Synchronization of weakly perturbed Markov chain oscillators

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Rate processes are simple and analytically tractable models for many dynamical systems that switch stochastically between a discrete set of quasistationary states; however, they may also approximate continuous processes by coarse-grained, symbolic dynamics. In contrast to limit-cycle oscillators that are weakly perturbed by noise, in such systems, stochasticity may be strong, and topologies more complicated than a circle can be considered. Here we apply a second-order time-dependent perturbation theory to derive expressions for the mean frequency and phase diffusion constant of discrete-state oscillators coupled or driven through weakly time-dependent transition rates. We also describe a method of global control to optimize the response of the mean frequency in complex transition networks.

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I. INTRODUCTION

The emergence of an oscillating mean field in large ensembles of noisy or nonidentical oscillators is considered the hallmark of synchronization as a collective effect \cite{1–4}. However, the term \textit{synchronization} is used differently in different contexts. It may also refer to anything from statistical correlations to phase synchronization or complete synchronization. For individual oscillators, it is appropriate to define synchronization as the adjustment of frequency due to an interaction \cite{5}. Here it is in this sense that we will study synchronization for weakly driven or coupled Markov chains. If a system with frequency $\omega_0$ is driven at frequency $\omega_1$, two effects may be observed: correlation of the state of the oscillator with the phase of the driving signal and adaptation of mean frequency $\Omega$. For deterministic limit-cycle oscillators, these effects correspond to phase and frequency synchronization, and for coupling stronger than a critical value, phase and frequency locking are observed. These extreme cases are not observed in the presence of noise, although the Kramers theory for weak noise predicts an exponentially small frequency difference within the synchronization regions (Sec. IV C). On the other hand, strong fluctuations, such as those present in noise-induced oscillations and stochastic cycles over a few states or with heterogeneous transition rates, should be considered an integral part of the system. In this case the first-order perturbation theory can predict the correlations induced by weak external perturbation. In addition, with the second-order perturbation theory presented in this paper, it is possible to quantify the change of frequency.

Stochastic processes over discrete and finite sets of states have long been used as conceptual models for stochastic oscillators and proven to capture the essential mechanisms of synchronization. Discrete-state Markov rate models are typically applicable for molecular machines. Enzymes and motor proteins can be considered as molecular machines that undergo operation cycles consuming energy and performing work. Often these operation cycles are well modeled by a finite set of configurations and reaction rates, which describe the speed of transitions between the states. The transitions occur randomly whenever a substrate molecule binds, ATP is converted, or through thermal activation. Molecular machines operating under nonequilibrium conditions are therefore stochastic oscillators that can be approximated by a continuous-time Markov chain model without detailed balance, i.e., with an average flow along the operation cycle [see Fig. 1(a)]. The mean frequency directly equates to the productivity of a molecular machine or the turnover rate of an enzymatic reaction, which can be increased or decreased depending on the situation. This can be achieved by driving the system purposefully at a resonant frequency.

If the frequencies of two oscillators are sufficiently similar, mutual coupling can increase the coherence of the stochastic oscillations. This nontrivial collective effect, where the order is an emergent property of the coupled system, is an important mechanism to reduce noise in biological oscillators on meso- and microscales. It has, for instance, been observed in ensembles of beating heart cells \cite{6}. In Ref. \cite{7} it was shown numerically that by coupling a few noisy gene regulatory circadian oscillators, phase diffusion for each oscillator decreases significantly. Another recent paper \cite{8} concludes that in some circumstances, the circadian protein phosphorylation cycle in cyanobacteria can dramatically enhance the robustness of the coexisting gene regulatory transcription-translation cycle. Both cycles were described by their respective rate processes, i.e., coupled stochastic oscillators of different types.

In this paper we apply the second-order perturbation theory to the master equation of a Markov rate process to derive expressions for its mean frequency $\Omega$. The adjustment of this frequency to frequency $\omega_1$ of a deterministic or stochastic driving signal is clearly identified as a parametric resonance phenomenon. For a simple jump process on a ring with discrete rotational symmetry we can also derive expressions for the phase diffusion constant in the case of external driving or mutual coupling between two stochastic oscillators. By exploiting the special structure of the second-order perturbation terms, we also present an algebraic method to maximize or minimize these terms under linear constraints on the perturbation and quadratic constraints on its power. Thus, it is possible to optimize pump flows in a diffusion process over a directed network in response to the driving frequency. Our work is closely related to the theories of stochastic resonance, ratchets,
and stochastic transport \cite{9,10} but should be viewed in the context of synchronization.

II. PERTURBATION THEORY FOR MARKOV RATE PROCESSES

A Markov rate process over a set of states \( n \in \{1, \ldots, L\} \) is described by its matrix \( W \) of transition rates and a master equation

\[
\dot{P}_{nn}(t|t_0) = \sum_{m=1}^{L} W_{nm}(t) P_{mn}(t|t_0)
\]

for forward-transition probabilities \( P_{mn}(t|t_0) \) to be in state \( n \) at time \( t \) when the system is in state \( n_0 \) at a previous time \( t_0 \). The matrix of transition rates is \( W_{nm} \geq 0 \) for \( m \neq n \) and \( \sum_n W_{nm} = 0 \), which guarantees conservation (\( \sum_n P_{nn} = 1 \)) and nonnegativity \( P_{nm} \geq 0 \) of the probability at all times. Also, because of this property one eigenvalue of \( W(t) \) is always zero. If this eigenvalue is nondegenerate, which is the case when the Markov chain is strongly connected, all other eigenvalues have negative real part, and the transition probability is asymptotically independent of the initial conditions:

\[
P_{nn}(t|t_0 \rightarrow -\infty) = p_n(t).
\]

The perturbation theory is applicable when \( W(t) \) is weakly dependent on time, i.e., for \( W(t) = W^0 + \epsilon V(t) \) and sufficiently small \( \epsilon \). From \( \sum_n W_{nm}(t) = 0 \) for all \( \epsilon \) follows \( \sum_n V_{nm}(t) = 0 \) at all times. The time-dependent perturbation ansatz for the problem

\[
\dot{P}(t|t_0) = [W^0 + \epsilon V(t)]P(t|t_0)
\]

is to expand \( P(t|t_0) \) as a series in powers of \( \epsilon \)

\[
P(t|t_0) = P^{(0)}(t|t_0) + \epsilon P^{(1)}(t|t_0) + \epsilon^2 P^{(2)}(t|t_0) + \cdots.
\]

Inserting this ansatz in Eq. (3) and sorting by powers of \( \epsilon \) the dynamics of any \( P^{(l)}(t|t_0) \) is determined as

\[
P^{(l)}(t|t_0) = W^{(l)}P^{(0)}(t|t_0) + V(t)P^{(l-1)}(t|t_0),
\]

which, given \( V(t) \) and \( P^{(l-1)}(t|t_0) \), is an inhomogeneous, linear ordinary differential equation that can be solved iteratively for each perturbation order.

The mean frequency of a single stochastic oscillator is proportional to a sum of directed and time-averaged flows \( \langle J_{nm}(t) \rangle_t = \langle W_{nm}(t)p_m(t) \rangle_t \). In general, we write

\[
\Omega = 2\pi \sum_{n,m} \Theta_{nm} \langle J_{nm}(t) \rangle_t,
\]

where \( \Theta_{nm} \in [-1,0,1] \) defines a Poincaré section and the direction in which the Poincaré section is crossed for each transition [Fig. 1(b)]. Note that \( \Omega \) depends on the choice of the Poincaré section, which can be arbitrary. Therefore, in general, \( \Omega \) is not equal to one of the relaxation frequencies of the system. In fact, for the two-state Markov model of stochastic resonance \cite{11}, which has the same form as Eq. (1), relaxation is not oscillatory at all, whereas \( \Omega \neq 0 \).

An alternative way to define mean frequency \( \Omega \) and, in addition, phase diffusion constant \( D \) of the stochastic oscillator is to lift the stochastic jump process to a periodic lattice with \( W_{n+L,m+L} = W_{nm} \). It is necessary to decide whether each transition is forward or backward [Fig. 1(b)]. Then the mean frequency and the phase diffusion constant are derived from the asymptotic behavior of the mean and variance of \( n(t) \), for instance, with respect to distribution \( P_{nn}(t+\tau|t) \) as

\[
\Omega = 2\pi \lim_{\tau \rightarrow \infty} \frac{1}{\tau} E \{n(t+\tau) - n_0\},
\]

\[
D = \left(2\frac{\pi}{L}\right)^2 \lim_{\tau \rightarrow \infty} \frac{1}{2\tau} \text{Var} \{n(t+\tau) - n_0\},
\]

where \( E \) and \( \text{Var} \) denote expected value and variance, respectively.

Because of asymptotic independence from the initial conditions, it is possible to perform a suitable average over the initial conditions \( n_0 \) and \( t \). The perturbation theory for this rate process over an infinite, periodic lattice requires the Bloch functions of the infinite, periodic difference operator obtained by the lift of \( W^0 \). We will derive expressions only for our simplest example, which is a one-dimensional ring with discrete rotational symmetry. For this example, the Bloch waves are plain harmonics. Dimensionless quantity \( c = \Omega/D \) is called the Péclet number and describes the coherence of a stochastic oscillator. Its value indicates the number of rotations for an ensemble of stochastic oscillators until coherence is lost.
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Note that, in the linear order of the perturbation, all time averages depend linearly on the time average of $V(t)$. Therefore, without loss of generality, we can absorb this time average into matrix $W^0$ and assume $\langle V(t) \rangle_t = 0$ henceforth. Nonlinear effects are first observed in the time averages of second-order perturbation terms $P^{(2)}$, $Q^{(2)}$, $D^{(2)}$ and $c^{(2)}$ of the transition probabilities and other quantities. We further assume that the perturbation is of the form

$$V(t) = H z(t) + H^* z^*(t), \quad (8)$$

where complex driving signal $z(t)$ has rotational symmetry such that

$$\langle z(t) \rangle_t = \langle z(t + \tau) z(t) \rangle_t = 0 \quad \text{and} \quad \langle z^*(t + \tau) z(t) \rangle_t = e^{i \Lambda t}. \quad (9)$$

Equations (8) and (9) include the case of harmonic driving $z(t) = \exp(-i \omega_1 t)$ where $\Lambda = i \omega_1$ and driving with a continuous or discrete stochastic signal with finite correlation time $|\text{Re}[\Lambda]|^{-1} < \infty$ and relaxation frequency $\text{Im}[\Lambda] = \omega_1$. Complex matrix $H$ assigns relative strength and phase to the perturbation of each transition. Thus we refer to $H$ as the driving protocol. Second-order shift $\Omega^{(2)}$ of the mean frequency is a real-valued function expressed as a sum of products between single entries of $H^*$ and $H$. A Hermitian operator corresponds to this quadratic form over the space of complex matrices, and we define the notation

$$\Omega^{(2)}(H^*, H) = 2 \pi \left[ \nu^0 [v(H^*, H) + v^*(H^*, H)], \right] \quad (10)$$

where the asterisk denotes the complex conjugate. The perturbation expansion of the transition probabilities and related quantities are written in terms of the left and right eigenvectors $v^{(k)}$ and $u^{(k)}$ of $W^0$, respectively, and corresponding eigenvalues $\lambda_k$, i.e., $W^0 v^{(k)} = \lambda_k v^{(k)}$ and $v^{(k)} W^0 = \lambda_k^* v^{(k)}$, where $\dagger$ denotes the complex conjugate transpose. We assume normalization $v^{(k)\dagger} u^{(k)\dagger} = \delta_{kk}$, completeness $\sum_k u^{(k)\dagger} = \mathbb{I}_{L \times C}$, and conventions $\lambda_0 = 0, p^{(0)} = u^{(0)}$, and $v^{(0)} = 1$, where $\mathbb{I}_{L \times L}$ is the identity matrix of size $L$ and $1 = (1, \ldots, 1)^T$. In Appendix A, we derive

$$v(A^*, B) = - \sum_{nm} \Theta_{nm} \sum_{k \neq 0} u^{(k)} m \left[ A_{nm} v^{(k)\dagger} B_{00}^{(0)} \right] - W^0_{nm} \sum_{k \neq 0} v^{(k)\dagger} A^{*} u^{(k)} v^{(k)\dagger} B_{00}^{(0)} \left( \lambda_{k'} + \lambda_{k} \right) \quad (11)$$

for the bilinear form $v(A^*, B)$ in Eq. (10). Equations (10) and (11) capture the essence of frequency adjustment toward the frequency of the driving signal (Sec. IV) and also demonstrate the resonance nature of synchronization. Denominators $(\lambda_k + \Lambda)$ in the sums in Eq. (11) are minimized when $\text{Im}[\Lambda] = -\text{Im}[\lambda_k]$, i.e., when the driving frequency matches the relaxation frequency of some mode $k$.

III. OPTIMIZATION

This section describes a corollary technique to determine the complex driving protocol $H$ that optimizes the second-order perturbation response of the mean frequency. In Sec. IV D, we apply this technique to a Markov rate jump process on a random network.

Being able to express the time-averaged second-order responses as a Hermitian form in terms of eigenvalues and eigenfunctions of the unperturbed transition matrix, we can formulate an optimization problem that can be solved using linear algebra. Consider a quadratic form

$$f(x) = x^H F x \quad (12)$$

with $x \in \mathbb{C}^N$ and a Hermitian matrix or operator $F^\dagger = F$. Linear constraints may be given as

$$P^H x = 0, \quad (13)$$

where the column vectors of $P \in \mathbb{C}^{N \times l}$ with $l < N$ can, without loss of generality, be assumed to be orthonormal, i.e., $P^H P = I_{l \times l}$. The nullspace of $P$ is spanned by another set of orthonormal vectors given by the column vectors of matrix $Q \in \mathbb{C}^{N \times (N-l)}$ such that $Q^H I = I_{(N-l) \times (N-l)}$ and $P^H Q = 0$. The purpose of optimization is to maximize or minimize $f(x)$ subjected to the linear constraints in Eq. (13) and the constraint on a cost function given by a positive definite matrix $S$ as

$$x^H S x = 1. \quad (14)$$

This problem is related to the variational problem in quantum mechanics to minimize or maximize

$$\Phi^H M \Phi \quad \text{subject to} \quad ||\Phi||^2 = 1 \quad (15)$$

for some Hermitian operator $M$. It can be easily shown that the solution $\Phi_{\text{opt}}$ is the normalized eigenfunction of $M$ with the largest or smallest eigenvalue. To put our optimization problem in this form, we need to project into the nullspace of $P$ and apply a similarity transformation that changes the constraint in Eq. (14) on the cost function into the normalization condition on $\Phi$. This is achieved by the following rules:

$$\Sigma = Q^H S Q, \quad M = \Sigma^{-1/2} Q F Q \Sigma^{-1/2}, \quad x = \Sigma^{-1/2} \Phi. \quad (16)$$

Note that with positive definite $S$, projected matrix $\Sigma$ is also positive definite, so that $\Sigma^{-1/2}$ is well defined and Hermitian.

When applied to the problem of finding an optimal driving protocol, vector $x \in \mathbb{C}^{L^2}$ represents complex matrices $H \in \mathbb{C}^{L \times L}$. The linear constraints can be as simple as requiring $H_{nm} = 0$ for impossible transitions from state $m_0$ to $n_0$ or transitions that cannot be perturbed. More subtle constraints can be assigned on the phase relations. For example, setting $H_{n_0,m_0} = e^{i \pi} H_{n_0,m_0}$ describes periodic switching in preference between two possible target states that are reached from state $m_0$, which is a very simple model for intersections with a traffic light. Optimizing the frequency and phase relations between traffic lights in a road network is a classic problem in transport theory.

The quadratic form $f(x) = x^H F x$ that is to be optimized can, for instance, be second-order frequency shift $\Omega^{(2)}(H^*, H)$. By using the explicit expressions given in Eqs. (10) and (11), we can also optimize $\Omega^{(2)}_{nm}$ at a fixed driving frequency $\omega_1$. A driving protocol that maximizes the response to changes in the driving frequency can be said to have good synchronizing properties.
rates for a forward jump process \( (γ \text{ and perturbation depending on time and state.}) \) (b) Transition is divided into a time-independent forward bias, diffusion part, and perturbation depending on time and state. (b) Transition rates for a forward jump process \((γ = 0)\) on a two-dimensional periodic lattice with attractive coupling in two directions. The speed of the transitions is indicated by different emphasis of the arrows.

**IV. EXAMPLES**

**A. Jump process on a ring**

In this section, we present the results for the mean frequency and the phase diffusion constant for a sequential change of states with discrete rotational or translational symmetry of transition rates \( W^0 \). We consider \( L \) states that are visited in sequence with forward transition rates \( W^0_{n+1,n} = L(1 + γ) \) and backward transition rates \( W^0_{n-1,n} = Lγ \) [Fig. 2(a)]. We directly consider the lift of the finite-state oscillator to the periodic lattice with translational symmetry \( W^0_{n+1,m+1} = W^0_{nm} \). The time scale is chosen such that the mean frequency of the unperturbed oscillator is \( ω_0 = 2π \). The transition rates can be split as \( W^0 = W^B + W^D \) into forward jump process \( W^B \) and unbiased diffusion part \( W^D \) [see Fig. 2(a)]. As discussed in Sec. IV C, these two rate processes give rise to the deterministic part and the diffusion of the stochastic oscillator in the continuum limit.

Suppose that the perturbations are given only to the forward bias in the form of a traveling wave

\[
V_{nm} = W^B_{nm} A(t) \cos[k_0m - φ(t)],
\]

where phase \( φ(t) \) and amplitude \( A(t) \) may be deterministic or stochastic. In terms of Eq. (8), we have

\[
H_{nm} = W^B_{nm} \frac{1}{L} e^{i k_0 m}, \quad z(t) = A(t) e^{-i φ(t)},
\]

and \( z(t) \) must satisfy Eq. (9). Because of the translational symmetry of the system the eigenfunctions of \( W^B \) and \( W^D \) are harmonics given by \( u^{(k)} = 2π u^{(k)} \) = \( \exp(i k n) \) with wave numbers \( k \in (-π, π] \). The eigenvalues are given by

\[
W^B_{nm} u^{(k)} = λ^B_{nm} u^{(k)} = L(e^{-i k} - 1) u^{(k)},
\]

\[
W^D_{nm} u^{(k)} = λ^D_{nm} u^{(k)} = Lγ(e^{i k} + e^{-i k} - 2) u^{(k)},
\]

\[
W^0_{nm} u^{(k)} = λ^0_{nm} u^{(k)} = (λ^B_{nm} + λ^D_{nm}) u^{(k)}.
\]

Furthermore, the action of the perturbation on an eigenmode creates two eigenmodes of wave numbers \( k \pm k_0 \), i.e.,

\[
H u^{(k)} = \frac{1}{2} \left( λ^B_{nm} u^{(k + k_0)} + λ^B_{nm} u^{(k - k_0)} \right), \quad H^* u^{(k)} = \frac{1}{2} \left( λ^B_{nm} u^{(k + k_0)} + λ^B_{nm} u^{(k - k_0)} \right).
\]

In Appendix B, by using the perturbation expansion of the characteristic function for the random process, we determine mean frequency \( Ω \), phase diffusion constant \( D \), and Péclet number \( c = Ω/D \) up to the second order in perturbation strength \( ε \) as

\[
ω_0 = 2π, \quad D_0 = \frac{2π^2}{L} (1 + 2γ), \quad c_0 = ω_0/D_0,
\]

\[
\frac{Ω}{ω_0} = 1 - ε^2 \frac{1}{4} \left( \frac{c}{λ^{B}_{0}} + \frac{c}{λ^{D}_{0}} \right)
\]

\[
D = 1 + ε^2 \frac{1}{4(1 + 2γ)} \left( \frac{1 - γ}{λ^{B}_{0}} + \lambda^{D}_{0} \right)^2 + \frac{3λ^{B}_{0} + 2λ^{D}_{0}}{λ^{B}_{0} + λ^{D}_{0}} + c.c.
\]

\[
\frac{c}{c_0} = 1 + ε^2 \left( \frac{Ω^{(2)}}{ω_0} - \frac{D^{(2)}}{D_0} \right)
\]

\[
= 1 + ε^2 \frac{1}{2(1 + 2γ)} \left( \frac{1 - γ}{λ^{B}_{0}} + \frac{L}{λ^{D}_{0}} + \frac{3λ^{B}_{0} + 2λ^{D}_{0}}{λ^{B}_{0} + λ^{D}_{0}} + c.c. \right)
\]

where c.c. denotes the complex conjugated terms. The six parameters of this model are length of the ring \( L \) (or alternatively the time scale of the jump process on the infinite lattice), diffusion parameter \( γ \), wave number \( k_0 \) of the driving traveling wave, dissipation rate \( -\text{Re}[Λ] \) and frequency \( \text{Im}[Λ] \) of the driving signal, and the perturbation strength \( ε \). If we drive deterministically with \( z(t) = \exp(-iω_0 t) \) the exponent \( Λ = iω_1 \) is purely imaginary. The shifts of the mean frequency and the Péclet number for this case are shown in Figs. 3(a) and 3(b), respectively. If the stochastic oscillator is coupled to the mean field from a finite ensemble or a single other stochastic oscillator, \( Λ \) will have a negative real part that adds to the negative real part of \( λ^{B}_{0} \). Thus the resonance effect decreases because the absolute values of the denominators in Eqs. (22)–(24) increase. For instance, with \( z(t) = \exp[-ik_1 n_1(t)] \), depending on another unperturbed oscillator with stochastic jump process \( n_1(t) \), the exponent of the autocorrelation function is simply \( Λ = -λ^{B}_{0} \) [see Eq. (B27) in Appendix B]. Driving with another stochastic oscillator of identical length, i.e., with similar stochasticity but variable mean frequency, the frequency shift is weaker, and we observe no enhancement in the Péclet number [see Figs. 3(c) and 3(d)].
simulations were performed using a generalized Gillespie algorithm.

So far, in the time-dependent perturbation theory, averaging Monte Carlo simulations \([Figs. 3(e) and 3(f)]\) of the coupled rate processes performed with the usual Gillespie algorithm (see Appendix C). We have verified our results against \((\omega_0/\omega_1)^2 L\) solutions over the initial conditions \([\omega_0=0.6, \omega_1=0.2] \Omega = \pi/L, -2\pi/L \) and \(\beta = \pi/2\). Numerical simulations were performed using a generalized Gillespie algorithm for time-dependent transition rates. Averages were taken over 1000 runs for 1000 time units each with the parameters \(\gamma = 0.1, \varepsilon = 0.5\).

The inset in subfigure (a) applies to all subfigures.

B. Mutually coupled oscillators

Unlike an externally driven stochastic oscillator, system \(n(t) = (n_0(t),n_1(t))\) of two mutually coupled stochastic oscillators is autonomous. As we pointed out, it is permitted to take an average over the initial conditions in Eqs. (7). So far, in the time-dependent perturbation theory, averaging over time has considerably simplified the equations. Now, the perturbation does not depend on time, i.e., \(z(t) = 1\) in Eq. (8).

Instead, we average \(P_{n_0+n_1} (\tau | 0)\) over the initial conditions \(n_0\) (see Appendix C). We have verified our results against Monte Carlo simulations [Figs. 3(e) and 3(f)] of the coupled rate processes performed with the usual Gillespie algorithm for time-independent transition rates.

We again consider the lift of the combined system \(n = (n_0,n_1) = n_0 e^0 + n_1 e^1\) to an infinite periodic lattice [see Fig. 2(b)]. The unperturbed transition rates may be given by isotropic diffusion part \(W^{diff}\) and forward bias \(W^0\) with different values in the directions \(e^0\) and \(e^1\):

\[
\begin{align*}
W^{0} &= W^{diff}, \\
W_{n_0+n_1,n_0}^{diff} &= W_{n_0+n_1,n_1}^{diff} = L \gamma, \\
W_{n_0+n_1,n_0}^{diff} &= L, W_{n_0+n_1,n_1}^{diff} = \frac{\omega_1}{2\pi} L.
\end{align*}
\]

Coupling shall be given by matrix

\[
\begin{align*}
V_{n_0+n_1,n_0} &= L \cos(\mathbf{k}^0 \mathbf{n} + \beta), \\
V_{n_0+n_1,n_1} &= \frac{\omega_1}{2\pi} L \cos(\mathbf{k}^0 \mathbf{n} - \beta).
\end{align*}
\]

Term \(\mathbf{k}^0 \mathbf{n}\) denotes the inner product of \(\mathbf{n}\) with wave vector \(\mathbf{k}^0\). In contrast to external driving, here phase shift \(\beta\) does have physical effects. For \(\mathbf{k}^0 = (2\pi L^{-1}, -2\pi L^{-1})\) and \(\beta = \pi/2\) coupling is attractive, whereas for \(\beta = 0\) cosine coupling inhibits synchronization. The former case is presented in Figs. 3(e) and 3(f).

Due to symmetry, the eigenfunctions of the unperturbed system are harmonics \(\nu_n^{\mathbf{k}} = (2\pi)^2 u_{\mathbf{k}}^{\mathbf{n}} = e^{i\mathbf{k} \mathbf{n}}\). We define

\[
\begin{align*}
\lambda_{\mathbf{k}}^{diff} &= \lambda_{\mathbf{k}}^{\mathbf{0}} + \gamma_{\mathbf{k}}^{diff}, \\
\lambda_{\mathbf{k}}^{\mathbf{0}}(\beta) &= \lambda_{\mathbf{k}}^{\mathbf{0}} + \frac{\omega_1}{2\pi} \lambda_{\mathbf{k}}^{\mathbf{0}} e^{-i\beta}, \\
\lambda_{\mathbf{k}} &= \lambda_{\mathbf{k}}^{\mathbf{0}}(0) + \lambda_{\mathbf{k}}^{diff},
\end{align*}
\]

where \(\gamma_{\mathbf{k}}^{diff}\) and \(\lambda_{\mathbf{k}}^{diff}\) are given by Eq. (19) and \(\mathbf{k} = (k_0,k_1)\). In Appendix C we determine the frequency, phase diffusion constant, and Péclet number of the first oscillator to the second-order in perturbation strength \(\varepsilon\) as

\[
\odder = \frac{\Omega}{\omega_0} = \left\{ 1 - \varepsilon^2 \left[ \frac{1}{4} \left( \frac{2}{\lambda_{\mathbf{k}}^{\mathbf{0}}(\beta)} + \frac{\lambda_{\mathbf{k}}^{\mathbf{0}}}{\lambda_{\mathbf{k}}^{diff}} \right) \right] \right\}.
\]

\[
D = \frac{D_0}{\omega_0} = 1 + \varepsilon^2 \left[ \frac{1}{4(1 + 2\gamma)} \left[ \frac{2 \lambda_{\mathbf{k}}^{\mathbf{0}}(\beta)}{\lambda_{\mathbf{k}}^{\mathbf{0}} + \gamma_{\mathbf{k}}^{\mathbf{0}} + i 2\gamma \text{Im}[\lambda_{\mathbf{k}}^{\mathbf{0}}]} \frac{\lambda_{\mathbf{k}}^{\mathbf{0}}}{\lambda_{\mathbf{k}}^{diff}} \right] - \frac{2(\lambda_{\mathbf{k}}^{\mathbf{0}} + L)}{\lambda_{\mathbf{k}}^{\mathbf{0}}} + \text{c.c.} \right] + \text{c.c.}.
\]

In contrast to the driven system, frequency \(\omega_1\) of the second oscillator enters both the denominator and numerator of the expressions via \(\lambda_{\mathbf{k}}^{\mathbf{0}}\) and \(\lambda_{\mathbf{k}}^{\mathbf{0}}(\beta)\), respectively. As a result, the second-order perturbation terms do not vanish in the limit \(\omega_1 \to \infty\), since the phases of the fast and the slow oscillator remain correlated. Another remarkable difference is the possibility of an increase in coherence for two mutually coupled, identical, or weakly nonidentical stochastic oscillators [Fig. 3(f)].
While the time-dependent perturbation theory can be used to approximate the frequency and the phase diffusion constant under driving by another stochastic oscillator, the transition rates in the combined system \( n_0(t), n_1(t) \) are not time dependent, irrespective of mutual or unidirectional coupling. The mean frequency in the combined autonomous system can, in principle, be obtained nonperturbatively from the stationary probability distribution, i.e., the explicit zero eigenvector of the combined time-independent matrix of transition rates. In Ref. [12] this was performed for a master-slave pair of two-state oscillators, and frequency locking was observed for strong coupling. The phase diffusion constant cannot be obtained in such a simple way. In Ref. [13] a method has been recently found to determine the phase diffusion constant in a periodically driven system from the cyclostationary solution of an extended, periodically driven, linear ordinary differential equation. In that case, the lift to the periodic lattice and the generating function approach is not necessary. It would be worth investigating whether this method can be generalized to a larger number of states, nonsequential transition networks, or stochastic driving.

C. Continuum limit

The purpose of this section is to illustrate that our theory is accurate in the limit of weak coupling and strong noise and complements the Kramers rate theory in the other asymptotic regime of strong coupling and weak noise.

In the limit \( L \to \infty \), the biased jump process on a ring network with harmonic driving corresponds to a continuous, weakly perturbed limit-cycle oscillator in the Kuramoto phase approximation

\[
\dot{\varphi} = 2\pi [1 + \varepsilon \cos(\varphi - \omega_0 t)] + \sqrt{2D_0} \xi(t)
\]

with delta correlated white noise \( \langle \xi(t)\xi(t') \rangle = \delta(t-t') \). To obtain a finite phase diffusion constant, \( \gamma/L \) must be kept constant such that, in the limit \( L \to \infty \),

\[
\frac{4\pi^2y}{L} \to D_0.
\]

The second-order perturbation terms for the frequency shift, the phase diffusion constant, and the Péclet number in the continuum limit are given in Appendix B. However, exact nonperturbative expressions for these quantities can be found. Substituting \( \varphi = \varphi_0 + \varphi \) and \( \Delta = (\omega_1 - 2\pi) \) in Eq. (31), we obtain the stochastic Adler equation

\[
\dot{\varphi} = -\Delta + 2\pi \varepsilon \cos \varphi + \sqrt{2D_0} \xi(t),
\]

for which the mean frequency and the phase diffusion constant are known in terms of special integrals \([14, 15]\). Here we will use the continuum limit to compare the different regimes in which our perturbation theory and the Kramers theory are valid. Noting that we use a time scale such that \( \omega_0 = 2\pi \), we introduce dimensionless parameters \( x = \Delta/D_0 \) and \( y = 2\pi \varepsilon /D_0 \). The continuum limit of the second-order perturbation frequency shift in Eq. (22) is given in Appendix B [Eq. (B31b)]. In terms of \( x \) and \( y \), Eq. (B31b) can be rewritten as

\[
\frac{\Omega - \omega_0}{D_0} \approx \frac{y^2}{2} \frac{x}{1 + x^2}.
\]

A value of this derivative larger than unity is unphysical, because then the frequency response is stronger than the change in the driving frequency. Therefore, we say \( y = \sqrt{2} \) is an upper bound for the region in which our perturbation theory is valid. On the other hand, the Kramers approximation gives the rate of phase slips in the Arnold tongue region for small noise strength \( D_0 \) or strong coupling. Equation (33) can be written as a gradient system

\[
\dot{\varphi} = -U'(\varphi) + \sqrt{2D_0} \xi(t),
\]

\[
U(\varphi) = (\Delta \cdot \varphi - 2\pi \varepsilon \sin \varphi).
\]

The tilted sine potential is shown in Fig. 4. It has a local minimum at \( \varphi_0 \) and the Kramers theory requires (\( U_{\text{max}} - U_{\text{min}} \)) / \( D_0 \) \( \gg 1 \). We can parametrize the condition (\( U_{\text{max}} - U_{\text{min}} \)) / \( D_0 \) = 1 by \( 0 \leq \varphi_0 \leq \pi/2 \) and obtain

\[
y = \frac{1}{2(\sin \varphi_0 - \varphi_0 \cos \varphi_0)}, \quad x = \pm \sqrt{\frac{1}{2} \sin \varphi_0}.
\]

The Kramers rate \( r_k \) over a potential barrier is given \([16]\) by

\[
r_k = \frac{1}{2\pi} \sqrt{\frac{U''_{\text{min}}}{U''_{\text{max}}}} \frac{1}{D_0} \left( e^{U_{\text{max}} - U_{\text{min}}} / D_0 \right).
\]

From \( \varphi = (\dot{\varphi})_r = \omega_1 + (\dot{\varphi})_l \) and by subtracting the backward jump rates from the forward jump rates, we obtain

\[
\Omega - \omega_0 \approx x - y \sin(\varphi_0) e^{2(\sin \varphi_0 - \varphi_0 \sin \varphi_0)(1 - e^{-2\pi x})}.
\]

Figure 5 compares the regions in the \((x, y)\) parameter space where the Kramers approximation holds with that where our second-order perturbation theory is valid at the resonance point. Small noise or large coupling strengths will result in strongly nonlinear behavior near the resonant frequency and render the perturbation theory invalid. In contrast, the regime
of the Kramers approximation can always be reached by increasing the coupling strength, or if $|x/y| < 1$, by reducing the noise strength. Figure 5(b) shows the rescaled frequency shift at $x = \Delta / D_0 = 1$ as a function of $y = 2\pi\varepsilon / D_0$. At low coupling strengths $y < \sqrt{2}$, the second-order perturbation theory approximates the quadratic behavior of the frequency shift, whereas at high coupling strengths, the Kramers rate theory describes the exponential deviation from frequency locking. Therefore, it is appropriate to consider the adaptation of frequency in our theory as a weak form of synchronization.

D. Ring with random shortcuts

While the jump process on a ring is a discrete approximation for a limit-cycle oscillator with finite phase diffusion, this one-dimensional geometry may not reflect the topology of a more complicated oscillatory or transport process. A way to incorporate a more complex topology is to use a network of states that are not visited in sequence. Here we study the effect of adding $E_{rc}$ random shortcuts to a unidirectional ring network of size $L$ [Fig. 6(a)]. The resulting Markov chain is strongly connected ensuring that the eigenvalue $\lambda_0 = 0$ is nondegenerate.

The strength of the resonance depends on coherence ratio $r_k = \text{Im}[\lambda_k] / \text{Re}[\lambda_k]$ of resonant eigenmode $k$. The corresponding eigenvector $u^{(k)}$ and its complex conjugate quantify collective oscillations and are associated with nonzero stationary probability fluxes in the system. Figure 6(a) shows a typical network for $L = 30$, $E_{rc} = 15$, i.e., shortcut density $\sigma = 0.5$, where each node $n$ is located on a circle at phase $\vartheta_n = 2\pi n / L$. The same network is shown in Fig. 6(b) with $\vartheta_n = \arg(u^{(k)}_n)$, where $u^{(k)}$ is the right eigenvector of $W_0$ for the eigenvalue with the maximal ratio $r_k$. Forward transitions are highlighted blue (solid) and backward transitions red (thin, dashed). With respect to this arrangement, the flow is strongly biased in the forward direction.

We investigated the dependence of the expected maximum value $r_{max}$ of the coherence ratio on shortcut density $\sigma = E_{rc} / L$. The eigenvalues of the transition matrices are localized in a region of the complex plane with a negative real part and are symmetric with respect to the line of real numbers [Fig. 6(c)]. The slope of a line originating at zero and tangent to that region scales with some power of $\sigma$ in both limits $\sigma \to 0$ and $\sigma \to \infty$ [Fig. 6(d)]. The decrease in $r_{max}$ with increasing $\sigma$ is the result of stronger mixing because of the random shortcuts. In the region of the topological crossover at $\sigma = 1$ [17] we observe that the coherence ratio is approximately equal to unity [see Figs. 6(c) and 6(d)].

When the transition rates are perturbed harmonically with frequency $\omega_1$ and driving protocol $H$ that encodes the relative
amplitudes and phases between the perturbations, the mean frequency will change by a value given approximately by Eqs. (10) and (11) in Sec. II. Using the technique described in Sec. III we optimized the driving protocol for the example shown in Fig. 6(b) such that, for a fixed driving frequency $\omega_1$ responsiveness $\partial\omega_2/\partial\omega_1$ of the frequency shift is maximized. We assumed that the perturbation cost is the same for every transition. We chose $\omega_1$ to be one of the apparent resonant frequencies in Fig. 7(a). Synchronization, i.e., the adaptation of the mean frequency to the driving frequency, both positive and negative, is observed in Fig. 7(b) when these driving protocols are kept constant and $\omega_1$ is changed.

V. DISCUSSION

We have presented second-order perturbation analysis for the frequency of periodically or stochastically driven, time-continuous Markov chain models. In the case of a biased jump process on a ring, we also determined the second-order perturbation terms for the phase diffusion constant. This simplest discrete-state model of a stochastic oscillator can adapt its mean frequency toward the frequency of a deterministic or stochastic driving signal. The mechanism for this weak form of synchronization is a parametric resonance of the driving signal with an oscillatory relaxation mode of the stochastic process. We also showed that, when two identical stochastic oscillators are attractedly coupled, they can increase their Péclet number, a measure for the coherence of oscillations, above the level of the uncoupled system. Furthermore, the explicit expressions for the frequency shift could be used to optimize the perturbations in a complex transition network.

Phase response can be viewed as a perturbation theory for the mean frequency of an oscillator. The techniques presented in this paper could therefore be useful in developing effective phase models for noise-induced or strongly noise-perturbed oscillations [15]. The explicit perturbation expressions for the frequency shift of a periodically driven Markov rate process may be used to estimate the unperturbed transition rates. In the case of enzymatic reactions, it may thus be possible to examine the reaction kinetics by measuring the response of the turnover rate to changes in the frequency and wave form of the driving signal.

Of course, the theory presented in this paper is also applicable to other nonequilibrium systems that are reasonably well described by stochastic cycles through the states of a Markov chain as well as stochastic transport or growth processes. In Ref. [18], periodic driving of the transition rates in the asymmetric exclusion process, which models transport through a cell membrane, was studied perturbatively. The resonance of the mean flow with the driving frequency in the second order in the perturbation strength was found, which agrees with our general results.

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APPENDIX A

In this section we will derive general expressions for the time-dependent transition probabilities of a driven Markov process up to the second order in perturbation strength. Transition probabilities $P_{nm}(t,t_0)$ obey the following equation:

$$\dot{P}(t|t_0) = [W^0 + \varepsilon V(t)]P(t|t_0)$$  \hspace{1cm} (A1)

with

$$V(t) = H z(t) + H^* z^*(t).$$  \hspace{1cm} (A2)

We assume that complex driving signal $z(t)$ is stationary and symmetric such that

$$\langle z(t + \tau) z(t) \rangle = \langle z(t) \rangle = 0 \quad \text{and} \quad \langle z^*(t + \tau) z(t) \rangle = e^{i\lambda \tau} \quad \text{for} \quad \tau > 0.$$  \hspace{1cm} (A3)

The perturbation expansion of $P(t|t_0)$ will be given in terms of eigenvectors and eigenvalues of unperturbed transition rates $W^0$. We denote the left and right eigenvectors as $v^{(k)}$ and $u^{(k)}$, respectively, and the corresponding eigenvalues as $\lambda_k$. The completeness and orthogonality relations for the eigenvectors are

$$\sum_k u^{(k)} v^{(k)^*} = \delta_{kk}$$

and

$$\sum_k v^{(k)*} u^{(k)} = \delta_{kk},$$

respectively. In the main text we also set $v^{(0)} = 1$ for $1 \leq n \leq L$. If $W^0$ is an infinite periodic difference operator, the eigenvalues are distributed continuously in a finite number of bands. Then the completeness relation is expressed as a sum over different bands and for each band an integral over wave numbers $(-\pi, \pi)$. The orthogonality relation is given by the product of a Kronecker delta for the band number and a Dirac delta function for the wave numbers within the bands. The nondegenerate zero eigenvalue corresponding to the unique stationary solution of the unperturbed problem is denoted as $\lambda_0 = 0$. By inserting the perturbation ansatz

$$P(t|t_0) = P^{(0)}(t|t_0) + \sum_l \varepsilon^l P^{(l)}(t|t_0)$$  \hspace{1cm} (A4)
In Eq. (A1) and sorting by powers of $\epsilon$, the dynamics of each perturbation term is given by a linear, inhomogeneous, ordinary differential equation:

$$\dot{P}^{(l)}(t|t_0) = W^{(l)} P^{(l)}(t|t_0) + V(t) P^{(l-1)}(t|t_0). \quad (A5)$$

In terms of Fourier modes,

$$\pi^{(l)}_{kk}(t|t_0) = v^{(k)} P^{(l)}(t|t_0) u^{(k)}, \quad q^+_{kk} = v^{(k)} H u^{(k)}, \quad q^-_{kk} = v^{(k)} H^* u^{(k)}.$$  \quad (A6)

and using the completeness relation, Eq. (A5) becomes

$$\dot{\pi}^{(l)}_{kk}(t|t_0) = \lambda_k \pi^{(l)}_{kk} + \sum_{k'} \left[ q^+_{kk'} \pi^{(l-1)}_{k'}(t|t_0) - q^-_{kk'} \pi^{(l-1)}_{k}(t|t_0) \right]. \quad (A7)$$

Note that because $v^{(0)} | H = v^{(0)} | H^* = 0$, coefficients $q_{kk'} \pm \lambda_k$, vanish for $k = 0$. We will explicitly exempt these modes from the sum or integral in Eq. (A7). From the initial condition $P_{nn}(t|t_0) = P_{nn}^{(0)}(t|t_0) = \delta_{nn}$ follows $\pi^{(l)}_{kk}(t|t_0) = \delta_{l0} \delta_{kk}$, and therefore

$$\pi^{(0)}_{kk}(t|t_0) = e^{\lambda_k t} \delta_{kk}.$$  \quad (A8)

All higher-order perturbation terms are given by convolutions of the inhomogeneous part with an exponential kernel. Defining

$$L^{0}_{kk}[f](t) = \int_0^t f(t + t') e^{(\lambda_k - \lambda_k^*) t'} dt', \quad (A9)$$

the general solution to Eq. (A7) for $l > 0$ is

$$\pi^{(l)}_{kk}(t|t_0) = e^{\lambda_k t} \sum_{k'} q^+_{kk'} L^{0}_{k'} \left[ \pi^{(l-1)}_{kk'}(t) \right] + e^{\lambda_k t} \sum_{k'} q^-_{kk'} L^{0}_{k'} \left[ \pi^{(l-1)}_{kk'}(t) \right], \quad (A10)$$

The first- and second-order perturbation terms are thus, respectively,

$$\pi^{(1)}_{kk}(t|t_0) = e^{\lambda_k t} q^+_{kk'} L^{0}_{k'} \left[ \pi^{(0)}_{kk'}(t) \right] + e^{\lambda_k t} q^-_{kk'} L^{0}_{k'} \left[ \pi^{(0)}_{kk'}(t) \right], \quad (A11)$$

and

$$\pi^{(2)}_{kk}(t|t_0) = e^{\lambda_k t} \sum_{k^0} q^+_{kk'k^0} \pi^{(1)}_{k^0k'} \left[ \pi^{(0)}_{kk'}(t) \right] + e^{\lambda_k t} \sum_{k^0} q^-_{kk'k^0} \pi^{(1)}_{k^0k'} \left[ \pi^{(0)}_{kk'}(t) \right] + O(\epsilon^2). \quad (A12)$$

Here $O(\epsilon^2)$ denotes the products of the form $z(t + t') z(t + t''')$ and $z'(t + t') z''(t + t''')$, which will vanish when averaged over $t$. The remaining nonvanishing time averages are expressed by incomplete Laplace transformations of complex autocorrelation function $c(\tau) = \langle z(t + t') z(t) \rangle$, or its complex conjugate. We observe

$$\begin{align*}
\langle z^n(t + t') L^{0}_{kk}[z](t) \rangle _0 & = \int_0^t \langle z^n(t + t') z(t + t') \rangle e^{-(\lambda_k - \lambda_k^*) t'} dt' \\
& = \int_0^t c(t - t') e^{-(\lambda_k - \lambda_k^*) t'} dt' \\
& = \int_0^t c(t') e^{(\lambda_k - \lambda_k^*) (t - t')} dt' \\
& = e^{(\lambda_k - \lambda_k^*) t} \mathcal{L}^0_{kk}[c](t) \\
\end{align*}$$

(A13)

and

$$\begin{align*}
\langle L^n_{kk} \left[ z^* L^{0}_{kk}[z'] \right](t) \rangle _0 & = \int_0^t \langle L^n_{kk} \left[ z^* L^{0}_{kk}[z'](t') \right] e^{-(\lambda_k - \lambda_k^*) t'} dt' \\
& = \mathcal{L}^0_{kk}[L^n_{kk} \left[ L^{0}_{kk}[c](t') \right]]. \\
\end{align*}$$

(A14)

For $c(\tau) = \exp(\Lambda \tau)$, the incomplete Laplace transformations are solved explicitly. For shorthand notation, we define $g^\Lambda_{kk} = \lambda_k - \lambda_k - \Lambda$. Then

$$e^{\lambda_k t} L^{0}_{kk}[c](t) = \frac{e^{\lambda_k t}}{g^\Lambda_{kk}} \left( e^{\lambda_k t} - 1 \right)$$

(A16)

and

$$e^{\lambda_k t} \mathcal{L}^0_{kk}[L^n_{kk} \left[ L^{0}_{kk}[c](t) \right]](t) = \frac{e^{\lambda_k t}}{g^\Lambda_{kk}} \left( \frac{e^{\lambda_k t} - 1}{g^\Lambda_{kk}} - \frac{e^{\lambda_k t} - 1}{\lambda_k - \lambda_k} \right).$$

(A17)

For $k' = k$, this equation takes the form

$$e^{\lambda_k t} \mathcal{L}^0_{kk}[L^n_{kk} \left[ c \right]](t) = \frac{e^{\lambda_k t}}{g^\Lambda_{kk}} \left( \frac{e^{\lambda_k t} - 1}{g^\Lambda_{kk}} - \frac{e^{\lambda_k t} - 1}{\lambda_k - \lambda_k} \right).$$

(A18)

For $k \neq 0$ and $k'' \neq 0$, the dependence on the initial conditions is removed by taking limit $\tau \to \infty$ in Eqs. (A16) and (A17):

$$\lim_{\tau \to \infty} e^{\lambda_k t} \mathcal{L}^0_{kk}[L^n_{kk} \left[ c \right]](t) = -\delta_{k0} \frac{1}{g^\Lambda_{kk}},$$

(A19)

$$\lim_{\tau \to \infty} e^{\lambda_k t} \mathcal{L}^0_{kk}[L^n_{kk} \left[ c \right]](t) = \delta_{k0} \frac{1}{\lambda_k - g^\Lambda_{kk}}.$$
Mean frequency \( \omega \) is proportional to the time-averaged flow over this Poincaré section times 2\( \pi \),
\[
\Omega = 2 \pi \sum_{n,m} \Theta_{nm} \left( J_{nm}(t) \right)_t
\]
and
\[
J_{nm}(t) = W_{nm}^{(0)} P_m^{(0)} + \varepsilon \left[ W_{nm}^{(1)} P_m^{(1)} + V_{nm}(t) P_m^{(0)} \right] + \varepsilon^2 \left[ W_{nm}^{(2)} P_m^{(2)} + V_{nm}(t) P_m^{(1)} \right].
\]
The time averages in the linear order of \( \varepsilon \) are zero because of Eq. (A3). We are now able to express the mean frequency in the second order of \( \varepsilon^2 \) by using Eqs. (A11), (A12), (A16), (A17), and (A19). We have
\[
\left( V_{nm}(t) P_m^{(1)}(t) \right)_t = \sum_{k \neq 0} u_n^{(k)} \left( H_{nm}(\varepsilon \pi k_0) v_0^{(0)*} + h_{nm}^* (\varepsilon \pi k_0) v_0^{(0)*} \right) = - \sum_{k \neq 0} u_n^{(k)} \left( H_{nm}^* \frac{q_{k0}}{g_{k0}} + h_{nm}^* \frac{q_{k0}}{g_{k0}} \right) v_0^{(0)*}.
\]
As well as
\[
W_{nm}^0 \left( P_m^{(2)}(t) \right)_t = \sum_{k \neq 0} u_m^{(k)} W_{nm}^0 \left( \pi k_0 \right) v_0^{(0)*} = \sum_{k' \neq k} u_m^{(k)} W_{nm}^0 \left( \frac{q_{k0}^* q_{k0}}{\lambda_k g_{k0}^2} + \frac{q_{k0}^* q_{k0}}{\lambda_k g_{k0}^2} \right) v_0^{(0)*}.
\]
Shift \( \Omega^{(2)} \) in the mean frequency to the order \( \varepsilon^2 \) as a function of the complex driving protocol \( H \) is given in terms of the bilinear form \( v(A^*, B) \):
\[
\Omega^{(2)}(H^*, H) = 2 \pi \left[ v(H^*, H) + v^*(H^*, H) \right].
\]
The bilinear form \( v(A^*, B) \) is obtained from Eqs. (A15) and (A20)–(A23) as
\[
v(A^*, B) = - \sum_{nm} \Theta_{nm} \sum_{k \neq 0} u_m^{(k)} v_0^{(0)*} \left[ A_{nm}^* \frac{v^{(k)}(t) B^{(0)}}{\lambda_k + \Lambda} - W_{nm}^0 \sum_{k \neq 0} \frac{v^{(k)}(t) A^* u^{(k)}(t) B^{(0)} \Lambda_k}{\lambda_k + \Lambda} \right].
\]
If exponent \( \Lambda \) depends in some way on frequency \( \omega_1 \) of complex driving signal \( z(t) \), the responsiveness of the frequency shift to \( \omega_1 \) is calculated from
\[
\partial_{\omega_1} v(A^*, B) = \partial_{\omega_1} A \sum_{nm} \Theta_{nm} \sum_{k \neq 0} u_m^{(k)} v_0^{(0)*} \left[ A_{nm}^* \frac{v^{(k)}(t) B^{(0)}}{\lambda_k + \Lambda} - W_{nm}^0 \sum_{k \neq 0} \frac{v^{(k)}(t) A^* u^{(k)}(t) B^{(0)} \Lambda_k}{\lambda_k + \Lambda} \right].
\]

**APPENDIX B**

We consider a biased jump process on a ring of states with forward transition rates \( W_{n+1,n}^0 = L(1 + \gamma) \) and backward transition rates \( W_{n,n+1}^0 = L \gamma \). The stationary probability flow in this system is \( J_{n,n+1}^0 = (W_{n+1,n}^0 - W_{n,n+1}^0) / L = 1 \); i.e., the time scale is chosen such that the mean period is one and the mean frequency is \( \omega_0 = 2 \pi \). We split the transition rates into an unbiased jump process with forward and backward rates \( W_{n\pm 1,n}^0 = L \gamma \) and bias \( W_{n+1,n} = L \). The perturbation shall be given as a traveling wave to the forward transitions
\[
V_{nm}(t) = W_{nm}^\Pi A(t) \cos[k_0 m - \varphi(t)].
\]
Amplitude \( A(t) \) and phase \( \varphi(t) \) of the perturbation specify complex driving signal \( z(t) = A(t) \exp[-i \varphi(t)] \), which is supposed to be symmetric and posses a discrete or continuous rotational symmetry. Then, the eigenvectors of the unperturbed system are simple harmonics, and the perturbation couples a mode \( k \) to its neighboring modes \( k \pm k_0 \). The mean frequency in the second order in perturbation strength can be calculated from the expressions derived in Appendix A. Here we will use a different approach and derive the frequency shift as well as the asymptotic phase diffusion constant from the generating function of the biased jump process on an infinite lattice. In this case, eigenmodes
\[
v_n^{(k)} = 2 \pi u_n^{(k)} = e^{ikn}, \quad k \in (-\pi, \pi)
\]
form a complete and orthogonal set with
\[
\int_{-\pi}^{\pi} v_n^{(k)}(t) u_m^{(k)}(t) \, dk = \delta_{nm} \quad \text{and} \quad \int_{-\pi}^{\pi} v_n^{(k)}(t) u_m^{(k)}(t) \, dk = \delta(k - k').
\]
The eigenvalues follow from
\[
\sum_{m=-\infty}^{\infty} W_{nm}^0 u_m^{(k)} = \lambda_k u_n^{(k)}
\]
\[
= L(1 + \gamma)\left( u_n^{(k)} - u_{n-1}^{(k)} \right) + L \gamma \left( u_n^{(k)} - u_{n+1}^{(k)} \right) = [L(1 + \gamma)e^{-ik} - 1] + L \gamma (e^{ik} - 1) u_n^{(k)}
\]
They can be divided into diffusion and flow parts as
\[
\lambda_k = \lambda_k^{\Delta} + \lambda_k^{\epsilon}, \quad \lambda_k^{\Delta} = L(e^{-ik} - 1),
\]
\[
\lambda_k^{\epsilon} = L \gamma(e^{ik} + e^{-ik} - 2).
\]
Perturbation \( V = H z(t) \) to \( H^* z(t) \) applied to mode \( u^{(k)} \) creates two neighboring modes \( u^{(k \pm k_0)} \) as
\[
H u^{(k)} = \frac{1}{2} \lambda_k^{\Delta} u^{(k + k_0)}, \quad H^* u^{(k)} = \frac{1}{2} \lambda_k^{\Delta} u^{(k - k_0)},
\]
so that
\[
q_{kk'}^{+} = v^{(k)} H u^{(k')} = \frac{1}{2} \lambda_k^{\Delta} \delta(k - k' - k_0),
\]
\[
q_{kk'}^{-} = v^{(k)} H^* u^{(k')} = \frac{1}{2} \lambda_k^{\Delta} \delta(k - k' + k_0).
\]
The mean frequency and the phase diffusion constant are defined as

$$\Omega = \frac{2\pi}{L} \lim_{\tau \to \infty} \frac{1}{\tau} \mathbb{E}[n(t + \tau) - n(t)],$$
$$D = \left(\frac{2\pi}{L}\right)^2 \lim_{\tau \to \infty} \frac{1}{2\tau} \text{Var}[n(t + \tau) - n(t)].$$

Taking the long time limit makes $\omega$ and $D$ independent from the initial conditions. Therefore, we can start with a localized distribution at $n(t) = 0$ and average over time, i.e., study the asymptotic behavior of the moments of the time-averaged transition probability $\langle P_{n0}(t + \tau | t) \rangle$. These moments are conveniently calculated from the characteristic function

$$\phi_{x}(x) = \mathbb{E}[e^{ixn}] = \sum_{n} e^{inx} \langle P_{n0}(t + \tau | t) \rangle,$$

as

$$\mathbb{E}[n] = -i\phi'(0), \quad \text{and} \quad \text{Var}[n] = -\phi''(0) + \phi(0).$$

The perturbation expansion of the time-averaged transition probabilities is

$$\langle P_{n0}(t + \tau | t) \rangle = P_{n0}^{(0)}(\tau) + \varepsilon^2 \langle P_{n0}^{(2)}(t + \tau | t) \rangle.$$  (B11)

In Appendix A, we have already derived the Fourier modes of the transition probabilities up to the second order of $\varepsilon^2$:

$$\phi_{x}(x) = \sum_{n} e^{inx} \langle P_{n0}(t + \tau | t) \rangle = \int_{-\pi}^{\pi} \sum_{n} e^{inx} u_{n}^{(0)}(\pi_{kk}^{(0)})_{l_{t}} + \varepsilon^{2} \langle \pi_{kk}^{(2)} \rangle_{l_{t}} dk',$$

$$= e^{i\lambda_{x} \tau} + \varepsilon^{2} \int_{-\pi}^{\pi} \langle \pi_{kk}^{(2)} \rangle_{l_{t}} dk.'$$

(B12)

Here we have used

$$\int_{-\pi}^{\pi} e^{ixn} \pi_{kk}^{(0)} = \delta(k + x), \quad \pi_{kk}^{(0)} = e^{i\lambda_{x}} \delta(k - k').$$

(B13)

From Eq. (A12), we see that the second-order perturbation term of average Fourier coefficient $\langle \pi_{kk}^{(2)} \rangle_{l_{t}}$ contains products of the form $q_{kk}^{+} q_{kk}^{-} q_{kk}^{++}$, and with Eq. (B7), it follows that these terms are nonzero only if $k = k'$ and $k'' = k \neq k_{0}$. All integrals over the Fourier modes are thus evaluated explicitly, and we find

$$\int_{-\pi}^{\pi} \langle \pi_{kk}^{(2)} \rangle_{l_{t}} dk' = \frac{1}{4} e^{i\lambda_{x} \tau} \frac{\lambda_{x}^{0} + \lambda_{k}^{0}}{\lambda_{x}^{0} - \lambda_{k}^{0}} - \frac{1}{4} e^{i\lambda_{x} \tau} \frac{\lambda_{x}^{0} - \lambda_{k}^{0}}{\lambda_{x}^{0} + \lambda_{k}^{0}} - \frac{1}{4} e^{i\lambda_{x} \tau} \frac{\lambda_{x}^{0} + \lambda_{k}^{0}}{\lambda_{x}^{0} - \lambda_{k}^{0}} - \frac{1}{4} e^{i\lambda_{x} \tau} \frac{\lambda_{x}^{0} - \lambda_{k}^{0}}{\lambda_{x}^{0} + \lambda_{k}^{0}} - \frac{1}{4} e^{i\lambda_{x} \tau} \frac{\lambda_{x}^{0} + \lambda_{k}^{0}}{\lambda_{x}^{0} - \lambda_{k}^{0}} - \frac{1}{4} e^{i\lambda_{x} \tau} \frac{\lambda_{x}^{0} - \lambda_{k}^{0}}{\lambda_{x}^{0} + \lambda_{k}^{0}}.$$

(B14)

Inserting Eq. (B14) into Eq. (B12), we find that the generating function has the form

$$\phi(x) = e^{i\lambda_{x} \tau} [1 + \varepsilon^{2} h(x)].$$

To emphasize the structure of $h(x)$ and take the derivative, we introduce some more shorthand notations:

$$h_{0}^{+}(x) = \frac{\lambda_{x}^{0} + \lambda_{k}^{0}}{4 \lambda_{x}^{0} - \lambda_{k}^{0}}, \quad h_{0}^{-}(x) = \frac{\lambda_{x}^{0} - \lambda_{k}^{0}}{4 \lambda_{x}^{0} + \lambda_{k}^{0}} - 1,$$

$$h_{1}^{+}(x) = \frac{\lambda_{x}^{0} + \lambda_{k}^{0}}{4 \lambda_{x}^{0} - \lambda_{k}^{0}}, \quad h_{1}^{-}(x) = \frac{\lambda_{x}^{0} - \lambda_{k}^{0}}{4 \lambda_{x}^{0} + \lambda_{k}^{0}} - 1,$$

so that

$$\phi(x) = h_{1}^{+}(x) h_{2}^{+}(x) + h_{1}^{-}(x) h_{2}^{-}(x) + [h_{1}^{+}(x) + h_{1}^{-}(x)] h_{x}^{+}(x).$$

(B17)

The necessary derivatives of the eigenvalues are

$$\frac{d}{dx} h_{x}^{+}(x) = i \left( \lambda_{x}^{0} + L \right),$$

$$\frac{d}{dx} h_{x}^{-}(x) = \left( \lambda_{x}^{0} + L \right),$$

$$\frac{d^{2}}{dx^{2}} h_{x}^{+}(x) = -2 \gamma \text{Im} \left[ \lambda_{x}^{0} \right],$$

$$\frac{d^{2}}{dx^{2}} h_{x}^{-}(x) = - \gamma \left[ \lambda_{x}^{0} + 2 \gamma L \right],$$

and with Eqs. (B15) and (B10), it follows that

$$\mathbb{E}[n] = Lx - \varepsilon^{2} ih'(0) \quad \text{and} \quad \text{Var}[n] = L(1 + 2\gamma)x - \varepsilon^{2} h'(0).$$

(B18)

(B19)

Functions $h_{0}^{+}(x)$ and all their derivatives at zero varnish when divided by $x$ in the limit $x \to \infty$

$$\lim_{x \to \infty} \frac{d^{n}}{dx^{n}} h_{0}^{+}(x) = 0 \quad (n = 0, 1, \ldots).$$

(B20)

Only derivatives of $h_{1}^{+}(x) \lambda_{x}^{0} \tau$ at zero remain in that limit. With Eq. (B16) and recalling that $g_{kk}^{x} = \lambda_{x} - \lambda_{k}^{0} + \Lambda$, we have

$$h_{1}^{+}(0) = \frac{1}{4} \frac{\lambda_{k}^{0}}{\lambda_{k}^{0} + \Lambda},$$

$$h_{1}^{-}(0) = \frac{1}{4} \frac{\lambda_{k}^{0} + L}{\lambda_{k}^{0} + \Lambda} - \frac{\lambda_{k}^{0}}{\lambda_{x}^{0} + \lambda_{k}^{0} + \Lambda^{2}} \left( \lambda_{x}^{0} + \lambda_{k}^{0} + \Lambda^{2} \right),$$

and

$$h_{1}^{-}(0) = h_{1}^{+}(0)^{*} \quad \text{and} \quad h_{1}^{-}(0) = h_{1}^{+}(0)^{*}.$$
and
\[ D = \left( \frac{2\pi}{L} \right)^2 \lim_{\tau \to \infty} \frac{1}{\tau} \text{Var}[n] \]
\[ = \frac{2\pi^2}{L} (1 + 2\gamma) - \varepsilon^2 \left( \frac{2\pi}{L} \right)^2 \lim_{\tau \to \infty} \frac{1}{2\tau} h''(0), \quad (B23) \]
so that \( D_0 = (1 + 2\gamma)2\pi^2/L \) and
\[ \frac{D}{D_0} = 1 + \varepsilon^2 \frac{1}{L(1 + 2\gamma)} \left[ 2h''(0) \frac{d^2\lambda_{\pm}}{dx^2}\bigg|_{x=0} + \text{c.c.} \right] \]
\[ + h''(0) \frac{d^2}{dx^2}\lambda_{\pm}\bigg|_{x=0} + \text{c.c.} \]
\[ = 1 + \varepsilon^2 \frac{1}{(1 + 2\gamma)} [2h''(0) - h''(0) + \text{c.c.}] \]
\[ = 1 + \varepsilon^2 \frac{1}{4(1 + 2\gamma)} \left[ 2\lambda_{\pm} \left( \lambda_{\pm}^2 + i2\gamma \text{Im} \left[ \lambda_{\pm}^n \right] \right) \right] \]
\[ = \frac{3\lambda_{\pm}^n + 2L}{\lambda_{\pm}^n + \Lambda} + \text{c.c.}, \quad (B24) \]
The relative Péclet number is
\[ \frac{c}{c_0} = 1 + \varepsilon^2 \left( \frac{\Omega^2}{\omega_0} - \frac{D(0)}{D_0} \right) \]
\[ = 1 + \varepsilon^2 \frac{1}{2(1 + 2\gamma)} \left[ (1 - \gamma)\lambda_{\pm}^n + L \right] \]
\[ - \frac{\lambda_{\pm}^n}{\lambda_{\pm}^n + i2\gamma \text{Im} \left[ \lambda_{\pm}^n \right] + \Lambda} + \text{c.c.}, \quad (B25) \]
where c.c. denotes the complex conjugate terms.

From the generating function Eq. (B9) also follows the autocorrelation of the driving signal if it is taken from another stochastic oscillator as \( z(t) = \exp[-ikn_1(t)] \):
\[ \langle z(t + \tau)|z(t)\rangle = \sum_n e^{ikn} \langle P_{t+\tau}(t + \tau)|t\rangle, \quad (B26) \]
and with Eqs. (B9) and (B15), we find
\[ \langle z(t + \tau)|z(t)\rangle = \phi(k) = e^{\lambda_{\pm}i\tau}, \quad (B27) \]
i.e., \( \Lambda = \lambda_{\pm} \) in Eq. (A3). Because the time scale of the driving signal can be different from the perturbed stochastic oscillator, eigenvalue \( \lambda_{\pm} \) is not necessarily given by Eq. (B5).

The continuum limit can be taken if \( L, \gamma \to \infty \) and
\[ D_0 = \frac{2\pi^2}{L} (1 + 2\gamma) \quad (B28) \]
is kept constant. For harmonic driving, angle \( \theta = 2\pi n/L \) in the continuum limit evolves according to
\[ \dot{\theta} = 2\pi [1 + \varepsilon \cos(\theta - \omega_1 t)] + \sqrt{2D_0} \xi(t), \quad (B29) \]
where \( \xi(t) \) is white noise with \( \langle \xi(t)\xi(t') \rangle = \delta(t - t') \). After substituting \( \varphi = \theta - \omega_1 t \), we obtain the stochastic Adler equation:
\[ \ddot{\varphi} + 2\pi - \omega_1 + 2\pi \varepsilon \cos(\varphi) + \sqrt{2D_0} \xi(t), \quad (B30) \]
for which the mean frequency and the phase diffusion constant are known [14,15]. With \( k_0 = 2\pi/L \) and \( \Delta = \omega_1 - 2\pi \), the eigenvalues and the perturbation terms become
\[ \lambda_{\pm}^{L} \to -i2\pi, \quad \lambda_{\pm}^{df} \to -D_0, \quad (B31a) \]
\[ \Omega \to 1 + \varepsilon^2 \frac{\sqrt{\Delta}}{D_0 + \Delta^2}, \quad (B31b) \]
\[ D \to 1 + \varepsilon^2 \left[ \frac{2\pi^2 (\Delta^2 - D_0^2)}{(D_0^2 + \Delta^2)^2} + \frac{2\pi^2}{D_0^2 + \Delta^2} \right], \quad (B31c) \]

**APPENDIX C**

Here we consider two stochastic jump processes \( n_0(t) \) and \( n_1(t) \) on a two-dimensional, periodic lattice. While we have used the time-dependent perturbation theory in the case of external driving, the coupled system \( n = (n_0, n_1) = n_0 e^0 + n_1 e^1 \) is autonomous. Nevertheless, because the time-independent perturbation theory is a special case of the time dependent perturbation theory, we can set \( z(t) = 1 \) and follow the derivations in Appendix B with a slight modification. The time averages have to be replaced with a suitable average over initial state \( n^0 = n(0) \). For this, we introduce unitary translation operator \( T_n^\alpha \) with \( [T_n^\alpha P(t^n)]_{mn} = P_{n^0+\alpha m} \). Transition probabilities \( P(t) = P(t + \tau|t) \) are now independent of \( t \), but Fourier modes
\[ \pi_{kk}(n^0, \tau) = \nu^{(k)} T_n^\alpha P(t) T_n^\alpha u^{(k)} \quad (C1) \]
and
\[ q_{kk}(n^0) = \nu^{(k)} T_n^\alpha H T_n^\alpha u^{(k)} \quad (C2) \]
depend explicitly on the initial conditions. Due to translational symmetry, transition matrix \( W^0 \) and the translation operator \( T_n^\alpha \) commute.

Again, we divide the transition rates into isotropic diffusion part \( W^{df} \) and a bias in the forward directions:
\[ W^\alpha = W^{df} + W^\alpha, \]
\[ W^{df}_{n+\alpha e^\sigma,n} = W^{df}_{n+\alpha e^\sigma,n} = L \gamma, \quad (C3) \]
\[ W^\alpha_{n+\alpha e^\sigma,n} = L, \quad W^\alpha_{n+\alpha e^\sigma,n} = \frac{\omega_1}{2\pi} L. \]
The oscillators are coupled weakly through their forward jump rates with the perturbation
\[ V_{n+\alpha e^\sigma,n} = L \cos(k^0 n + \beta), \quad V_{n+\alpha e^\sigma,n} = \frac{\omega_1}{2\pi} L \cos(k^0 n - \beta). \quad (C4) \]
Here \( k^0 n \) denotes the inner product. The eigenfunctions of the unperturbed system are harmonics \( i_n^{(k)} = (2\pi)^2 u^{(k)}_n = e^{ikn} \). With \( k = (k_0, k_1) \), we define
\[ \lambda_{\pm}^{L} = L(e^{-ik} - 1), \quad \lambda_{\pm}^{df} = L(e^{ik} + e^{-ik} - 2), \quad \lambda_{\pm}^{k_0} = \lambda_{\pm}^{k_0} + \lambda_{\pm}^{k_1}, \quad \lambda_{\pm}^{k_0}(\beta) = \lambda_{\pm}^{k_0} + \frac{\omega_1}{2\pi} \lambda_{\pm}^{k_1} e^{i2\beta}, \quad (C5) \]
\[ \lambda_{\pm} = \lambda_{\pm}^{L}(0) + \lambda_{\pm}^{df}. \]
Then
\[ W^{(k)} = \lambda^{(k)} u^{(k)}, \quad H^{(k)} = \frac{1}{2} e^{i\beta k} (\beta) u^{(k+k)}, \]
\[ H^{(k)} = \frac{1}{2} e^{-i\beta k} (\beta) u^{(k-k)}, \] (C6)
and
\[ q^{+}_{kk} = e^{-ik^{(k)} x} e^{-i\beta k} (\beta) d(k-k'), \]
\[ q^{-}_{kk} = e^{-ik^{(k)} x} e^{-i\beta k} (\beta) d(k-k' + k). \] (C7)

We see that the average over the initial conditions in Eqs. (A11) and (A12) removes all terms linear in \( q^{\pm}_{kk} \) and quadratic terms \( q^{+}_{kk} q^{-}_{kk} \) and \( q^{+}_{kk} q^{+}_{kk} \). The generating function is written as
\[ \phi(x) = \sum_{n} e^{i\lambda n} \langle P^{(n)}(\tau) \rangle n = e^{\lambda \tau + \gamma^{2} \tau} \int_{-\pi}^{\pi} dK^{(2)} \langle n_{0} \rangle, \] (C8)
and from Eqs. (A12), (A18), and (C7) and setting \( \Lambda = 0 \), we have
\[ \int_{-\pi}^{\pi} \langle n_{0} \rangle dK = \frac{1}{4} e^{x^{(k)} x} \frac{\lambda^{(k)} x - \lambda - \tau}{\lambda - \lambda^{(k)} x} \left( \frac{e^{x^{(k)} x - \lambda - \tau} - 1}{\lambda - \lambda^{(k)} x} \right) + \frac{1}{4} e^{x^{(k)} x} \frac{\lambda^{(k)} x - \lambda - \tau}{\lambda - \lambda^{(k)} x} \left( \frac{e^{x^{(k)} x - \lambda - \tau} - 1}{\lambda - \lambda^{(k)} x} \right). \] (C9)

As in the previous section, the generating function is of the form
\[ \phi(x) = e^{\lambda \tau + \gamma^{2} \tau} [1 + \gamma^{2} h(x)], \] (C10)
and the mean and variance of the position of the first oscillator are given, respectively, by
\[ \mathbb{E}[n_{0}] = -i \partial_{\lambda} \phi(x) \big|_{x=0}, \]
\[ \text{Var}[n_{0}] = -\partial_{\lambda}^{2} \phi(x) \big|_{x=0} + \left( \partial_{\lambda} \phi(x) \big|_{x=0} \right)^{2}. \] (C11)

With the definitions
\[ h^{+}_{1}(x) = \frac{1}{4} \lambda^{(k)} x - \lambda + \frac{\lambda^{(k)} x - \lambda - \tau}{\lambda - \lambda^{(k)} x}, \]
\[ h^{-}_{1}(x) = \frac{1}{4} \lambda^{(k)} x - \lambda - \lambda^{(k)} x - \lambda - \tau, \]
\[ h_{1}(x) = \frac{1}{4} \lambda^{(k)} x - \lambda - \lambda^{(k)} x, \]
\[ h_{2}(x) = \frac{1}{4} \lambda^{(k)} x - \lambda + \lambda^{(k)} x - \lambda - \tau, \] (C12)
function \( h(x) \) can be written as
\[ h(x) = h^{+}_{1}(x) h^{-}_{1}(x) - [h^{+}_{1}(x) h^{-}_{1}(x) - h^{+}_{1}(x) h^{-}_{1}(x)] \tau. \] (C13)

Again, only the derivatives of terms \( h^{+}_{1}(x) \lambda_{-x}^{\pm} (\pm \beta) \) remain in the limit \( \tau \to \infty \). The derivatives of the eigenvalues with respect to the first component of \( x \) are the same as those for the single oscillator
\[ \partial_{\lambda} \lambda^{(k)} x - \lambda + \lambda^{(k)} x \big|_{x=0} = \frac{d}{dx} \lambda^{(k)} x - \lambda + \lambda^{(k)} x \big|_{x=0} = \frac{d}{dx} \lambda^{(k)} x - \lambda + \lambda^{(k)} x \big|_{x=0}, \]
given by Eq. (B18) in Appendix B. Thus, functions \( h^{+}_{1}(x) \) and their derivatives at zero are
\[ h^{+}_{1}(0) = \frac{1}{4} \lambda^{(k)} x - \lambda + \lambda^{(k)} x, \quad \partial_{\lambda} h^{+}_{1}(x) \big|_{x=0} = \frac{1}{4} \left[ \frac{\lambda^{(k)} x - \lambda + \lambda^{(k)} x}{\lambda} - \frac{\lambda^{(k)} x - \lambda - \lambda^{(k)} x}{\lambda^{(k)} x} \right]. \] (C14)
as well as \( h^-(0) = h^+(0)^* \) and \( \partial_{\eta} h^-(0) = \partial_{\eta} h^+(0)^* \). The mean frequency and phase diffusion constant of the first oscillator are then determined as
\[
\Omega = \frac{2\pi}{L} \lim_{t \to \infty} \frac{1}{\tau} \mathbb{E}[n_t] = 2\pi - \epsilon^2 \frac{2\pi}{L} \lim_{t \to \infty} \frac{1}{\tau} i \partial_{\eta} h^+(x) \bigg|_{x=0} = 2\pi + \epsilon^2 \frac{2\pi}{L} i \left[ h^+_1(x) \partial_{\eta} \lambda^\eta_{-\lambda}(\beta) + h^-_1(x) \partial_{\eta} \lambda^\eta_{-\lambda}(\beta) \right] \bigg|_{x=0}
= 2\pi \left( 1 - \epsilon^2 \frac{1}{4} \left[ \frac{\lambda^\eta_{k^0}(\beta)}{\lambda_{k^0}} + \frac{\lambda^\eta_{-k^0}(\beta)}{\lambda_{-k^0}} \right] \right),
\]
(C16)
\[
D = \left( \frac{2\pi}{L} \right)^2 \lim_{t \to \infty} \frac{1}{\tau} \text{Var}[n] = \frac{2\pi^2}{L} (1 + 2\gamma) - \epsilon^2 \left( \frac{2\pi}{L} \right)^2 \lim_{t \to \infty} \frac{1}{\tau} \partial_{\eta}^2 h(x) \bigg|_{x=0},
\]
(C17)
so that
\[
\frac{D}{D_0} = 1 + \epsilon^2 \frac{1}{L(1 + 2\gamma)} \left[ 2\partial_{\eta} h^+_1(x) \partial_{\eta} \lambda_{-\lambda}(\beta) + h^+_1(x) \partial_{\eta} \lambda^\eta_{-\lambda}(\beta) + \text{c.c.} \right] \bigg|_{x=0}
= 1 + \epsilon^2 \frac{1}{1(1 + 2\gamma)} \left[ 2i \partial_{\eta} h^+_1(x) - h^+_1(x) + \text{c.c.} \right] \bigg|_{x=0}
= 1 + \epsilon^2 \frac{1}{4(1 + 2\gamma)} \left[ \frac{\lambda^\eta_{k_0}(\beta) \left( \lambda^\eta_{i\lambda} + i2\gamma \text{Im}[\lambda^\eta_{i\lambda}] \right)}{\lambda^2_{k_0}} - \frac{2(\lambda^\eta_{i\lambda} + L) + \lambda^\eta_{k_0}(\beta)}{\lambda^2_{k_0}} + \text{c.c.} \right],
\]
(C18)
and for the Péclet number
\[
\frac{c}{c_0} = 1 + \epsilon^2 \left( \frac{\Omega}{\omega_0} - \frac{D}{D_0} \right) = 1 + \epsilon^2 \frac{1}{2(1 + 2\gamma)} \left[ \frac{\lambda^\eta_{i\lambda} - \gamma \lambda^\eta_{k_0}(\beta) + L}{\lambda^2_{k_0}} - \frac{\lambda^\eta_{k_0}(\beta) \left( \lambda^\eta_{i\lambda} + i2\gamma \text{Im}[\lambda^\eta_{i\lambda}] \right)}{\lambda^2_{k_0}} + \text{c.c.} \right].
\]
(C19)