An infinite-period phase transition versus nucleation in a stochastic model of collective oscillations

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Received 16 June 2011
Accepted 21 August 2011
Published 22 September 2011

Online at stacks.iop.org/JSTAT/2011/P09023
doi:10.1088/1742-5468/2011/09/P09023

Abstract. A lattice model of three-state stochastic phase-coupled oscillators has been shown by Wood et al (2006 Phys. Rev. Lett. 96 145701) to exhibit a phase transition at a critical value of the coupling parameter, leading to stable global oscillations. We show that, in the complete graph version of the model, upon further increase in the coupling, the average frequency of collective oscillations decreases until an infinite-period (IP) phase transition occurs, at which point collective oscillations cease. Above this second critical point, a macroscopic fraction of the oscillators spend most of the time in one of the three states, yielding a prototypical nonequilibrium example (without an equilibrium counterpart) in which discrete rotational ($C_3$) symmetry is spontaneously broken, in the absence of any absorbing state. Simulation results and nucleation arguments strongly suggest that the IP phase transition does not occur on finite-dimensional lattices with short-range interactions.

Keywords: classical Monte Carlo simulations, critical exponents and amplitudes (theory), finite-size scaling, stationary states
1. Introduction

Models of coupled oscillators exhibit a surprising range of symmetry-breaking transitions to a globally synchronized state. In the paradigmatic Kuramoto model, for instance, oscillators with intrinsic frequencies $\omega_i$ coupled via their continuous phases $\theta_i$ can exhibit stable collective oscillations, with breaking of time-translation symmetry [1]–[5]. When $\omega_i = 0$, the Kuramoto model acquires an additional $\theta_i \rightarrow -\theta_i$ reflection symmetry, which can also be broken at the onset of synchronization, yielding surprising response properties [6].

Within the class of discrete phase models, the paper–scissors–stone game is an example of a system with three absorbing states which can present either global oscillations or spontaneous breaking of discrete rotational ($C_3$) symmetry [7]–[11]. More recently, Wood, Van den Broeck, Kawai and Lindenberg proposed a family of models of phase-coupled three-state stochastic oscillators which can undergo a phase transition to a synchronized state [12]–[15] for sufficiently strong coupling. From here on each of them is called a Wood et al cyclic model (WCM). Although the WCM also has $C_3$ symmetry, it has no absorbing state.

As noted previously [14], the first WCM [12] has an unusual behavior after the phase transition to a synchronized state. As the phase coupling is further increased, collective oscillations slow down while oscillators remain partially synchronized. Here we have a closer look at this phenomenon, in particular addressing whether or not there is a second phase transition with breaking of $C_3$ symmetry.

The paper is organized as follows. In section 2 we review the WCM and the essentials of the known transition to a synchronized phase. Our results regarding a putative second phase transition are described in section 3. We conclude in section 4.

doi:10.1088/1742-5468/2011/09/P09023
2. The model

In the WCM, the phase $\phi_x$ at site $x$ ($x = 1, \ldots, N$) can take one of three values: $\phi_x = j_x 2\pi/3$, where $j_x \in \{0, 1, 2\}$ (for simplicity, from here on we employ $x$ to denote sites and $j$ to denote states—always modulo 3). The only allowed transitions are those from $j$ to $j+1$ (see figure 1), with transition rates

$$g_{j,j+1} = g \exp \left[ a \frac{(N_{j+1} - N_j)}{z} \right],$$

where $g$ is a constant (which can be set to unity without loss of generality), $a$ is the coupling parameter, $N_j$ is the number of nearest neighbors in state $j$, and $z$ is the number of nearest neighbors. The rates are invariant under cyclic permutation of the indices, i.e., belong to the symmetry group $C_3$ of discrete rotations, and strongly violate the detailed balance condition.

The mean-field approximation is obtained by replacing $N_j/z$ in the argument of the exponential of equation (1) with the corresponding probability, $P_j$, leading to the following set of equations:

$$\dot{P}_j = g_{j-1,j}P_{j-1} - g_{j,j+1}P_j,$$

where now $g_{j,j+1} = g \exp[a(P_{j+1} - P_j)]$. These also represent the equations for a complete graph in the limit $N \to \infty$: since $z = N - 1$, we can replace $N_j/N$ with $P_j$ in the infinite-size limit. Normalization reduces these equations to a two-dimensional flow in the $(P_1, P_2)$ plane (figure 2).

Starting from the complex order parameter,

$$v = \frac{1}{N} \sum_{x=1}^{N} e^{i\phi_x},$$

one can define quantities that characterize the collective behavior of the system. A configuration with $v \neq 0$ has unequal numbers of sites in the three states, which led

doi:10.1088/1742-5468/2011/09/P09023
Figure 2. Phase space \( (P_1, P_2) \). Gray triangles correspond to the allowed region \( 0 \leq P_2 \leq 1 - P_1 \). \( a \) increases from left to right. In the upper panels, dashed (solid) lines denote the nullclines \( \dot{P}_1 = 0 \) (\( \dot{P}_2 = 0 \)). Lower panels show trajectories from complete graph simulations with \( N = 5000 \) (thick lines) and solutions of the mean-field equations (thin lines with arrows). Panel (a) shows that \( \vec{P}_{1/3}^* \) is the only fixed point in the system for \( a = 1.6 > a_c \). (b) \( \vec{P}_{1/3}^* \) is unstable, with trajectories converging to a stable limit cycle, which corresponds to sustained global oscillations (see also figure 3(a)). (c) For \( a = 3 \), the nullclines approach one another in three additional positions, thereby creating three symmetric saddle-node ghosts. (d) The limit cycle is almost as large as the perimeter of the allowed region. Note the overcrowding of lines near the ghosts, with consequent anharmonicity in the time series (figure 3(b)). (e) For \( a = 3.25 \) there are six additional fixed points which are born simultaneously in three saddle-node bifurcations at \( a = a_c \). (f) In this case, if the system starts from the symmetric unstable fixed point \( \vec{P}_{1/3}^* \), fluctuations determine to which stable fixed point it goes. Closed (open) symbols denote stable (unstable) fixed points.

Kuramoto to propose

\[
    r \equiv \langle |v| \rangle_t \langle s \rangle
\]

as the order parameter for synchronization [1, 3, 12], where \( \langle \rangle_t \) denotes a time average over a single realization (in the stationary state), and \( \langle \rangle_s \) an average over independent realizations. Note that \( r > 0 \) is consistent with, but does not necessarily imply, globally synchronized oscillations. The latter situation is characterized by a periodically varying phase of \( v \) [16]–[19].

In the mean-field equations (2), the transition to the synchronized regime is associated with a supercritical Hopf bifurcation at \( a = a_c = 1.5 \). The trivial fixed point
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\[ \vec{P}_{1/3} = (P_1 = 1/3, P_2 = 1/3) \] loses stability at \( a = a_c \), and a limit cycle encircling this point appears (figures 2(a) and (b)) [12]. For \( a \geq a_c \), sustained oscillations in \( P_q \) characterize synchronization among the oscillators (figure 3(a)). Correspondingly, \( r \) grows continuously, \( \sim (a - a_c)^\beta \), at the transition (figure 3(c)), with a mean-field exponent \( \beta = 1/2 \) [12]. The scaled variance \( \chi_r \equiv L^d \langle \langle |v| \rangle^2 \rangle_s - r^2 \) diverges with the system size at criticality, as shown for simulations of complete graphs using \( L^d = N \) in figure 3(d).

Note that the phase transition to stable collective oscillations is associated with the breaking of the continuous time-translation symmetry: the time series \( P_0(t), P_1(t) \) and \( P_2(t) \) become periodic for \( a \geq a_c \). But since they are statistically identical, except for a phase, \( C_3 \) symmetry still holds. Increasing \( a \) above \( a_c \) enhances synchronization among the oscillators, leading to increasing oscillation amplitudes, as shown in figures 2(d) and 3(b).

3. The second phase transition

3.1. Mean-field theory and the complete graph

Wood et al. found that the increasing amplitude of oscillation is accompanied by a decreasing angular frequency \( \omega = 2\pi/\langle \tau \rangle \), where \( \langle \tau \rangle \) is the mean time between peaks in \( P_k \) (figures 3(a)–(c)). This can be understood qualitatively from the exponential dependence of the rates of equation (1) on the neighbor fractions: when a state is highly populated, the rate at which oscillators leave this state becomes very small. In addition, the rate at which a highly populated state attracts oscillators from its predecessor is also very high.

We study the effect of a further increase in the coupling \( a \), beyond the range of values investigated in [12], using mean-field theory and analysis on a complete graph of \( N \) sites. Note that the state of the latter is completely specified by two integer random variables, \( N_1 \) and \( N_2 \). We study the complete graph (for \( N \) up to 1000) via exact (numerical) stationary solution of the master equation [20] and (for \( N \) ranging from 10 to \( 10^4 \)) via event-driven Monte Carlo simulations. The results of the two approaches are fully consistent.

In mean-field theory, when \( a \) reaches an upper critical value \( a^c \approx 3.10243991564 \), three symmetric saddle-node bifurcations occur simultaneously, and the period of the collective oscillations diverges. Above \( a^c \), there are three symmetric attractors in the system, and threefold rotational \( (C_3) \) symmetry is spontaneously broken: figure 2(f) illustrates how the final fate of three different trajectories starting from the same initial conditions is determined by fluctuations. Moreover, the transition can be considered reentrant, in the sense that time-translation symmetry, which had been broken at \( a_c \), is restored at \( a^c \) (although, unlike in other cases seen in the literature [21, 22], the phase above \( a^c \) is not exactly the same as that below \( a_c \), because of \( C_3 \) symmetry breaking).

Analogously to what occurs in a condensation or ferromagnetic phase transition [23], the freezing of the relative occupancy of each state does not imply that individual oscillators freeze as well. The frequencies of individual oscillators do decrease with increasing \( a \), but only vanish in the limit \( a \to \infty \), when one of the states is fully occupied.

It is convenient to define an order parameter that is identically zero (in the infinite-size limit) for \( a > a^c \). Since the angular frequency \( \omega \) does not fulfill the usual requirements for an order parameter [24], we propose

\[ |\psi| \equiv \frac{1}{N} \sum_{x=1}^{N} \left( \delta_{0,jx} + e^{2\pi i/3} \delta_{1,jx} + e^{-2\pi i/3} \delta_{2,jx} \right) \gamma_x, \tag{5} \]

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Figure 3. Panels (a) and (b) show the evolution of $P_1$ for $a = 1.6$ and 3, respectively. Points represent simulations for the complete graph with $N = 10000$ whereas lines show the mean-field solution. (c) Dependence of $r$ and $\omega$ on $a$, exhibiting the two phase transitions. (d) $\chi_r$ versus $a$ showing peaks at the transitions (system sizes as in (c)).
where $\delta$ is the Kronecker delta and $\gamma_x \equiv g_{j_x, j_{x+1}}$ is the transition rate at site $x$ (see equation (1)). Thus $|\psi|$ reflects not only the configuration, but also the rate at which the latter evolves.

On the complete graph, $\gamma_x$ is the same for all sites $x$ in the same state $j$. Denoting this rate by $\gamma_j$, the order parameter can be rewritten for mean-field (MF) analysis as

$$|\psi|^2_{\text{MF}} = \left[ 2 \sum_{j=0}^{2} P_j^2 \gamma_j^2 \right] - P_0 \gamma_0 P_1 \gamma_1 - P_1 \gamma_1 P_2 \gamma_2 - P_2 \gamma_2 P_0 \gamma_0. \quad (6)$$

Note that both the disordered phase ($a < a_c$) and the frozen phase ($a > a_c$) have stable fixed points, so $P_j^* \gamma_j = P_{j+1}^* \gamma_{j+1}$ from equation (2). This in turn yields $|\psi| = 0$ in equation (6) for both cases (a similar line of reasoning can be applied directly to equation (5)).

Figure 4 shows the variation of $\langle |\psi| \rangle$ ($\equiv \langle \langle |\psi| \rangle \rangle_s$) with $a$ in mean-field theory, and on the complete graph, confirming that $|\psi|$ functions as an order parameter for detecting both phase transitions. The MF critical behavior is $\langle |\psi| \rangle \sim (a - a_c)^{1/2}$ for $a \gtrless a_c$ and $\langle |\psi| \rangle \sim (a^c - a)^{1/2}$ for $a \gtrless a^c$. In the pair approximation for a lattice with coordination number $z = 6$ we find $a_c = 1.97579$ and $a^c = 4.71727$; for $z = 8$ the corresponding values are $a_c = 1.80198$ and $a^c = 4.10028$. In the pair approximation, no transition is observed for $z \leq 4$.

On the complete graph, the order parameter decays with the system size as $\langle |\psi| \rangle \sim N^{-1/4}$ at $a = a_c$ and as $N^{-0.4203(3)}$ at $a = a^c$, as shown in figure 5. The first result is typical of mean-field-like scaling with the system size at a continuous phase transition, and can be understood as follows. Our numerical results show that with increasing $N$, as expected, the stationary probability distribution in the $P_1 - P_2$ plane becomes (for $a = a_c$) increasingly concentrated around the symmetric point $(1/3, 1/3)$. The maximum of the
distribution, \( p_{\text{max}} \), scales as \( N^{-3/2} \) and the number of points at which the probability is relatively large (i.e., \( p(N_1, N_2) \geq p_{\text{max}}/2 \)) has \( \sim N^{3/2} \) growth. In terms of the densities \( P_i = N_i/N \), the probability distribution is concentrated on a region of radius \( \rho \sim N^{-1/4} \).

Note that \(|\psi|^2\) is a quadratic form of the variables \( \xi_i \equiv P_i - 1/3 \) and that \( \psi = 0 \) for \( \xi_1 = \xi_2 = 0 \). Thus for the purposes of scaling analysis we may write the stationary average of \(|\psi|\) as

\[
\langle |\psi| \rangle \sim \int_0^{N^{-1/4}} \frac{\xi^2 \, d\xi}{\int_0^{N^{-1/4}} \xi \, d\xi} \sim N^{-1/4}
\]

with \( \xi = \sqrt{\xi_1^2 + \xi_2^2} \). At the first transition (\( a = a_c \)) the same argument applies to the order parameter \( r \), and we indeed find numerically \( r \sim N^{-0.26} \).

At the second transition, the probability distribution is peaked near the three stable fixed points of the mean-field analysis, and so has a form very different than that found at \( a_c \). It is therefore not surprising that \( \langle |\psi| \rangle \) decays with a different exponent. The value of 0.4203(3) (which does not correspond to any simple ratio of integers) nevertheless remains something of a puzzle. We suspect that understanding this result will require an asymptotic (large-\( N \)) analysis of the Fokker–Planck equation for the probability density in the \( P_1-P_2 \) plane, a task we defer to future work.

doi:10.1088/1742-5468/2011/09/P09023
3.2. Hypercubic lattices

Motivated by the appearance of a novel phase transition in mean-field theory and on the complete graph, we search for such a transition on finite-dimensional structures, i.e., the simple cubic lattice and its four-dimensional (hypercubic) analog, in systems of \( N = L^d \) sites. We do observe on these lattices a marked tendency for \( \omega \) to fall rapidly with \( a \) for values well above the first transition at \( a_c \) (\( a_c \simeq 2.345, \simeq 1.900 \) and \( \simeq 1.750 \) for \( d = 3, 4 \) and 5, respectively [12]). We also observe a broad maximum in \( \chi_r \) at some point in this range. To infer the existence of a phase transition, rather than a rapid but smooth variation of \( \omega \) and other properties with \( a \), we require evidence of emergent singularities as the linear system size \( L \) is increased.

The global dynamics may be characterized in terms of a process that begins with all sites in state \( j \), and ends when the fraction of sites remaining in this state falls to \( \leq 1/3 \). Let the ‘escape time’ \( T \) denote the mean duration of this process, averaged over realizations of the stochastic dynamics. If there is an infinite-period phase transition at some critical value \( a_c \), then \( T \) should show \( \sim L^z \) growth at this point, where \( z \) denotes the dynamic critical exponent. Although simulations reveal an exponential growth of \( T \) with \( a \) at fixed \( L \), they show that for large, fixed \( a \), the escape time decreases with increasing system size! The mean period (see figure 6) exhibits two regimes for \( a > a_c \): for smaller \( a \), relatively slow, exponential growth with \( T \sim e^a \) essentially independent of system size, and, for larger \( a \), rapid exponential growth with \( T \sim L^{-d} e^{\kappa a} \), with \( \kappa \gg 1 \). We identify the latter regime as that of single-droplet nucleation [25]. In the former regime, the density of ‘advanced’ sites (i.e., in state \( j + 1 \) in a configuration with the majority in state \( j \)) is large enough that essentially no barrier to the cyclic dynamics exists. The (smooth) crossover between these regimes occurs at a larger \( a \) value for larger system sizes.

The results for the mean transition rate \( T \) on the four-dimensional hypercubic lattice, \( T \sim L^{-d} \exp[\kappa a] \), clearly suggest a nucleation mechanism. The exponential factor represents a barrier to nucleation, that is, the mean time to the formation of a critical cluster. (By a ‘critical’ cluster we mean one that is equally likely to grow as to diminish in size; larger clusters tend to grow while smaller ones tend to shrink.) Consider a region in which all spins are in the same state. The rate at which a given spin flips to the next state (e.g., \( 1 \rightarrow 2 \)) is \( e^{-a} \), leading to the exponential growth in \( T \). The prefactor \( 1/L^d \) corresponds to essentially independent contributions due to each site in the system. (Given the extremely low density of clusters in this regime, we can treat the nucleation and initial growth of a cluster as occurring in isolation, without interference from other clusters.)

The constant \( \kappa \) in the exponential reflects the probability of formation of a critical cluster, which in turn depends on the critical cluster size, \( \ell^* \). Intuitively, the nucleation barrier exists because an isolated advanced site has a higher rate for advancing once again, and then rapidly returning to the majority state, than it does for inducing a neighbor to advance. Given the difficulty of visualizing four-dimensional space, we begin by studying nucleation on the square lattice. Of course, the model does not exhibit an ordered phase in two dimensions, but this issue can be bypassed by using small systems (a lattice of \( 10 \times 10 \) sites, in the present case). Then the system remains near one of the three attractors (almost all sites in the same state) except for the rapid transitional periods following a nucleation event. (In larger systems, global synchronization is quickly lost and

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Figure 6. Four-dimensional hypercubic lattice: inset: mean period $T$ versus $a$ for (upper to lower) $L = 28, 32, 40, \text{and } 50$; main graph: scale period $T^* \equiv L^d T$; system size increasing upwards.

the system breaks into a number of domains.) On such a small lattice, one may easily identify transitional regimes between, for example, states with majority 1 and majority 2. We define the transition (in the macroscopic sense) as the moment when 20% of sites are in state 1, with nearly all the rest (70% or more) in state 2. To estimate $T$ we determine the time required for a large number (4000, in practice) of transitions to occur, in event-driven Monte Carlo simulations. A typical evolution is shown in figure 7. We find that the mean transition time grows exponentially with $a$; for $a \geq 7$ we find $\kappa \approx 1.68(1)$. This suggests that the critical cluster size is approximately 3–4.

Several other pieces of evidence support this estimate. First, in the pre-transition regime, the number $N_2$ of sites in state 2 is seen to fluctuate between 0 and 3 or so. In the example shown in figure 7, the second time $N_2$ grows to 4, a transition occurs. Typical formats of a growing cluster are shown in figure 8; the cluster is seen to be compact, as expected for a growth probability that is highest for the sites outside the cluster having the largest number of neighbors within it.

A semiquantitative analysis of the cluster size distribution in the pre-transition regime can be developed as follows. Consider first the density $\rho_2$ of isolated sites in state 2, in a background of 1s. The rate for a given 1 to change to 2 is $e^{-a}$, while an isolated 2 (with all neighbors in state 1) flips to state 3 at unit rate. There are also transitions of the form

\begin{equation}
\ln T^* = \frac{1}{\kappa} \ln a + C
\end{equation}

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where $C$ is a constant. For $a \geq 7$, we find $\kappa \approx 1.68(1)$, suggesting that the critical cluster size is approximately 3–4.
21 → 22. The total rate for an isolated 2 to become a dimer is $4e^{-a/2}$. Thus we have a rate equation of the form

$$\dot{\rho}_2 = e^{-a} \rho_1 - \rho_2 - 4e^{-a/2} \rho_1 \rho_2 \simeq e^{-a} - \rho_2$$

leading to $\rho_2 \simeq e^{-a}$ in the quasi-stationary (QS) regime. By the QS regime we mean the long-lived metastable state in which the great majority of sites are in state 1. Similarly,
let \( \rho_{22} \) denote the probability that a given site is the leftmost (or lowermost) site of an isolated dimer (the restriction is to avoid double counting). Then we have

\[
\dot{\rho}_{22} \simeq 4e^{-a/2}\rho_2 - 2e^{-a/4}\rho_{22} - O(e^{-a/2}\rho_{22})
\]

so \( \rho_{22} \simeq 2e^{-5a/4} \) in the QS state. Consider next the density \( \rho_L \) of L-shaped trimers. The gain term in the equation for \( \rho_L \) is \( 4e^{-a/2}\rho_{22} \) while the dominant loss term (due to the formation of a four-site, square cluster) is simply \( -\rho_L \) (the transition rate is unity). Thus \( \rho_L \simeq 8e^{-7a/4} \). The exponential factor, \( e^{-7a/4} \), is quite close to the factor \( e^{-1.68(1)a} \) found in simulations, suggesting that this is essentially the critical cluster. (Note that this is the smallest cluster with a larger probability of growing than of shrinking.) Applied to the simple cubic lattice, this line of reasoning suggests that the critical cluster consists of seven sites (in the form of a cube with one corner removed) and yields \( \kappa_3 = 11/3 \). (A crude estimate for the value in four dimensions can be obtained by supposing that \( \kappa_4 \approx (\kappa_3^2/\kappa_2) \times \kappa_3 \); this yields \( \kappa_4 \approx 7.7 \), reasonably close to the simulation value, \( \kappa_4 = 8.43(7) \).)

The nucleation scenario outlined above essentially rules out the second transition on any finite-dimensional lattice with short-range interactions. There is always a cluster size such that growth is more likely than shrinkage, and once such a cluster forms, a transition of the global state is likely. From the perspective of equilibrium phase transitions, this is not surprising: the ordered state of any (finite) Ising or Potts model is always subject to a reversal of magnetization at a finite rate. (This rate, however, decays with system size in equilibrium, as a finite cluster is always more likely to shrink than to grow, in the absence of an external field.) In the present cyclic dynamics, by contrast, there is no impediment to growth (no ‘surface tension’, as it were), once a cluster has attained the critical size.

4. Conclusions

We have identified a second phase transition in the cyclic model proposed by Wood et al [12]–[15]. This continuous transition is of a fundamentally different nature than the first, as it corresponds to an infinite-period transition at which macroscopic properties freeze and a discrete rotational (\( C_3 \)) symmetry is spontaneously broken, in the absence of an absorbing state. We find that an appropriate order parameter for the second transition involves the local transition rates. While the phase transition is observed on the complete graph, we argue that it does not exist on finite-dimensional structures with short-range interactions. The question naturally arises of whether such an infinite-period transition exists in systems with long-range interactions. This does indeed appear likely, for interactions that decay sufficiently slowly with distance. In the case of graphs with interactions between many pairs of sites, independent of distance, we expect to observe an infinite-period transition if (in the infinite-size limit) a given site interacts with a finite fraction of the others. We plan to investigate these questions, as well as the behavior on scale-free networks, in future work.

Acknowledgments

VRVA and MC acknowledge financial support from CNPq, FACEPE, CAPES, FAPERJ and special programs PRONEX, PRONEM and INCEMAQ. RD acknowledges financial support from CNPq.

doi:10.1088/1742-5468/2011/09/P09023
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