Additive noise quenches delay-induced oscillations

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received 14 March 2013; accepted in final form 3 June 2013
published online 4 July 2013

PACS 02.30.Ks – Delay and functional equations
PACS 05.45.-a – Nonlinear dynamics and chaos
PACS 02.50.Fz – Stochastic analysis

Abstract – Noise has significant impact on nonlinear phenomena. Here we demonstrate that, in opposition to previous assumptions, additive noise interferes with the linear stability of scalar nonlinear systems when these are subject to time delay. We show this by performing a recently designed time-dependent delayed center manifold (DCM) reduction around a Hopf bifurcation in a model of nonlinear negative feedback. Using this, we show that noise intensity must be considered as a bifurcation parameter and thus shifts the threshold at which delay-induced rhythmic solutions emerge.

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Introduction. – Noise has a strong influence on the dynamics of systems subject to time-delayed retroaction. Random fluctuations impact laser dynamics [1], gene regulatory networks [2], postural and mechanical control [3], which all exhibit retarded self-interactions. The presence of time delay also often appears along stochastic synchronization in complex networks [4–6]. In neuroscience, the propagation and time evolution of neural activity has been shown to vary according to the combined action of sustained stochastic driving and recurrent feedback, key components of brain circuits [7]. As such, proper characterization of systems away from the stationary regime is essential in order to understand how these elements conspire and give rise to the complex phenomena we observe in experiments.

Systems evolving under mixtures of noise and nonlinear feedback express divergent properties from their deterministic counterparts, where noise has been shown to provoke or prevent bifurcations [8]. However, such noise-induced stability shifts in scalar systems are thought to be caused only by a subgroup of stochastic influences, namely multiplicative or parametric noise. This specific type of noise has indeed been shown to be the source of state transitions in nonlinear systems, while its implication on the stability of delay equations is a well-documented phenomenon [9–11].

In the past, numerous studies have determined or assumed that additive noise, that is stochastic fluctuations independent of the state variable, has no impact on the stability of nonlinear systems [12], even in the presence of delays [13]. This implies that the deterministic expression of the dynamics fully exposes the mean behavior of an otherwise stochastic system. This view has since been challenged in mean-field–typesystems [14,15] and physical pattern-forming systems [16]. More recently, it has been demonstrated that additive noise stabilizes delay nonlinear systems near non-oscillatory instabilities [17,18]. In this letter, we show that additive stochastic fluctuations cause a shift in the eigenspectrum in a nonlinear scalar delayed differential equation (DDE) in the vicinity of a Hopf bifurcation and de facto shape delay-induced rhythms. By doing so, we state that despite the current belief, additive noise changes significantly the stability of delayed oscillatory nonlinear systems and that the additive noise intensity has to be considered as a bifurcation parameter.

To demonstrate this, we adopt the following strategy. We first derive a time-dependent delayed center manifold (DCM) representation of a generic scalar DDE unfolded around a supercritical Hopf bifurcation. Then, adiabatic elimination is performed based on the characteristic separation of time scales involved in the vicinity of a non-hyperbolic fixed point. We show that the explicit time dependence of the DCM reveals noise-dependent components responsible for the depletion of rhythmic solutions.
for negative nonlinearities, and consequently exposes the impact of additive noise on the system’s equilibrium and its stability. We then apply our theory to reconstruct the associated stochastic bifurcation diagram of the ensemble-averaged noisy solutions, and validate it via numerical simulations.

Model. – The work is motivated by non-homogeneous retarded equations with the generic form

\[ \dot{x}(t) = -x(t) + \gamma f(x(t - \tau)) + I(t), \]  

(1)

exhibiting a smooth nonlinearity \( f(x) \) and some additive random fluctuations \( I(t) \), in the Itô sense. Let us consider the expansion of eq. (1) around the stationary state \( x_0 = \gamma f(x_0) \) up to third order, which yields the polynomial system

\[ \dot{u}(t) = -u + \eta u + \kappa u^2 + \nu u^3 + \sqrt{2D}\xi(t), \]  

(2)

where \( \eta = \gamma f'(x_0) \), \( \kappa = \gamma f''(x_0)/2 \) and \( \nu = \gamma f'''(x_0)/3 \) are non-zero constants, with the small deviations from the stationary state \( u = u(t) \) and \( u_\tau = u(t - \tau) \). Here, the additive random fluctuations are defined by \( I(t) = \sqrt{2D}\xi(t) \) with the noise intensity \( D > 0 \) such that \( \langle \xi(t)\xi(t') \rangle = \delta(t - t') \) where \( \langle \cdot \rangle \) denotes the ensemble average.

Figure 1 shows significant differences between the dynamics of eq. (2) for \( D = 0 \) and \( D > 0 \) close to the bifurcation point. While virtually masked by local fluctuations in single-trial trajectories, the investigation of the ensemble average reveals that additive noise stabilizes the oscillations by preventing the bifurcation. This is caused by a shift of the eigenspectrum performed by noise and delay. The impact of noise decreases as \( |\epsilon| \) increases, i.e., away from the bifurcation, the time scale separation vanishes and the system’s ensemble average becomes somewhat passive to additive noise.

**Unfolding around the Hopf bifurcation.** – In order to explain this novel phenomenon, the subsequent paragraphs first neglect noise and follow the standard analysis of DCM reduction for deterministic delay equations [19–22]. Analyzing the stationary state’s linear stability, we consider the linear terms in eq. (2),

\[ \dot{u}(t) = -u + \eta u_\tau \]  

(3)

and focus on the Hopf instability assuming \( u(t) = u_0 \exp(\lambda t), \lambda \in \mathbb{C}, u_0 = \text{const.} \) Then, there is a critical \( \lambda_c = \pm iw_c \) that satisfies

\[ \lambda_c = -1 + \eta_\epsilon e^{\lambda_c \tau_c}, \]  

(4)

where \( w_c = \sqrt{\eta^2 - 1} \) is the corresponding Hopf frequency. Introducing the control parameter \( \epsilon = \eta - \eta_\epsilon \), it is convenient to rewrite eq. (2) as

\[ \dot{u}(t) = L(u, \eta_c) + F(\epsilon, u) + \sqrt{2D}\xi(t), \]  

(5)

with \( L(u, \eta_c) \equiv -u + \eta_c u_\tau \) and \( F(\epsilon, u) \equiv \epsilon u_\tau + \kappa u^2 + \nu u^3 \). In this formulation, the Hopf bifurcation occurs for \( \epsilon = 0 \).

**DCM reduction neglecting noise.** – Equation (3) and the characteristic equation (4) define eigenvectors \( \Phi(\theta) = (\phi_1(\theta), \phi_2(\theta)) \), where \( \theta \) is a parametrization of the delay, i.e., \( -\tau \leq \theta \leq 0 \). They span the center subspace and are determined by \( \pm iw_c \phi_j(0) = L(\phi_j, \eta_c) - \phi_j(0) + \eta_c \phi_j(-\tau) \) \( j = (1, 2) \) yielding

\[ \Phi(\theta) = (\phi_1(\theta), \phi_2(\theta)) = (e^{i\omega_c \theta}, e^{-i\omega_c \theta}). \]  

(6)
The associated adjoint basis of the center subspace $\Psi(s) = (\psi_1(s), \psi_2(s))^T = (d_{c}^{-iws}, d_{e}^{iws})^T$, $d \in \mathbb{C}$, obeys $(\Phi, \Psi) = \mathbb{I}$, where $(a(\theta), b(\theta))$ is a bilinear form defined by [19]

$$a(b) = a(0)b(0) - \int_{-\tau}^{0} \int_{0}^{\theta} a(\xi - \theta)b(\xi)d\xi[\delta(\theta)].$$

(7)

Here, $[\delta(\theta)] = (-\delta(\theta) + \eta_0(\theta + \tau))d\theta$. According to the center manifold theorem, phase space can be split into center $(C)$ and stable $(S)$ parts, i.e., $C \oplus S$, such that the stables modes converge to the center manifold. We thus have

$$u_t(\theta) = \Phi(\theta)z + h(\theta, \bar{z}) \in C$$

for some stable manifold $h$, where $z = [z_1(t), z_2(t)]$ is the amplitude on the center parameter equation.

The state probability density $p(z, h_t)$ of the center modes $z$ evolving according to eq. (9) considers $h_t$ as a time-dependent random variable and obeys the Fokker-Planck equation,

$$\frac{\partial p(z, h_t)}{\partial t} = \left\{ \psi^2(0)D \frac{\partial^2 p(z, h_t)}{\partial z^2} - \frac{\partial}{\partial z_1} \left( iw_z + \psi_1(0)F[g(z) + h_t] \right) p(z, h_t) \right\} + \left\{ \psi^2(0)D \frac{\partial^2 p(z, h_t)}{\partial z^2} - \frac{\partial}{\partial z_2} \left( -iw_z + \psi_2(0)F[g(z) + h_t] \right) p(z, h_t) \right\},$$

(12)

where $g(z) = \Phi(\theta)z + h(\theta, \bar{z})$ is the time-independent contribution to the center manifold. Here the dependence on $\theta$ and the trivial dynamics of the unfolding parameter $c$ have been dropped for clarity.

The characteristic time scale separation involved near non-hyperbolic fixed points implies that the time-dependent functional $h_t(t)$ and the noisy input $\xi(t)$ are much smaller than the center modes $z$ and retain the time scale separation of the unperturbed system. By virtue of the separable ansatz in eq. (10) for which the temporal correction $h_t$ is independent of $z$, the adiabatic elimination [16,25] of the fast-time-dependent terms applies by performing an ensemble average of $p(z, h_t)$ over $h_t$ for all $\theta$'s in eq. (12) using the associated density $q(h_t)$. Specifically, for weak noise one separates the slow and fast evolution by $p(z, h_t) = p_0(z, t)q(h_t)$ and expand the nonlinearity $F$ with respect to $h_t$ around the center modes.

Assuming stationarity of $h_t$ and a Gaussian profile for $p(h_t)$ with vanishing mean and variance $\sigma^2 = \sigma^2(D) = \int_{0}^{\tau} h_t^2(\theta, s)p(h_t(\theta, s))ds > 0$ for a short-time window $T \ll \pi / w_c$ shorter than the time scale of the center dynamics

$$\int_{t}^{t+\tau} F[g(z(t)) + h_t(\theta, s)]p(h_t(\theta, s))ds \approx \epsilon g_x(z(t)) + \kappa (g_x^2(z(t)) + \sigma^2) + \nu \left( g_x(z(t)) + 3g_x(z(t))\sigma^2 \right).$$

with $g_x(z(t)) = \Phi(-\tau)z(t) + h(z(t), -\tau)$. The resulting adiabatic Fokker-Planck equation determines the temporal evolution of the probability density $p_0(z, t)$ whose mean value $\bar{z}$ obeys the non-autonomous order parameter equation

$$\dot{z}(t) = c_o + B\bar{z}(t) + \Psi(0)F(e^*, \Phi(z)(t) + h(z(t), \theta),$$

(13)

with the effective control parameter $e^*(D) = \epsilon + \mu(D)$, $c_o(D) = \kappa \sigma^2(D)$ and $\mu(D) = 3\mu \sigma^2(D)$.

A comparison between the full and reduced description reveals the mechanism by which additive noise alters the properties of the noise-free system. To see this, note that
Fig. 2: (Colour on-line) Delayed stochastic Hopf bifurcation diagram. Amplitudes of the nascent oscillatory solutions as a function of noise intensity \((D)\) and distance from the bifurcation \((\epsilon)\). The Hopf threshold is located at \(\epsilon = 0\) (dashed line across plots for comparison). (a) Generic bifurcation diagram in the absence of noise \((D = 0)\). (b) As noise is injected additively in the system, the onset of stable oscillatory solutions retreats in parameter space compared to the noise-free case depicted in panel (a), giving rise to significantly different behavior. The panel shows the amplitudes of the ensemble-averaged solutions \(\langle u(t) \rangle\) of eq. (2). (c) Amplitudes of the corrected system (14) capture the additive noise effect. Deviations between the noisy diagram and the DCM reduction prediction (between panels (b) and (c)) appear as \(D\) and \(|\epsilon|\) increase pointing to the approximative nature of the separation ansatz (10). Other parameters are taken from fig. 1.

eq. (13) may be viewed as the reduced dynamics of the non-autonomous DDE,

\[
\dot{Z}(t) = c_0(D) + L(Z(t), \eta_c) + F(\epsilon^*(D), Z(t))
\] (14)

This equation is our major result. By direct comparison with eq. (5), two effects can be seen in this system with cubic nonlinearity: i) Since \(c_0 \neq 0\), for \(D > 0\) the system has a new equilibrium repositioned in phase space and, ii) the noise shifts the control parameter by a factor of \(\mu\) and thus alters the stability of the system. Note that eq. (14) converges to eq. (5) when \(D = 0\). Summarizing, the sign and magnitude of the nonlinear parameters \(\kappa\) and \(\nu\) in the expanded system fully determine the qualitative impact of the additive perturbations on stability. This finding demonstrates the purely nonlinear nature of phenomenon shown in fig. 1. Moreover, by re-scaling time by \(t \rightarrow \tau t\), one may see that \(c_\alpha, \mu \sim \tau\) (details not shown), and hence, the position of the stationary state and the system stability depend on both the noise intensity and the delay.

To test our theory, we have numerically investigated the behavior of the DDE in eq. (2) with \(D = 0\) and \(D > 0\) where the delayed nonlinearity is taken to be negative \((\gamma < 0)\). To compare the original dynamics with the DCM corrected system in eq. (14), we have chosen a regime of large delays where the center modes oscillate at a small frequency and where the impact of noise on the dynamics is known to be weak [24]. To determine \(\sigma^2\), we have integrated numerically the linear system in eq. (11), applied the corresponding definition of \(h_\epsilon\), and computed the stationary variance \(\sigma^2 \approx (1 - 2\text{Re}(d) \cos(\omega_\epsilon \tau))^2 \langle H^2 \rangle\), which holds true for \(D > 0\) small. The constant \(d\) is the normalization constant of the adjoint basis \(\Psi(s)\). We note that, to the lowest order, one can derive expressions for the variance of \(\langle H^2 \rangle\) based on eq. (11) and expose the conjugate dependence of \(\sigma^2\) on both the delay and noise intensity [26].

While set in such a regime of slow oscillations, fig. 2 shows that additive noise shifts the Hopf bifurcation to smaller values and thus stabilizes the system. The corrected system (14) captures very well the dynamics altered by additive noise and further allows to predict the position of the noise-shifted Hopf bifurcation in parameter.
space. Indeed, the DCM reduction scheme captures both qualitatively and quantitatively the effect of additive noise on the dynamics of the original equation in the neighborhood of the bifurcation whenever $D$ and $|\epsilon|$ are chosen small. It further reveals the characteristic quenching action of the random fluctuation on the ensemble-averaged solutions experienced in the presence of negative delayed feedback.

One may ask whether the effect revealed, i.e., the shift of the control parameter and the oscillation amplitude, is prominent enough to play a significant role in the system’s dynamics. Figure 3 shows oscillation amplitudes subjected to the control parameter $\epsilon$ for two noise levels. It reveals that the oscillation amplitudes are practically indistinguishable for $\epsilon < -0.25$ or $\epsilon > 0.02$, i.e., away from the original deterministic bifurcation threshold. Hence the additive noise effect on the ensemble average of the system activity is observable near deterministic stability threshold but disappears far from the bifurcation point.

**Discussion.** – Our results reveal that additive noise perturbs the stability of scalar delayed systems and shifts the instability threshold in parameter space, demonstrating that noise intensity must be considered as a bifurcation parameter. Using time-dependent DCM reduction, we have exploited the characteristic time scale separation emerging near non-hyperbolic fixed points to perform adiabatic elimination. For the cubic system considered, our approach reveals the effect of additive noise on linear and cubic terms. This perturbs the stability of the systems and shifts the onset of oscillatory solutions.

This novel finding has several implications. First, the effect suggests a general mechanism for on-line suppression of rhythmic activity by external stochastic driving in the presence of inhibitory feedback, despite purely deterministic model parameters. We conjecture that this mechanism might provide support to experimental paradigms seeking to suppress undesired or detrimental rhythmic regimes in recurrent systems.

The work considers the supercritical Hopf bifurcation for a delayed nonlinearity with negative gain ($\gamma < 0$). Although our theory is not specific to this case, we note that in the vicinity of a subcritical instability, additive noise causes the solutions to diverge and that, in most practical applications, the supercritical case prevails. This case nonetheless suggests that noise in the subcritical margin might have the opposite effect: it might enhance oscillatory behavior. This remains to be studied in future work.

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