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Excitable systems with noise and delay, with applications to control: Renewal theory approach

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We present an approach for the analytical treatment of excitable systems with noise-induced dynamics in the presence of time delay. An excitable system is modeled as a bistable system with a time delay, while another delay enters as a control term taken after Pyragas [K. Pyragas, Phys. Lett. A 170, 421 (1992)] as a difference between the current system state and its state τ time units before. This approach combines the elements of renewal theory to estimate the essential features of the resulting stochastic process as functions of the parameters of the controlling term.

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I. INTRODUCTION

Oscillations induced by noise are observed in many nonlinear systems with dissipation including neural networks, lasers, chemical reactions, biomembranes, etc. (see Ref. [1] and references therein). At present, three different types of systems are being recognized that demonstrate noise-induced oscillations: bistable systems [2], systems close to Andronov-Hopf bifurcation [3], and excitable systems [1].

Much effort has been put into the development of the analytic description of excitable systems. The approach of the renewal theory [4] was successfully adopted to describe the real excitable systems with continuous dynamics by a two-state [2] or by a three-state [5,6] discontinuous stochastic process. In the frames of the renewal theory, it is assumed that the system is allowed to be only in a finite number of states, and the noise makes the system switch between the states in a random manner. The time the system spends in a given state before undergoing the next transition is called residence time in this state. For the validity of the renewal theory it is crucial that the distribution function of the residence time in a given state does not change in time. Therefore, in the case of an n-state system, its entire dynamics is determined by n distribution densities of the residence times, also known as residence time densities (RTDs). For the FitzHugh-Nagumo system the RTDs were explicitly calculated in Ref. [2] using the Fokker-Planck equation approach suggested in Ref. [7]. Phenomenological three-state model of the excitable system with the arbitrary RTDs was discussed in Refs. [5,6].

The situation becomes more complicated when a delay term is introduced into an excitable system, and the random process in it becomes non-Markovian. This problem arises for example in relation to the problem of controlling noiseinduced motion.

Usually noise-induced oscillations possess a certain time scale that is dependent on the parameters of the applied noise, e.g., its intensity. One possible way to define this time scale for an excitable system is to estimate the power spectrum that would normally contain one or more peaks, and to take the inverse of the frequency of the highest peak to be the main period of oscillations. In Ref. [8] the idea was introduced to control the properties of oscillations induced merely by external noise by applying time-delayed feedback force

F(t) in the Pyragas form [9] $F(t)=k[x(t-\tau)-x(t)]$, where τ is time delay and k is the feedback strength. It has been shown that the time scale and the coherence of noise-induced oscillations can be changed by adjusting solely the delay time in the controlling force. Moreover, an almost piecewise-linear dependence of the main period on the delay time was revealed. Remarkably, a similar behavior of the main period was found in systems with time-delayed feedback that were either excitable, or close to Andronov-Hopf bifurcation [8,10]. For a van der Pol oscillator this phenomenon was explained by means of computation of the power spectrum of oscillations analytically: by using either linearization [11,12], or mean-field approximation [12,13].

The theory of the excitable systems with time delay is still missing. Some progress was made in Ref. [14], where the impact of the delayed feedback on the current dynamics of the system was considered in the mean-field approximation. Here we propose an alternative approach based on the analysis of renewal processes with history-dependent RTDs. As mentioned above, the renewal theory can only be applied in the case when the RTDs do not change in time. This condition is obviously violated for the processes with time delay, in which RTDs are dependent on history.

In order to overcome this problem, we introduce *equilibrium* RTDs by averaging over all possible histories. This approach is somewhat similar to the one suggested in Ref. [15] for calculating the RTD in the case of stochastic resonance, where the noise-induced switchings between the two potential wells are considered in the presence of a weak periodic perturbation. Namely, in Ref. [15] the unknown RTD is obtained by averaging the known distribution of escape times (time it takes to escape from a certain well) over the known probability that the system spends a certain amount of time in the left well before entering the right well. In contrast to this, we show that in the case of *history-dependent* renewal process, the equilibrium RTDs are given by the solution of an integral equation. This equation is derived for an arbitrary, and solved analytically for a moderate, delay time.

In order to compare the analytic results of the modified renewal theory with the numerical results for an excitable system, we design an excitable system from a bistable system with a nonsymmetric potential and with a time delay, following the idea of the delay-induced excitability proposed in Ref. [16]. To this system we add the controlling force F(t)

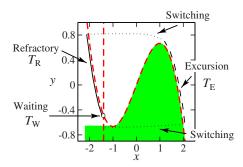


FIG. 1. (Color online) Phase plane of FitzHugh-Nagumo system, Eqs. (1), in an excitable regime. Dashed gray lines, null clines; empty circle, fixed point; shaded area, the area from which an excursion can start. Stages of one full oscillation are shown: black dashed line, excursion; black solid line, refractory; dotted lines, switchings between the two branches (very fast).

as above. The parameters of the two-state model used for analytics are matched to the parameters of the bistable system via the Kramers formula for the transition rates Ref. [17–19]. We employ the hybrid approach by combining the renewal theory results with the equilibrium RTDs of the history-dependent process in order to approximate the power spectrum of the noise-induced oscillations. Incoherence maximization due to delay is demonstrated for positive feedback strength k only.

II. THE MODEL

A. Excitability

Before describing the way to construct an excitable system from a bistable system with delay following [16], we need to explain the concept of excitability. A popular toy model of an excitable system is a FitzHugh-Nagumo system

$$\dot{x} = \left(x - \frac{x^3}{3} - y\right) \frac{1}{\epsilon},$$

$$\dot{y} = x + a + D\zeta(t),\tag{1}$$

where $\zeta(t)$ describes random fluctuations with Gaussian distribution, zero mean, and unity variance. The null clines of this system, which are the curves defined by $\dot{x}=0$ and $\dot{y}=0$ (assuming that D=0) are shown in Fig. 1 by gray dashed lines. Usually this system is considered at $\epsilon \leq 1$ in order to provide time scales separation described below. At a > 1 the system has a single stable fixed point (empty circle in Fig. 1), and no oscillatory dynamics without noise (D=0). When Gaussian noise is applied (D>0), the behavior of the system changes drastically. Namely, while the value of $D\xi(t)$ is small, the system oscillates randomly around the fixed point during the waiting phase of duration T_W [note the horizontal plateaus of the realization of x(t) in Fig. 2(a)]. But when the values of $D\xi(t)$ are large enough to throw the phase point into the gray area in Fig. 1, the system quickly tends to the right-hand branch of the cubic parabola along an almost horizontal path (lower dotted line in Fig. 1), since due to the

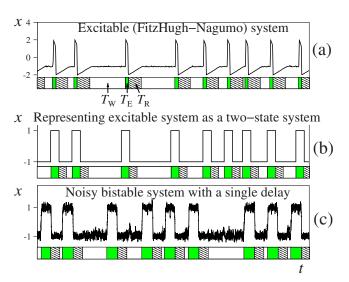


FIG. 2. (Color online) (a) Realization x(t) of an excitable system, Eq. (1). (b) Stages of oscillations in (a) modeled by a two-state process. (c) Realization of a noisy bistable system with delay, Eq. (2). Lower panels of (a)–(c) show stages of the process: white, waiting with duration T_W ; shaded, excursion with duration T_E ; patterned, refractory with duration T_R .

smallness of ϵ the horizontal component \dot{x} of the phase velocity is much larger than its vertical component \dot{y} . Then the system enters its excursion stage in which the phase point slowly crawls upwards along the right-hand branch of the parabola during T_E time units (black dashed line in Fig. 1): this motion is smeared by noise, but its velocity is almost unaffected by the latter, at least in average. As soon as the phase point reaches the top of the right-hand parabola branch, the vector flow swiftly carries it towards the left-hand branch (upper dotted line in Fig. 1). Finally, the system enters its refractory stage as the phase point slowly crawls downwards towards the fixed point during time T_R (black solid line in Fig. 1), again almost unaffected by noise in average. Then the process is repeated.

A typical realization of a stochastic process x(t) occurring in system (1) looks similar to the profile shown in Fig. 2(a), where an x variable from the FitzHugh-Nagumo system in the excitable regime is plotted. The cells in the lower panel show different stages of the process: white, waiting; shaded, excursion; patterned, refractory. One can single out two essential features of this motion induced merely by external noise. First, one can distinguish between very fast motion between the two branches of the cubic parabola, and slow motion along the branches. If ϵ is very small, one can assume that the switching between the branches occurs instantly, i.e., the system can be only in one of the two states corresponding to the two branches of the parabola. Second, the durations T_E and T_R of the excursion and of the refractory stages can be approximately regarded as independent of noise and constant in any event of a large excursion in the phase space, while the duration T_W of the waiting stage is completely determined by noise. The distribution density of T_W depends both on noise and on the design of the system.

The dynamics of the FitzHugh-Nagumo system in the excitable regime can be approximated by a two-state stochastic

process s(t) which can take only two values, say, $s=\pm 1$ [Fig. 2(b)]. We assume that in the excursion stage the system is in state s=+1, and in the waiting stage in s=-1. In addition, we assume that in the refractory stage s(t) does not change and assign s=-1 as in the waiting stage. Therefore, residence time in the state (-1) is equal to the sum of the refractory and waiting times (T_R+T_W) .

The dependence of the transition rate λ on the parameters of the FitzHugh-Nagumo system is rather complex and can be obtained only numerically. However, for a bistable system this information is readily available via the Kramers formula [17–19]. One can also neglect the intrawell dynamics of a bistable system and match the parameters of the latter with the parameters of a two-state system.

B. Bistable excitable system with delayed feedback

The idea of the present approach is to construct an excitable system from a bistable system by including a delay term into the latter, following the concept of delay-induced excitability proposed in Ref. [16]. Namely, in Ref. [16] it was shown that a two-state system with a single time delayed feedback T can behave as an excitable one if the double-well potential describing the system is asymmetric. A characteristic feature of this model is that there is a locking of state, i.e., the transition from one state into another is allowed to occur not earlier than T seconds after the moment of the previous transition. This feature is illustrated in Fig. 2(c) where a realization of such a bistable system is shown, with the deeper potential well being located around the value of x=-1. Here, one can recognize the same stages that occur in an excitable system [compare with Fig. 2(a)]: waiting marked as a white cell in the lower panel, excursion marked by shaded, and refractory marked by a patterned cell. Note, that in the bistable system with a single delay T the following condition is automatically satisfied by the way of construction: T_E $=T_R=T$. As a result, the distribution densities of the waiting times in both states are shifted to the right by T as compared to the case without delay.

We would like to assess the effect of the controlling term F(t) on this system, in which the second delay appears. The second delay term is due to the delayed feedback force.

The dynamics of the bistable system with two time delays is described by the equation

$$\dot{x}(t) = -\frac{\partial U[x, x_T]}{\partial x} + kf(x_\tau - x) + \sqrt{2D}\zeta(t), \qquad (2)$$

where x_T and x_τ denote the retarded variables x(t-T) and $x(t-\tau)$, respectively; T is the fixed excursion and refractory time, $f(x_\tau-x)$ is a controlling force, with the function f being in general a nonlinear function of its argument. τ is the delay time and k is the strength of the controlling force. We require that f must satisfy the condition f(z)=0 when z=0 in order to mimic the property of the Pyragas feedback force [9] that preserves the number and position of the fix points in the system. $\zeta(t)$ in Eq. (2) is Gaussian noise, and D is the noise strength.

The potential $U[x,x_T]$ in Eq. (2) is chosen to have two wells with minima located at $x=\pm 1$ that are separated by a

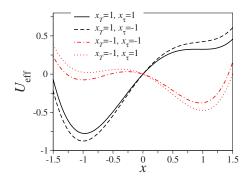


FIG. 3. (Color online) Effective potential $U_{\text{eff}}[x, x_T, x_\tau]$ (4) for k=0.05 and fixed x_T and x_τ as in the legend.

barrier with the maximum at x=-0.3, so that

$$U[x,x_T] = \frac{x^4}{4} - \frac{x^2}{2} - (0.1 + 0.2x_T)(x^3 - 3x),$$

$$\frac{\partial U[x, x_T]}{\partial x} = (x^2 - 1)(x - 0.3 - 0.6x_T). \tag{3}$$

Following Ref. [17], we assume the overdamped case when we can neglect the intrawell dynamics. It is well known that this approximation is valid for noise intensities much smaller than the height $\Delta U_{\rm eff}$ of the potential barrier, i.e., for $D \ll \Delta U_{\rm eff}$.

The special choice of the positions of two potential minima does not affect the results presented below. Also, the analytic results presented below are valid for arbitrary nonlinear controlling force f, provided that kf remains small, rendering x_T and x_τ to be either in the left or in the right well at any time moment. With this, the values they are taking are close to -1 or to +1, respectively, therefore we single out two states of the system that we denote as (-1) and (+1). In the analytical calculations below we will substitute the exact instantaneous values of x_T and x_τ by their approximate values ± 1 .

For the sake of simplicity, we illustrate most of our results (except in Sec. V B) for linear f, i.e., f(z)=z. In this case the evolution equation (2) can also be rewritten in terms of an effective potential $U_{\rm eff}$ which includes the controlling force kf,

$$U_{\text{eff}}[x, x_T, x_\tau] = U[x, x_T] + k \frac{x^2}{2} - kx_\tau x.$$
 (4)

With Eq. (4) the evolution equation (2) becomes

$$\dot{x}(t) = -\frac{\partial U_{\text{eff}}[x, x_T, x_\tau]}{\partial x} + \sqrt{2D}\zeta(t). \tag{5}$$

In Fig. 3 the effective potential $U_{\rm eff}$, given by Eq. (4), is shown for fixed x_T and x_τ that take values ± 1 . Consider all stages of one oscillatory cycle. Start from $x_T \approx +1$. From Fig. 3 it is clear that at any x_τ there exists only one well (left) and the particle trapped in it will remain there until x_T switches to approximately (-1). This is the refractory phase with duration T_R . At $x_T \approx -1$ the right well appears and the waiting

TABLE I. Transition rates λ between the states of the bistable system with delay at different values of x_T and x_T .

	$x_T = +1$ $x_T = +1$	$x_T = +1$ $x_{\tau} = -1$	$x_T = -1$ $x_\tau = +1$	$x_T = -1$ $x_\tau = -1$
$(-1) \rightarrow (+1)$	0	0	p+q	p
$(+1) \to (-1)$	∞	∞	0	0

phase begins, during which it is possible to jump from the left to the right well. Transition rate λ from (-1) to (+1) becomes dependent on x_{τ} ; for k > 0 it is larger when $x_{\tau} \approx +1$, i.e., when the left well is shallower. After the particle jumps into the right well, the system is in the excursion phase. Note that at $x_T \approx -1$ the right minimum is much deeper than the left one. Therefore, the particle trapped in it will remain there until x_T changes from (-1) to (+1), i.e., until the right well vanishes. After that the particle jumps from state (+1) to the state (-1) and the cycle repeats again.

The information on the transition rates is summarized in Table I. Here we denote by (p+q), q>-p, the transition rate from the state (-1) to the state (+1) if $x_{\tau} \approx +1$ and $x_{T} \approx -1$ and by p, p>0, the transition rate from the state (-1) to the state (+1) if $x_{\tau} \approx -1$ and $x_{T} \approx -1$. The transition rates are calculated from $U_{\rm eff}$ according to the Kramers theory [17] as follows:

 $p = \frac{1}{2\pi} \sqrt{-\partial_{xx} U_{\text{eff}}(x_m, -1, -1) \partial_{xx} U_{\text{eff}}(x_0, -1, -1)}$

$$\times \exp\left[-\frac{\Delta U_{\text{eff}}}{D}\right],$$

$$p + q = \frac{1}{2\pi} \sqrt{-\partial_{xx} U_{\text{eff}}(x'_m, -1, 1)\partial_{xx} U_{\text{eff}}(x'_0, -1, 1)}$$

$$\times \exp\left[-\frac{\Delta U'_{\text{eff}}}{D}\right],$$

$$(6)$$

where x_0 and x_m are the positions of the maxima and the left minima of the effective potential $U_{\rm eff}$ at $x_T = x_\tau = -1$ and x_0' and x_m' are the same quantities at $x_T = -1$, $x_\tau = +1$. $\Delta U_{\rm eff}$ and $\Delta U_{\rm eff}'$ are the potential differences between the maxima and the minima. Note that due to presence of the feedback term $k(x_\tau - x)$, the values of x_m and x_m' are not exactly equal to -1, and x_0 is not exactly -0.3. However, in analytic calculations we substitute $x_0 = x_0'$, x_m , and x_m' by their approximate values -0.3, -1, and -1, respectively. Throughout the paper we compare the analytic results derived from the two-state model with the results of simulation of a bistable system with two time delays.

III. TWO-STATE MODEL: ANALYTIC RESULTS

The continuous random process x(t) is approximated by the discrete random process $s(t) = \pm 1$ with infinitely fast (discontinuous) transitions from one state to the other. In order to match the parameters of the two-state model with

those of the bistable system, Eq. (2), the excursion and the refractory times must be set equal, i.e., $T_E = T_R$ [16]. However, for the sake of generality the analytic results presented below were obtained for the case of $T_E \neq T_R$. In what follows by ξ we denote the time of residence in a certain state: from the context it will be clear what state or phase is referred to.

Regarding the dependence of the transition rates on the history (Table I) we conclude that the RTD $\psi_+(\xi)$ in the state s=+1 is given by the δ function $\psi_+(\xi)=\delta(\xi-T_E)$. This assumption is justified if there is a strong time scale separation, meaning that the transition from the excited to the nonexcited state and backwards occurs almost instantaneously. Moreover, the RTD in the state s=-1 is zero for the first T_R seconds after the transition from the state s=+1 to the state s=-1. Consequently, the model contains only one unknown object, namely the distribution density $\psi_-(\xi)$ of the waiting time. It is related to the RTD $\Psi_-(\xi)$ in the state s=-1, via

$$\Psi_{-}(\xi) = \begin{cases} 0, & \xi \in [0; T_R], \\ \psi_{-}(\xi - T_R), & \xi \in [T_R; \infty). \end{cases}$$
 (7)

A. Small delay times τ : Nonvariable history

We start with the case of small delay times from the interval $\tau \in [0; T_E + 2T_R]$. The probability $P_-(\xi)$ of survival in the state (-1) during the waiting phase is given by the solution of the equation

$$-\psi_{-}(\xi) = \frac{\partial P_{-}(\xi)}{\partial \xi} = -\lambda_{-}(\xi)P_{-}(\xi), \tag{8}$$

where $\lambda_{-}(\xi)$ stands for the transition rate from s=-1 to s=+1. If $\lambda=$ const, which is valid for the bistable systems without delays, the statistics of the waiting times is exponential [4] with the distribution density ψ given by

$$\psi(\xi) = \lambda \exp(-\lambda \xi), \tag{9}$$

where the constant transition rate λ depends on the system parameters. The mean waiting time $\langle T_W \rangle$ is then given by $\langle T_W \rangle = 1/\lambda$. However, due to delay there appears a discontinuity in the transition rate $\lambda_-(\xi)$ (see Table I), therefore the solution of Eq. (8) on the intervals where $\lambda_-(\xi)$ is constant must be normalized in such a way that the survival probability $P_-(\xi)$ remains continuous.

Suppose that at time $t=t_0$ [Fig. 4(a)] the refractory phase has just finished and the system is in the very beginning of the waiting phase, implying that the waiting duration is equal to zero, $\xi=0$. Figure 4(a) shows the profile of the two-state process on the interval of time $t \in [t_0-T_E-2T_R;t_0]$. It is clear that no other profile s(t) is possible on this interval of time. This means that despite the fact that the solution of Eq. (8) depends on τ , this dependence remains the same during any waiting phase of any cycle, as long as τ is less than T_E+2T_R . Hence, the distribution density of the residence times in the state s=-1 does not change in time and the renewal theory [4,20] can be applied. Three different cases should be considered separately for $\tau \leq T_E+2T_R$.

Case 1: $\tau \in [0; T_R]$. The solution of Eq. (8) is then given by

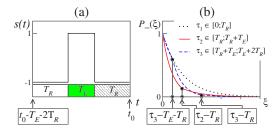


FIG. 4. (Color online) (a) Sketch of the profile of the two-state process $s(t) = \pm 1$ on the interval of time $[t_0 - T_E - 2T_R; t_0]$. (b) Three possible forms of the survival probability $P_-(\xi)$ in the waiting phase for $\tau < T_E + 2T_R$. Here, ξ is the duration of the waiting phase. $P_-(\xi)$ is shown for three different values τ_i of the delay time τ chosen arbitrarily from the intervals indicated in the legend. Filled circles on the ξ axis indicate points where the change of the slope of the corresponding $P_-(\xi)$ occurs.

$$P_{-}(\xi) = e^{-p\xi}, \quad \xi \in [0; \infty).$$
 (10)

This solution is shown by the dotted line in Fig. 4(b).

Case 2: $\tau \in [T_R; T_E + T_R]$. In this case the survival probability reads

$$P_{-}(\xi) = \begin{cases} e^{-(p+q)\xi}, & \xi \in [0; \tau - T_R], \\ e^{-p\xi} e^{-q(\tau - T_R)}, & \xi \in [\tau - T_R; \infty). \end{cases}$$
(11)

This solution is shown by the solid line in Fig. 4(b). Case 3: $\tau \in [T_E + T_R; T_E + 2T_R]$.

$$P_{-}(\xi) = e^{-p\xi} \begin{cases} 1, & \xi \in [0; \tau - T_E - T_R], \\ e^{-q\xi} e^{q(\tau - T_E - T_R)}, & \xi \in [\tau - T_E - T_R; \tau - T_R], \\ e^{-qT_E}, & \xi \in [\tau - T_R; \infty). \end{cases}$$
(12)

The dashed line in Fig. 4(b) corresponds to $P_{-}(\xi)$ in case 3. We can define the main period $T_{\rm main}$ as $T_{\rm main} = 2\pi/\Omega_{\rm max}$, where the frequency $\Omega_{\rm max}$ corresponds to the absolute maximum of the power spectrum of oscillations.

In Ref. [8] it has been shown numerically for the FitzHugh-Nagumo system that $T_{\rm main}$ depends almost piecewise linearly on τ . It was assumed there that the piecewise linearity in the main period vs delay time is a universal phenomenon that should occur in any system driven by noise including nonexcitable systems near Andronov-Hopf bifurcation. Here, a similar dependence of $T_{\rm main}$ on τ will be demonstrated for the system equation (2) analytically already for $\tau \in [0; T_E + 2T_R]$ and k < 0. From Eqs. (10)–(12) one can calculate the power spectrum of oscillations using the result of the renewal theory [4,20],

$$S(\omega) = \frac{2(\Delta x)^2}{\langle T_+ \rangle + \langle T_- \rangle} \frac{1}{\omega^2} \operatorname{Re} \left\{ \frac{\left[1 - \widetilde{\Psi}_-(i\omega)\right] \left[1 - \widetilde{\psi}_+(i\omega)\right]}{1 - \widetilde{\Psi}_-(i\omega)\widetilde{\psi}_+(i\omega)} \right\}, \tag{13}$$

where $\widetilde{\psi}_{+}(i\omega) = \exp(-i\omega T_{E})$ is the Laplace transform of ψ_{+} , $\widetilde{\Psi}_{-}(i\omega)$ is the Laplace transforms of Ψ_{-} from Eq. (7), $\Delta x = 2$ and $\langle T_{\pm} \rangle$ are the mean residence times in the states s = +1 and s = -1 of the two-state process, respectively. It is

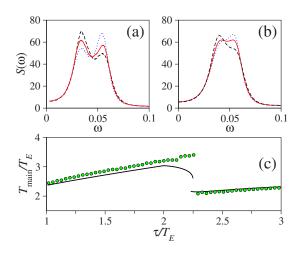


FIG. 5. (Color online) (a) Numerical power spectrum of the bistable system (2) for T=50 and three different τ close to τ_c . (b) Analytic power spectrum (13) for the same parameters as in (a) and T_E = T_R =55. (c) Numerical (circles) and analytic (line) main period of the oscillations.

clear that $\langle T_+ \rangle = T_E$ and $\langle T_- \rangle = T_R + \int_0^\infty \xi \psi_-(\xi) d\xi$.

We compare the analytic power spectrum (13) with the numerical power spectrum of the original bistable system (2) calculated with the parameters k=-0.05, T=50, D=0.053. Note that any switching event in the bistable system has a finite duration, i.e., the change of the coordinate x from (+1) to (-1) takes certain time. This leads to the increase of the duration of locking. Therefore to successfully match the parameters of the bistable system (2) to the parameters of the two-state model we set T_E = T_R =T+ Δ , where Δ is the time needed for the particle to descend from the maximum of the effective potential $U_{\rm eff}$ into its minimum. Δ is approximately set to 5 for |k|<0.05.

For the delay time τ between $2T_E$ and $3T_E$ the power spectrum has two maxima as shown in Fig. 5(a). For τ close to $2T_E$ the left maximum is higher than the right one [dashed line in Fig. 5(a)]. As τ increases from $2T_E$ to $3T_E$, the left maximum of $S(\omega)$ decreases and the right maximum increases. At critical τ_c the two maxima have equal heights as shown by the solid line in Fig. 5(a). For $\tau > \tau_c$ the right maximum is higher than the left one as shown by the dotted line in Fig. 5(a). Figure 5(b) shows the analytic power spectrum for the same parameters as in Fig. 5(a).

The main period increases almost linearly with τ for $\tau < \tau_c$, at $\tau = \tau_c$ the main period drops discontinuously and for $\tau > \tau_c$ it increases almost linearly again. This is shown in Fig. 5(c), where numerical results (circles) are compared with main period computed from the analytical power spectrum (13) (solid line).

B. Large delay times τ : Equilibrium RTDs

Now consider the case of the delay times larger than (T_E+2T_R) . This situation is qualitatively different from the one discussed in Sec. III A because for large delay times the history is variable. In other words, there are many possible profiles of the two-state system on the interval of time $[t_0]$

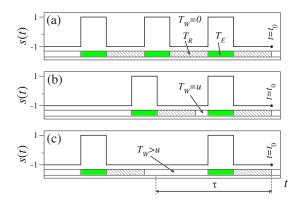


FIG. 6. (Color online) Schematic representation of the configuration of pulses in the case of $\tau \in [T_E + 2T_R; 2T_E + 3T_R]$. (a) The durations of the two latest waiting times are $T_W = 0$. (b) Durations of the latest waiting time is $T_W = u > 0$. (c) Durations of the latest waiting time is $T_W > u$.

 $-\tau$; t_0], where t_0 is the time moment when the current refractory phase has just finished. This situation is sketched in Fig. 6, where the time moment t_0 is indicated by the filled circle. The distribution of the residence times in the state s=-1 changes in time and now the results of the renewal theory cannot be applied directly. To overcome this problem we introduce the concept of equilibrium RTDs in the sense of the averaging over all possible histories. The equilibrium RTDs can be computed from the known instantaneous RTDs as shown below.

Consider a certain history $\tilde{s}(t)$ of the two-state stochastic process $s(t) = \pm 1$ on the interval of time $t \in [t_0 - \tau; t_0]$ inside which there are \tilde{N}_{τ} pulses. Note that inside an interval of length τ there can be not more than N_{τ} pulses, where N_{τ} = $[\tau/(T_E+T_R)]$ and $[\cdots]$ denotes an integer part of the number. Number these pulses with index j changing from 1 to $\tilde{N}_{\tau} \leq N_{\tau}$, with j=1 corresponding to the pulse closest to t_0 . Denote by u_i the duration of the waiting phase that precedes the jth pulse. If we know the number \tilde{N}_{τ} of pulses inside t $\in [t_0 - \tau; t_0]$ together with the ordered durations of the waiting phases u_j , $j=1,2,\dots,\tilde{N}_{\tau}$, we can unambiguously reconstruct $\tilde{s}(t)$. Let us consider all possible histories $\tilde{s}_k(t)$ and form an N_{τ} -dimensional "history" space $H_{\mathbf{u}}^{\tau}$ with vectors \mathbf{u}_{k} $=(u_{k,1},u_{k,2},\cdots,u_{k,N_{\tau}})$. If the number of pulses $\widetilde{N}_{\tau,k}$ for the given history $\tilde{s}_k(t)$ is less than the largest possible number N_{τ} , the redundant coordinates $u_{k,j}$, $j = N_{\tau,k} + 1, \dots, N_{\tau}$ are set to zero. Namely, if there is only one pulse on the given interval, we set $u_{k,1} = (\tau - T_E - T_R)$ and $u_{k,j} = 0, j = 2, ..., N_{\tau}$. If there are two pulses, we set $u_{k,2}=(\tau-2T_E-2T_R-u_{k,1})$ and $u_{k,j}=0$, j =3, \cdots , N_{τ} , etc. Therefore, the durations $u_{k,j}$ cannot be larger than $(\tau - T_E - T_R)$ by the way of construction. Obviously, $H_{\mathbf{u}}^{\tau}$ is in one-to-one correspondence with the set of all possible histories $\tilde{s}_k(t)$.

Next, split the space $H_{\mathbf{u}}^{\tau}$ into M equal N_{τ} -dimensional cells I_l , $l=1,2,\cdots,M$ of the form $I_l=[u_{l,1}\pm\eta/2]\times[u_{l,2}\pm\eta/2]\times\cdots\times[u_{l,N_{\tau}}\pm\eta/2]$, where η is the side length of each cell. $H_{\mathbf{u}}^{\tau}$ can be represented as a unity of N_{τ} -dimensional cells as follows:

$$H_{\mathbf{u}}^{\tau} = \{I_1 \cup I_2 \cup \cdots \cup I_M\}.$$

Consider one cell I_k . One can assume that all histories corresponding to the points inside this cell are approximately the same. Therefore, the waiting times that start immediately after t_0 and whose histories are approximately \mathbf{u}_k have approximately the same RTDs $\psi_-^{\mathbf{u}_k}$, which are solutions of Eq. (8).

Now consider some sufficiently long realization of the two-state stochastic process s(t). Suppose we are interested in some function f of the randomly changing waiting time ξ , and we wish to compute an average of f. There are two kinds of averages: over time and over the ensemble of realizations, which coincide if the underlying process is ergodic. The time average $\langle f \rangle$ can be calculated from a single realization as follows:

$$\langle f \rangle = \lim_{N \to \infty} \frac{1}{N} \sum_{i=1}^{N} f(\xi_i), \tag{14}$$

where N stands for the number of pulses in the realization. In a sufficiently long realization one can find a large number n_k of histories from the cell I_k . The waiting times occurring after these histories have (approximately) the same distribution $\psi_-^{n_k}$. In an infinitely long realization when $N \to \infty$, one can find *all* histories from the same cell, i.e., $n_k \to \infty$, with every history being found an infinite number of times. We now regroup the summands in Eq. (14) by collecting inside each bracket number k the values of k of the waiting times k occurring after the history from the same cell centered at k. We denote the respective values of k as k and k are the history from the same cell centered at k and

$$\langle f \rangle = \lim_{N \to \infty} \left(\frac{n_1}{N} \frac{1}{n_1} \left[f_1^{\mathbf{u}_1} + f_2^{\mathbf{u}_1} + \dots + f_{n_1}^{\mathbf{u}_1} \right] + \frac{n_2}{N} \frac{1}{n_2} \left[f_1^{\mathbf{u}_2} + f_2^{\mathbf{u}_2} + \dots + f_{n_2}^{\mathbf{u}_2} \right] + \dots + \frac{n_M}{N} \frac{1}{n_M} \left[f_1^{\mathbf{u}_M} + f_2^{\mathbf{u}_M} + \dots + f_{n_M}^{\mathbf{u}_M} \right] \right).$$
(15)

We introduce the equilibrium RTD $\psi_{-}^{eq}(\xi)$ in the sense that $\psi_{-}^{eq}(\xi)d\xi$ gives the probability for the waiting time to have the duration in the interval $[\xi; \xi+d\xi]$. The equilibrium RTD does not depend on history, i.e., $\psi_{-}^{eq}(\xi)d\xi$ is understood as the number N_{ξ} of the waiting times with the duration T_{W} that fall within the interval $[\xi; \xi+d\xi]$ divided by the total number N of pulses in any given realization of the stochastic process, in the limit as $N \rightarrow \infty$,

$$\psi_{-}^{\text{eq}}(\xi)d\xi = \lim_{N \to \infty} \frac{N_{\xi}}{N}.$$
 (16)

Clearly, the factors n_k/N in Eq. (15) are the probabilities of the history to be in the cell I_k . Denote these probabilities by $\mathcal{P}(\mathbf{u}_k)(d\eta)^{N_\tau}$, where $\mathcal{P}(\mathbf{u})$ is the corresponding distribution density. From the general considerations it is clear that $\mathcal{P}(\mathbf{u})$ depends solely on ψ_{-}^{eq} . The particular dependence of $\mathcal{P}(\mathbf{u}_k)$

on ψ_{-}^{eq} is to be determined separately for any given history \mathbf{u}_{k} .

The terms

$$\frac{1}{n_k} \underbrace{\left[f_1^{u_k} + f_2^{u_k} + \dots + f_{n_k}^{u_k} \right]}_{n_k}$$

in Eq. (15) are the time averages of $f(\xi)$, for ξ with the history determined by the cell centered around \mathbf{u}_k and with distribution density $\psi_-^{n_k}$. Now, since the process is assumed to be ergodic, a time average is equal to the ensemble average. The latter can be calculated with the knowledge of the RTD $\psi_-^{n_k}$ for the given history as $\int_0^\infty \psi_-^{n_k}(\xi) f(\xi) d\xi$.

Finally, in the limit $M \rightarrow \infty$ and $\eta \rightarrow 0$, Eq. (15) can be rewritten in the form

$$\langle f \rangle = \int_0^\infty \psi_-^{\text{eq}}(\xi) f(\xi) d\xi$$
$$= \int_{u \in H_{\mathbf{u}}^{\tau}} \mathcal{P}(\mathbf{u}) du_1 du_2 \cdots du_{N_{\tau}} \int_0^\infty \psi_-^{\mathbf{u}}(\xi) f(\xi) d\xi. \quad (17)$$

Regroup terms in Eq. (17) as follows:

$$0 = \int_0^\infty f(\xi)d\xi \left\{ -\psi_-^{\text{eq}}(\xi) + \int_{u \in H_{\mathbf{u}}^T} \mathcal{P}(\mathbf{u})\psi_-^{\mathbf{u}}(\xi)du_1du_2 \cdots du_{N_\tau} \right\}.$$

Note that Eq. (17) holds for an arbitrary integrable function f, which means that the expression in brackets is equal to zero. We therefore arrive at the following integral equation for the unknown equilibrium RTD,

$$\psi_{-}^{\text{eq}}(\xi) = \int_{\mathbf{u} \in H_{\mathbf{u}}^{\tau}} \mathcal{P}(\mathbf{u}) \psi_{-}^{\mathbf{u}}(\xi) du_1 du_2 \cdots du_{N_{\tau}}. \tag{18}$$

Equation (18) allows us to treat the renewal processes with nonidentically distributed waiting times. It generalizes the results of the renewal theory [4].

C. Equilibrium RTD for $\tau \in [T_E + 2T_R; 2T_E + 2T_R]$

We now use the derived equation (18) to calculate the equilibrium RTD $\psi_{-}^{\rm eq}$ in the case of $\tau \in [T_E + 2T_R; 2T_E + 2T_R]$. The reason for choosing this interval for τ instead of the whole interval $[T_E + 2T_R; 2T_E + 3T_R]$, where the history space $H_{\bf u}^{\tau}$ is one-dimensional is the following. If $\tau \in [T_E + 2T_R; 2T_E + 2T_R]$ the history can contain only one complete pulse and one half-complete pulse (see Fig. 6), depending on the duration of the previous waiting phase T_W . Therefore, this situation is simpler than the case of $\tau \in [T_E + 2T_R; 2T_E + 3T_R]$, where the history can contain up to two complete pulses. Denote the duration of the previous waiting phase by u and consider two possible cases.

Case 1. For $u \in [\tau - T_E - 2T_R; \infty)$ the history contains one complete pulse. In this case the solution of Eq. (8) is given by

$$P_{-}^{0}(\xi) = e^{-p\xi} \begin{cases} 1, & \xi \in [0; \ \tau - T_{E} - T_{R}], \\ e^{-q\xi + q(\tau - T_{E} - T_{R})}, & \xi \in [\tau - T_{E} - T_{R}; \ \tau - T_{R}] \\ e^{-qT_{E}}, & \xi \in [\tau - T_{R}; \ \infty). \end{cases}$$
(19)

Case 2. For $u \in [0; \tau - T_E - 2T_R]$ the history contains one complete pulse and one half-complete pulse. In this case the solution of Eq. (8) is

$$P_{-}^{u}(\xi) = e^{-p\xi} \times \begin{cases} e^{-q\xi}, & \xi \in [0; \ \tau - T_{E} - 2T_{R} - u], \\ e^{-q(\tau - T_{E} - 2T_{R} - u)}, & \xi \in [\tau - T_{E} - 2T_{R} - u; \ \tau - T_{E} - T_{R}], \\ e^{-q\xi + q(T_{R} + u)}, & \xi \in [\tau - T_{E} - T_{R}; \ \tau - T_{R}], \\ e^{-q(\tau - 2T_{R} - u)}, & \xi \in [\tau - T_{R}; \ \infty). \end{cases}$$
(20)

The connection between the probability density $\mathcal{P}(u)$ and ψ_{-}^{eq} is straightforward

$$\mathcal{P}(u) = \psi_{-}^{\text{eq}}(u). \tag{21}$$

Because u is constrained to the interval $[0; \tau - T_E - 2T_R]$, the relation, Eq. (18), is an equation for the unknown $\psi_-^{\text{eq}}(\xi)$ only on the same interval of $\xi \in [0; \tau - T_E - 2T_R]$. For $\xi > (\tau - T_E - 2T_R)$, Eq. (18) is no longer an equation, as will be shown below

For $\xi \in [0; \tau - T_E - 2T_R]$ we divide the integration interval U according to the restrictions on ξ and u as in Eq. (21) and

rewrite Eq. (18) as follows:

$$\begin{split} \psi_{-}^{\text{eq}}(\xi) &= p e^{-p\xi} \Biggl(1 - \int_{0}^{\tau - T_{E} - 2T_{R}} \psi_{-}^{\text{eq}}(u) du \Biggr) \\ &+ (p + q) e^{-(p + q)\xi} \int_{0}^{\tau - T_{E} - 2T_{R} - \xi} \psi_{-}^{\text{eq}}(u) du \\ &+ p e^{-p\xi} e^{-q(\tau - T_{E} - 2T_{R})} \int_{\tau - T_{E} - 2T_{R} - \xi}^{\tau - T_{E} - 2T_{R} - \xi} e^{qu} \psi_{-}^{\text{eq}}(u) du \,. \end{split}$$

We look for the solution of Eq. (22) in the form

$$\psi_{-}^{\text{eq}}(t) = Ae^{-(p+q)\xi},$$
 (23)

where *A* is some unknown constant. Plugging Eq. (23) into Eq. (22) and comparing the coefficients of the exponents $e^{-p\xi}$ and $e^{-(p+q)\xi}$ yields the constant *A*,

$$A = \frac{p(p+q)}{p + qe^{-(p+q)(\tau - T_E - 2T_R)}}. (24)$$

On the interval $\xi \in [\tau - T_E - 2T_R; \tau - T_E - T_R]$ Eq. (18) becomes

$$\psi_{-}^{\text{eq}}(\xi) = pe^{-p\xi} \left(1 - \int_{0}^{\tau - T_{E} - 2T_{R}} \psi_{-}^{\text{eq}}(u) du \right)$$

$$+ pe^{-p\xi} e^{-q(\tau - T_{E} - 2T_{R})} \int_{0}^{\tau - T_{E} - 2T_{R}} e^{qu} \psi_{-}^{\text{eq}}(u) du,$$
(25)

which is no longer an equation, because the right-hand side of Eq. (25) depends on the known ψ_-^{eq} on the interval $\xi \in [0; \tau - T_E - 2T_R]$. Similarly, the equilibrium RTD ψ_-^{eq} on the interval $\xi \in [\tau - T_E - T_R; \tau - T_R]$ is determined as

$$\psi_{-}^{\text{eq}}(\xi) = (p+q)e^{-(p+q)\xi}e^{q(\tau-T_E-T_R)}
\times \left(1 - \int_0^{\tau-T_E-2T_R} \psi_{-}^{\text{eq}}(u)du\right)
+ (p+q)e^{-(p+q)\xi}e^{qT_R} \int_0^{\tau-T_E-2T_R} e^{qu}\psi_{-}^{\text{eq}}(u)du.$$
(26)

Finally, the equilibrium RTD ψ_{-}^{eq} on the interval $\xi \in [\tau - T_R; \infty)$ reads

$$\psi_{-}^{\text{eq}}(\xi) = pe^{-p\xi}e^{-qT_{E}} \left(1 - \int_{0}^{\tau - T_{E} - 2T_{R}} \psi_{-}^{\text{eq}}(u)du \right) + pe^{-p\xi}e^{-q(\tau - 2T_{R})} \int_{0}^{\tau - T_{E} - 2T_{R}} e^{qu} \psi_{-}^{\text{eq}}(u)du. \quad (27)$$

Taking the integrals in Eqs. (25)–(27) with ψ_{-}^{eq} on the interval $\xi \in [0; \tau - T_E - 2T_R]$ from Eq. (23), one obtains the equilibrium RTD on the whole interval

$$\psi_{-}^{\text{eq}}(\xi) = A \begin{cases}
e^{-(p+q)\xi}, & \xi \in [0; \tau - T_E - 2T_R], \\
e^{-q(\tau - T_E - 2T_R) - p\xi}, & \xi \in [\tau - T_E - 2T_R; \tau - T_E - T_R], \\
\frac{(p+q)}{p} e^{-(p+q)\xi} e^{qT_R}, & \xi \in [\tau - T_E - T_R; \tau - T_R], \\
e^{-q(\tau - 2T_R) - p\xi}, & \xi \in [\tau - T_R; \infty).
\end{cases} (28)$$

The knowledge of the RTD allows us to compute the average of any given function g(s) of the stochastic variable s(t),

$$\langle g \rangle = \frac{g(-1)\langle T_{-} \rangle + g(+1)T_{E}}{T_{F} + \langle T_{-} \rangle}, \tag{29}$$

where $\langle T_- \rangle = T_R + \int_0^\infty u \psi_-^{\text{eq}}(u) du$ for $\tau \in [T_E + 2T_R; 2T_E + 2T_R]$. For the values of the delay time $\tau < T_E + 2T_R$ the equilibrium RTD must be replaced by the corresponding $\psi_-(u)$ from Eq. (8) with $P_-(u)$ from Eqs. (11) or (12).

We present here the results of the comparison of the analytic formula, Eq. (29), with the simulation of the original bistable system (2). Figures 7(a) and 7(b) show the average $\langle x \rangle$ vs delay time τ for D=0.043, k=0.05 and D=0.053, k=-0.01, respectively. Analytic result (29) (lines) is compared with $\langle x \rangle$ calculated numerically (circles). As in Fig. 5 the durations of the excited and the refractory phases are T_E = T_R =55.

For positive feedback strength [Fig. 7(a)] the average $\langle x \rangle$ increases with the delay until $\tau = 2T_E$, then it decreases when τ is in the interval $\tau \in [2T_E; 3T_E]$ and after that it increases again when $\tau \in [3T_E; 4T_E]$. The behavior of $\langle x \rangle$ on τ is exactly the opposite for negative feedback strength k=-0.01 as

it is shown in Fig. 7(b). We do not plot the variance $\langle x^2 \rangle$ because it is independent of delay: according to Eq. (29), for the stochastic process $s = \pm 1$ the variance is constant ($\langle s^2 \rangle = 1$).

Note that the nonmonotonous dependence of the average $\langle x \rangle$ on the delay time is qualitatively similar to the dependence of the variance $\langle x^2 \rangle$ on delay time which was derived in Ref. [21] for the van der Pol oscillator near the Andronov-Hopf bifurcation. This similarity shows again that the properties of the noise-induced oscillations in the excitable sys-

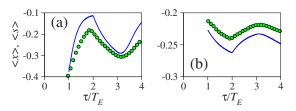


FIG. 7. (Color online) Comparison of the average x vs delay time calculated numerically (circles) with the average $\langle s \rangle$ of the two-state process (29) (lines). Parameters are (a) D=0.043, T=50, $T_E=T_R=55$, k=0.05 and (b) D=0.053, T=50, $T_E=T_R=55$, k=-0.01.

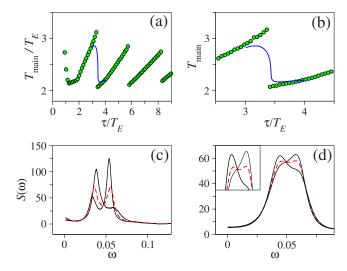


FIG. 8. (Color online) (a) The circles represent the main period $T_{\rm main}$ of the oscillations in units of $T_E = T_R = 55$ calculated numerically, the solid line shows the corresponding analytic counterpart computed from Eqs. (13) and (28). (b) Zoom of the region in (a) near $\tau = 3T_E$. (c) Numerical power spectrum for τ near the critical τ_c . Dashed line shows power spectrum at $\tau = \tau_c$. (d) Analytic power spectrum for $T_E = T_R = 49$. The inset in (d) shows a zoom of the area around the maximum of $S(\omega)$.

tems are closely related to the properties of nonexcitable noise-driven systems near bifurcations [8].

IV. POWER SPECTRUM IN THE MEAN-FIELD APPROXIMATION

The knowledge of the equilibrium RTD does not allow us to compute the power spectrum of noise-induced oscillations. According to Ref. [7] the power spectrum is given by the Fourier transform of the average product of probability currents $\langle j(t)j(t')\rangle$, where the current j(t) is a sum of δ -like pulses occurring at the moments t_n of switching: $j=2\Sigma_n(-1)^n\delta(t-t_n)$. Any switching time t_n is the time moment, when the transition from the state s=-1 to the state s=+1 or backwards occurs. The amplitude of the current j is given by +2 in case when s=-1 changes to s=+1 and by -2 otherwise. Unfortunately, the average $\langle j(t)j(t')\rangle$ cannot be represented in the form (15) and therefore the correlation function cannot be written in terms of the equilibrium ψ_{-1}^{eq} .

One possible approximation for the power spectrum is an analog of the mean-field approximation, when the known result for the power spectrum from the renewal theory [4,20] is calculated with the equilibrium ψ_{-}^{eq} obtained above.

We compare this version of the mean-field approach with the simulation of the bistable system (2) in Fig. 8. For D = 0.043, k = 0.05 and the rest of the parameters as in Fig. 7, in Figs. 8(a) and 8(b) the main period $T_{\rm main}$ of the noise-induced oscillations is given as a function of delay time τ . Both $T_{\rm main}$ and τ are shown in the units of $T_E = T_R$. Circles in Figs. 8(a) and 8(b) correspond to the results of simulation, lines show the main period computed from the analytic expression for the power spectrum (13) with the equilibrium

RTD (28). For simplicity we show only the part of the analytical curve for $\tau \in [3T_E; 4T_E]$. As we see, the mean-field analytics predicts correctly the critical delay time where the first branch switching occurs.

In Fig. 8(c) the numerical power spectrum is shown for three different values of τ close to its critical value τ_c . Figure 8(d) shows the analytic power spectrum for the same parameters as in Fig. 8(c) and $T_E = T_R = 49$.

V. INCOHERENCE MAXIMIZATION DUE TO DELAY

A. Bistable excitable system

The RTDs ψ_{-} derived in Sec. III, along with the equilibrium RTD (28), allow us to compute the coefficient of variation R [1], which serves as a measure of coherence of spiking

$$R = \frac{\sqrt{\langle T_{-}^2 \rangle - \langle T_{-} \rangle^2}}{T_F + \langle T_{-} \rangle}.$$
 (30)

The smaller the coefficient of variation R the more regular is the spiking. The effect of coherence resonance [22] manifests itself in the appearance of a minimum in the dependence of the coefficient of variation on the noise strength D. The opposite effect to coherence resonance is regarded as incoherence maximization [1]. It appears when there is a local maximum in the dependence of R on D.

Here we show that in the bistable system presented in Sec. II, delay induced incoherence maximization can be observed for positive feedback strength k. In terms of the twostate model with history dependent transition rates, Table I, positive feedback strength k corresponds to the inequality p +q > p, implying that the probability of transition from the state s=-1 to the state s=+1 is larger if $s_{\tau}=+1$. Since the durations of the excited T_E and the refractory T_R states are fixed, the dependence on noise is realized through the duration of the waiting phase T_W only. Obviously, in the limit of large noise strength $(D \rightarrow \infty)$, the duration T_W becomes vanishingly small, leading to almost regular spiking with the mean interspike interval given by T_E+T_R . For vanishing noise (D=0) the transition probability is negligibly small resulting again in the regular spike train. Therefore, for some finite noise strength we can expect to observe a local maximum of the spike incoherence.

Using the Kramers relation (6) [18,19] between the transition rates p and q and the noise strength D, we plot in Fig. 9 the coefficient of variation R as function of D. In Fig. 9(a) R is shown for positive feedback strength k=0.05, T_E = T_R =55, and different delay times τ as indicated in the legend. With no feedback ($\tau \le T_E$) the coefficient of variation decreases monotonically with D. However, at any τ larger than, and close to, T_E , a maximum appears at a certain noise strength D. This is evidence of incoherence maximization induced by the delay. If the feedback strength is negative, k=-0.05 [Fig. 9(b)], the incoherence maximization is absent at least for the delay times $\tau \le 4T_E$.

B. Comparison with the FitzHugh-Nagumo system

To demonstrate incoherence maximization predicted in Sec. V A, we use the FitzHugh-Nagumo system with nonlin-

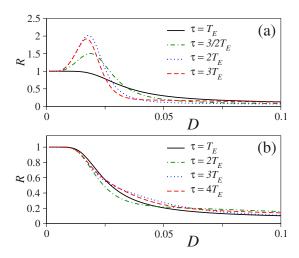


FIG. 9. (Color online) (a) Coefficient of variation R vs noise strength D for k=0.05 and T_E = T_R =55. (b) The same as in (a) for the negative feedback strength k=-0.05.

ear delayed feedback introduced through the activator x into the equation for inhibitor y,

$$\epsilon \dot{x} = x - \frac{x^3}{3} - y,$$

$$\dot{y} = x + a - K[x_{\tau} - x]^2 + D\zeta(t),$$
 (31)

where $x_{\tau} = x(t - \tau)$, $\epsilon = 0.01$, and a = 1.1.

The form of the feedback term in Eqs. (31) was chosen in such a way that the noise-induced dynamics in the waiting phase is similar to that of the two-state model introduced in Sec. III.

It should be emphasized that the FitzHugh-Nagumo system with *linear* feedback cannot be approximated by a two-state model with the transition rates as in Table I. To see why this is so, consider a typical pulse train given by variable x as function of time. This is shown in Fig. 4(a). Assume that the system is currently in the waiting phase, $x(t) \in$ waiting phase, and assume further that τ seconds ago the system was in the refractory phase, $x(t-\tau) \in$ refractory phase. Since the value of the state variable in the waiting phase is different from that in the refractory phase, unlike in the two-state model, we conclude that the feedback term which is given by $k[x(t)-x(t-\tau)]$ is not zero. However, this term becomes zero (up to fluctuations whose order is given by the noise strength D) if τ seconds ago the system was in the waiting phase.

This means that the current transition rate λ from the waiting phase to the excited phase changes depending on whether τ seconds ago the system was in the waiting or in the refractory phase. This contradicts the assumption that the transition rate depends on τ according to Table I.

The comparison between the FitzHugh-Nagumo model and the two-state model is possible only in the case of non-linear feedback, e.g., such as in Eqs. (31) and strong time scale separation, i.e., $\epsilon \leq 1$, when the concept of the transition rates can be applied. The nonlinear feedback term ensures that during the waiting phase $[x(t) \approx -1.1]$ the value of

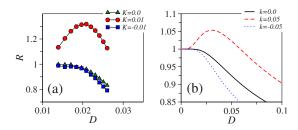


FIG. 10. (Color online) (a) Coefficient of variation R vs noise strength D for the FitzHugh-Nagumo system, Eqs. (31), with fixed delay time τ =4 and the feedback strength K as indicated in the legend. (b) Coefficient of variation R vs noise strength D for the two-state model with the parameters T_R =2.5, τ =4, and different feedback strength R given in the legend.

 $[x_{\tau}-x]^2$ becomes significant only if τ second ago the system was in the excited phase, i.e., if $x_{\tau} \approx 2$. On the other hand, if x_{τ} belongs to the refractory phase, the term $[x_{\tau}-x]^2$ is negligibly small, so that the transition rate from the nonexcited to the excited state is the same as in the original system without the feedback. Consequently, the transition rate is modified by the feedback only if x_{τ} belongs to the excited state.

It is easy to see from Eqs. (31) that negative values of the feedback strength K effectively decrease the transition rate, whereas the positive values of K increase the probability of transition. Therefore, based on the predictions made in Sec. V A, we conclude that delay-induced incoherence maximization should be observed for positive K.

This result is confirmed by numerically computing the coefficient of variation R vs noise strength D for the FitzHugh-Nagumo system, Eqs. (31), at fixed delay time τ =4, as shown in Fig. 10(a). For negative K and K=0, R decreases monotonically with D, however for positive K the coefficient of variation reaches a local maximum confirming incoherence maximization at certain noise strength D.

To match the parameters of the two-state model with those of the FitzHugh-Nagumo system, Eqs. (31), we set T_R =2.5, T_E =0.5, τ =4, and plot R vs noise strength D in Fig. 10(b) at different values of the feedback strength k (see legend). We see that the behavior of the coefficient of variation R on D in the two-state model qualitatively coincides with the numerical results obtained for the FitzHugh-Nagumo system Fig. 10(a).

VI. CONCLUSION

To conclude, we presented a two-state model of an excitable system with time-delayed feedback. In this model the state variable s takes only two values $s = \pm 1$ and the transition probability from one state into the other depends on the history of the process in a given way. To compare the results derived for the two-state model with the properties of a real excitable system we consider a bistable system (2) with the effective potential (4) which contains two delay times. One of them is fixed and is used to model the excitability and the other one is assigned to the delay time of the controlling feedback force.

Assuming that the durations of the excited phase and the refractory phase are noise independent, we conclude that the

only unknown and in general history-dependent quantity is the residence time density (RTD) $\psi_{-}(\xi)$ of the waiting phase. We show that for delay times less than the sum of the duration of the excited phase T_E and two durations of the refractory phase $2T_R$, the history is nonvariable, i.e., there is only one possible profile of the two-state process on the interval of time $[t_0 - \tau; t_0]$, where t_0 is the time moment when the latest refractory phase has just finished. Therefore, all the waiting times are identically distributed and the renewal theory can be applied. In this case $\psi_{-}(\xi)$ as function of the delay time τ is computed straight forwardly and represented by Eq. (8) with P_{\perp} determined by Eqs. (11) and (12). We use the results of the renewal theory [4,20] to obtain the power spectrum of the stochastic process s(t). The main period of the noise-induced oscillations calculated from the analytically known power spectrum shows piecewise linear dependence on the delay time. This analytical result confirms the similar finding obtained numerically in Refs. [8,10] for the FitzHugh-Nagumo system in the excitable regime.

For the delay times larger than (T_E+2T_R) the history becomes variable and the distribution density of the waiting times is no longer time independent. To handle renewal processes with history-dependent RTDs the equilibrium RTD $\psi_{-}^{eq}(\xi)$, Eq. (16), in the sense of the averaging over all possible histories is introduced. The integral, Eq. (18), for $\psi_{-}^{eq}(\xi)$ is derived for an arbitrary delay time τ . The solution of this equation is given for $\tau \in [T_E+2T_R; 2T_E+2T_R]$ by Eq. (28). The knowledge of $\psi_{-}^{eq}(\xi)$ allows us to calculate the average

of any given function f(s) of the stochastic process s be means of Eq. (29). This is an exact formula which is valid for any renewal process with the dependence on history. Unfortunately, the results of the renewal theory cannot be used directly to compute, e.g., the power spectrum of the noise-induced oscillations. However, an analog of the mean-field approximation is introduced when the expression for the power spectrum from the renewal theory is calculated with the equilibrium $\psi_-^{eq}(\xi)$.

The analytic results were compared with the results of numerical simulation of the bistable system (2). It is shown that the mean-field power spectrum predicts correctly the critical delay time when the first branch switching occurs in the dependence of the main period vs delay time [Figs. 8(a) and 8(b)].

Finally, we demonstrated incoherence maximization for positive feedback strength due to delay [Fig. 9(a)]. Hereby, the degree of incoherence measured by the coefficient of variation (30) is shown to posses a local maximum for increasing noise strength D and fixed delay time. For negative feedback strength incoherence maximization is not observed up to the delay times of $(2T_E + 2T_R)$.

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