) + \lambda_r(r, b - 1)P(r - 1, b) - \lambda_r(r, b)P(r, b) + \gamma_b(r, b + 1)P(r, b + 1) - \gamma_b(r, b)P(r, b) + \gamma_r(r + 1, b)P(r + 1, b) - \gamma_r(r, b)P(r, b).
\]  
(A1)

We set \( \gamma_b(r, 0) = 0 = \gamma_r(0, b) \) so that we cannot have a negative number of particles of either color. By multiplying both sides of (A1) by the arbitrary function \( U \) and summing over \( b \) and \( r \) we obtain the evolution equation for the quantity \( U \) averaged over dynamic trajectories, \( \langle U(r, b) \rangle \equiv \sum_{r,b=0}^{\infty} U(r, b)P(r, b) \):

\[
\partial_t \langle U(r, b) \rangle = \langle [U(r, b + 1) - U(r, b)]\lambda_b(r, b) \rangle + \langle [U(r + 1, b) - U(r, b)]\lambda_r(r, b) \rangle + \langle [U(r, b - 1) - U(r, b)]\gamma_b(r, b) \rangle + \langle [U(r - 1, b) - U(r, b)]\gamma_r(r, b) \rangle.
\]  
(A2)

Setting \( U(r, b) = b \) gives the rate of change of the mean number of blue particles,

\[
\partial_t \langle b \rangle = \langle \lambda_b(b, r) \rangle - \langle \gamma_b(b, r) \rangle.
\]  
(A3)

The corresponding equation for red particles is

\[
\partial_t \langle r \rangle = \langle \lambda_r(b, r) \rangle - \langle \gamma_r(b, r) \rangle.
\]  
(A4)

We can obtain closed-form equations for rates of change of particle number by making a mean-field approximation, replacing averages over functions \( f \) of \( r \) and \( b \) with functions \( f \) of the averages of \( r \) and \( b \), i.e. writing \( \langle f(r, b) \rangle = f(\langle r \rangle, \langle b \rangle) \). To simplify notation we then replace \( \langle r \rangle \rightarrow r \) and \( \langle b \rangle \rightarrow b \), so that (A3) and (A4) read

\[
\dot{b} = \lambda_b(r, b) - \gamma_b(b, r);
\]  
(A5)

\[
\dot{r} = \lambda_r(b, r) - \gamma_r(b, r).
\]  
(A6)

The size of the system is \( N = r + b \), and so its growth rate is

\[
\dot{N} = \dot{r} + \dot{b} = \lambda_b + \lambda_r - \gamma_b - \gamma_r.
\]  
(A7)

In equilibrium we must have the vanishing of (A5) and (A6), giving

\[
\lambda_b = \gamma_b
\]  
(A8)

and

\[
\lambda_r = \gamma_r.
\]  
(A9)

The rate of change of magnetization \( m \equiv (b - r)/(b + r) \) is

\[
\dot{m} = \frac{1}{N} \left[ \dot{b} - \dot{r} - m(\dot{b} + \dot{r}) \right] = \frac{1}{N} \left[ \lambda_b - \lambda_r - \gamma_b + \gamma_r - m(\lambda_b + \lambda_r - \gamma_b - \gamma_r) \right].
\]  
(A10)

The condition \( \dot{m} = 0 \) implies

\[
m_* = \frac{\lambda_b - \lambda_r - \gamma_b + \gamma_r}{\lambda_b + \lambda_r - \gamma_b - \gamma_r},
\]  
(A11)

in which all rates are understood to be evaluated at \( m = m_* \).
APPENDIX 2. IRREVERSIBLE MODEL OF GROWTH

The irreversible model described in the main text allows no particle removal, \( \gamma_b(r,b) = 0 = \gamma_r(r,b) \). Blue particles are added with an Arrhenius-like rate that assumes Ising-like interaction interaction energies between red and blue particles with coupling \( J \) and magnetic field \( h \) (we take \( k_B T = 1 \)),

\[
\lambda_b(b,r) = \frac{1}{2} \exp \left\{ J \frac{1 + m}{2} - J \frac{1 - m}{2} + h \right\} = \frac{1}{2} e^{mJ+h}.
\]  
(A12)

Here the spatial mean-field approximation is apparent, i.e. particles ‘feel’ only the average magnetization of the whole system. We have absorbed the particle coordination number, assumed to be constant, into \( J \). Similarly, red particles are added to the system with rate

\[
\lambda_r(b,r) = \frac{1}{2} \exp \left\{ J \frac{1 - m}{2} - J \frac{1 + m}{2} - h \right\} = \frac{1}{2} e^{-mJ-h},
\]  
(A13)

The averaged growth rate (A7) is

\[
\dot{N} = \cosh(mJ + h).
\]  
(A14)

The averaged time evolution of the system’s magnetization, (A10), is

\[
\dot{m} = N^{-1} [\sinh(mJ + h) - m \cosh(mJ + h)].
\]  
(A15)

This vanishes for

\[
m_* = \tanh(m_* J + h).
\]  
(A16)

Equations (A15) and (A16) are equations (2) and (3) of the main text.

The temporal evolution to the attractor (A16) differs in different regimes of parameter space. Consider the case of vanishing magnetic field. For a small departure \( \delta m \) from the attractor, \( m(t) = m_* + \delta m(t) \), we have from (A14) \( N \approx \cosh(m_* J) t \). Inserting this result into (A15) and using (A16) gives

\[
\dot{\delta m} \approx -\frac{1}{t} \left\{ \delta m \cosh(J\delta m) + [m_*^2 + m_* \delta m - 1] \sinh(J\delta m) \right\}.
\]  
(A17)

Expanding this equation in powers of \( \delta m \) gives

\[
\dot{\delta m} \approx \frac{1}{t} \left\{ [J(1 - m_*^2) - 1] \delta m - J m_* (\delta m)^2 + \frac{J^2}{6} [J(1 - m_*^2) - 3] (\delta m)^3 \right\}.
\]  
(A18)

In the unmagnetized region \( (J < 1) \) we have \( m_* = 0 \), and so \( \dot{\delta m} \approx (J - 1) \delta m / t \). Thus temporal relaxation to the attractor is algebraic, with a continuously varying exponent: \( \delta m \sim t^{J-1} \). In the magnetized region \( (J > 1, m_* \neq 0) \) relaxation to steady-state is also algebraic, \( \delta m \sim t^{J(1-m_r^2)-1} \). Close to the critical point, where \( J \approx 1 \), we have from (A16) that \( m_*^2 \approx 3(J - 1)/J^3 \), and so \( \delta m \sim t^{2(1-J)} \) to leading order in \( J - 1 \). Thus the (moduli of) exponents either side of the critical point are distinct. At the critical point we have \( J = 1 \) and \( m_* = 0 \), in which case the first two terms on the right-hand side of (A18) vanish. We then have \( \dot{\delta m} \propto - (\delta m)^3 / t \), and so \( \delta m \sim (\ln t)^{-1/2} \).

APPENDIX 3. REVERSIBLE MODEL OF GROWTH

For the reversible model we have constant rates of particle addition, \( \lambda_b(r,b) = pc \) and \( \lambda_r(r,b) = (1-p)c \), where \( c \) is a notional ‘solution’ concentration and \( p \) is the fraction of particles in solution that are blue. To make contact
with Ising model nomenclature we introduce the magnetic field \( h \) via 
\[
p \equiv e^h/(2 \cosh h).
\]
Particle unbinding rates are Arrhenius-like, appropriate for particles escaping from a structure via thermal fluctuations (we take \( k_B T = 1 \)):
\[
\gamma_b(b,r) = \frac{1 + m}{2} \exp \left\{ -\epsilon_s \frac{1 + m}{2} - \epsilon_d \frac{1 - m}{2} \right\} \\
= \frac{1 + m}{2} e^{-\Sigma - m \Delta},
\] (A19)
where the magnetization of the system is again 
\[
m \equiv \frac{(b - r)}{(b + r)}.
\]
We assume that blue particles possess energy of interaction \( -\epsilon_s \) with blue particles and \( -\epsilon_d \) with red particles (we have absorbed factors of coordination number, assumed to be constant, into these energetic parameters). We have defined the parameters 
\[
\Sigma \equiv \frac{(\epsilon_s + \epsilon_d)}{2} \quad \text{and} \quad \Delta \equiv \frac{(\epsilon_s - \epsilon_d)}{2}.
\]
The prefactor of the exponential ensures that blue particles leave the system with a rate proportional to their relative abundance in the system. For red particles we choose the unbinding rate
\[
\gamma_r(b,r) = \frac{1 - m}{2} \exp \left\{ -\epsilon_s \frac{1 - m}{2} - \epsilon_d \frac{1 + m}{2} \right\} \\
= \frac{1 - m}{2} e^{-\Sigma + m \Delta},
\] (A20)
Note that because \( m \) is not defined for \( r = b = 0 \) we additionally require \( \gamma_b(0,0) = 0 = \gamma_r(0,0) \).
Hence \((A5)\) and \((A6)\) become
\[
\dot{b} = pc - \frac{1 + m}{2} e^{-\Sigma - m \Delta},
\] (A21)
\[
\dot{r} = (1 - p) c - \frac{1 - m}{2} e^{-\Sigma + m \Delta},
\] (A22)
which, with \( p = 1/2 \), are Equations (1) of Ref. [16]. It is convenient to rescale time and concentration
\[
t \rightarrow e^\Sigma t
\] (A23)
and
\[
c \rightarrow e^{-\Sigma} c
\] (A24)
to get the simpler equations
\[
\dot{b} = pc - \frac{1 + m}{2} e^{-m \Delta},
\] (A25)
\[
\dot{r} = (1 - p) c - \frac{1 - m}{2} e^{m \Delta}.
\] (A26)
The growth rate \((A7)\) is
\[
\dot{N} = c - \cosh(m \Delta) + m \sinh(m \Delta).
\] (A27)
In this model there exists an equilibrium when rates of particle addition and removal balance. The the associated equation of state follows from \((A8)\) and \((A9)\), and is
\[
m_0 = \tanh(m_0 \Delta + h)
\] (A28)
with the associated concentration
\[
c_0^2 = (1 - m_0^2) \cosh^2 h.
\] (A29)
Note that the equilibrium concentration for \( h = 0 \) is the same for red \( (m_0 < 0) \) and blue \( (m_0 > 0) \) solutions, i.e. \( c_0 \) is unchanged upon setting \( m_0 \rightarrow -m_0 \).
The rate of magnetization evolution, Eq. (A10), is

\[ \dot{m} = \frac{1}{N} \left[ (1 - m^2) \sinh(m\Delta) - c (m - \tanh h) \right], \tag{A30} \]

which vanishes when

\[ c (m - \tanh h) = (1 - m^2) \sinh(m\Delta). \tag{A31} \]

In the main text we assume an Ising-like hierarchy for the interaction energies of this model, in which case \( \Delta = J \) and \( \Sigma = 0 \). With these choices Equations (A28), (A30), and (A31) are equations (3), (4), and (5) of the main text.

Analysis of (A31), for \( h = 0 \), gives rise to Fig. 1(a) of the main text. The function on the right-hand side of (A31) vanishes at \( m = 0 \) and at \( m = \pm 1 \), and has one turning point for positive \( m \) and one for negative \( m \). When \( \Delta < \sqrt{6} \) this function takes its largest positive gradient, \( \Delta \), at the origin. Therefore it intersects the function \( cm \) on the left-hand side of (A31) three times if \( c < \Delta \) (with two non-negative solutions, \( m_\pm \), stable to perturbations of \( m \), and one, at \( m = 0 \), unstable) and only once (at \( m = 0 \)) if \( c > \Delta \). When \( c = \Delta \) all solutions lie at \( m = 0 \). The solutions \( m_\pm \) move smoothly away from \( m \) as \( c \) is decreased below \( \Delta \), and do so as

\[ m_\pm \sim \left( \frac{6}{\Delta} \cdot \frac{\Delta - c}{6 - \Delta^2} \right)^{1/2}. \tag{A32} \]

Thus at the point \( c = \Delta \) (for \( h = 0 \) and \( \Delta < \sqrt{6} \)) we have a continuous nonequilibrium phase transition separating zero- and nonzero magnetization solutions to Eq. (A31).

For \( h = 0 \) and \( \Delta \geq \sqrt{6} \) we can have zero, three or five solutions to (A31), depending on the value of \( c \), and respectively zero, two and three of those solutions are stable.

As for the irreversible model, temporal relaxation to the attractor \( m_\star \) varies by parameter regime. Expanding (A27) and (A30) for \( m(t) = m_\star + \delta m(t) \) gives, for \( m_\star = 0 \),

\[ \partial_t \delta m \approx \frac{1}{t} \left\{ \frac{\Delta - c}{c - 1} \delta m - \frac{\Delta(6 - \Delta^2)}{6(\Delta - 1)} (\delta m)^3 \right\}. \tag{A33} \]

In the unmagnetized region \( c > \Delta \) we then have \( \delta m \sim t^{\frac{\Delta-c}{c-1}} \). At criticality (\( \Delta = c \)) we have \( \delta m \sim (\ln t)^{-1/2} \).

Expanding (A27) and (A30) for \( m(t) = m_\star + \delta m(t) \) and using (A32) reveals that in the magnetized region we have \( \delta m \sim t^{\frac{6c-\Delta}{c-1}} \).