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Multidimensional characterization of stochastic dynamical systems based on multiple perturbations and measurements

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Generalized nonlinear response theory is presented for stochastic dynamical systems. Experiments in which multiple measurements of dynamical quantities are used along with multiple perturbations of parameters of dynamical systems are described by generalized response functions (GRFs). These constitute a new type of multidimensional measures of stochastic dynamics either in the time or the frequency domains. Closed expressions for GRFs in stochastic dynamical systems are derived and compared with numerical non-equilibrium simulations. Several types of perturbations are considered: impulsive and periodic perturbations of temperature and impulsive perturbations of coordinates. The present approach can be used to study various types of stochastic processes ranging from single-molecule conformational dynamics to chemical kinetics of finite-size reactors such as biocells. © 2015 AIP Publishing LLC. [http://dx.doi.org/10.1063/1.4917527]

I. INTRODUCTION

Multidimensional spectroscopic techniques span a broad range of frequencies from NMR to the UV and use numerous types of pulse sequences and detection modes.¹ They all share one common thread: the molecule is excited by several pulses followed by a single detection event. nD techniques involve *n* perturbations followed by a single detection event and are represented by causal response functions. Thinking more broadly, one can envision other types of signals that involve n perturbations and m measurements. These measure both response and spontaneous fluctuations and are not causal. In the linear regime, there are two possible signals (excitation followed by measurement or two measurements) but they are connected by the fluctuation dissipation $(FD)^{2-4}$ relation and carry the same information.⁵ This is not the case for the nonlinear response since there is no universal FD theorem. The basic ideas are not limited to optical pulses and hold for any type of perturbation (electrical, mechanical, chemical concentrations, etc.) In this paper, we discuss several examples of such new types of multidimensional signals.

The idea of using multiple perturbations and measurements has been discussed previously. Reference 6 considered generalized response functions (GRFs) corresponding to a series of perturbations of open quantum system followed by a number of measurements. That formalism allowed the calculation of van der Waals forces between molecules subject to time-dependent coupling. GRFs corresponding to impulsive perturbations of spin systems were considered in Ref. 7 and used the derived generalized multi-point correlation functions to study stochastic dynamics of disordered spin models. The idea of mixing multiple perturbations with measurements was extended in Ref. 8, to arbitrary type of time-dependent perturbation, and applied to examine classical dynamics of nonlinear Hamiltonian systems. That formalism was further extended to non-Hamiltonian systems, such as chemical reactions, and was used to study nonlinear kinetics and connectivity of species in chemical systems.⁹

In the present paper, we introduce a generalized response theory of stochastic dynamical systems. The formalism is based on a perturbative expansion of the Fokker-Planck equation around steady state. The Fokker-Planck approach can be used for equilibrium¹⁰ as well as non-equilibrium¹¹ systems; thus, the developed formalism is applicable to systems in thermal equilibrium as well as in non-equilibrium steady-state.

The paper is organized as follows. Expressions for nonlinear response functions of a general stochastic process are derived in Secs. II and III. Section IV considers the special case of impulsive perturbations. The concept of GRF is introduced in Sec. V. Section VI introduces generalized susceptibilities for a harmonic perturbation. GRFs of exactly solvable Ornstein-Uhlenbeck process (OUP) are calculated in Sec. VII and compared with analytical expressions. 2D GRFs corresponding to perturbations of temperature in bistable and nonlinear stochastic systems at thermal equilibrium are calculated in Sec. VIII, and the quantitative information contained in GRFs is analyzed. GRFs of a stochastic chemical network are calculated in Sec. IX and explored for signatures of chemical connectivity. We conclude with a discussion in Sec. X.

II. THE FOKKER-PLANCK EQUATION

A general time-dependent stochastic process in *N*-dimensions, $\mathbf{x}(t)$, can be described either by stochastic differential equations or, equivalently, its time-dependent probability density $W(\mathbf{x},t)$ which satisfies a Fokker-Planck equation¹⁰

$$\frac{\partial W(\mathbf{x},t)}{\partial t} = \hat{L}(\mathbf{x},t)W(\mathbf{x},t),\tag{1}$$

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where $\hat{L}(\mathbf{x},t)$ is the Fokker-Planck operator,

$$\hat{L}(\mathbf{x},t) = -\frac{\partial}{\partial x_i} D_i(\mathbf{x},t) + \frac{\partial^2}{\partial x_i \partial x_j} D_{ij}(\mathbf{x},t), \qquad (2)$$

where $D_i(\mathbf{x}, t)$ is a vector of drift coefficients and $D_{ij}(\mathbf{x}, t)$ is a matrix of diffusion coefficients.

Since the conditional probability $P(\mathbf{x},t|\mathbf{x}',t')$ is the probability distribution $W(\mathbf{x},t)$ for particular initial condition $W(\mathbf{x},0) = \delta(\mathbf{x} - \mathbf{x}')$, then it should also obey Eq. (1),

$$\frac{\partial P(\mathbf{x},t|\mathbf{x}',t')}{\partial t} = \hat{L}(\mathbf{x},t)P(\mathbf{x},t|\mathbf{x}',t').$$
(3)

A formal solution of Eq. (3) with the initial conditions $P(\mathbf{x},t|\mathbf{x}',t) = \delta(\mathbf{x} - \mathbf{x}')$ then reads¹⁰

$$P(\mathbf{x},t|\mathbf{x}',t') = T_{+} \exp\left[\int_{t'}^{t} \hat{L}(\mathbf{x},\tau)d\tau\right] \delta(\mathbf{x}-\mathbf{x}'), \qquad (4)$$

where T_+ is the time ordering operation, which orders time to increase from right to left in the products of $\hat{L}(\mathbf{x},\tau)$ of Taylor-expansion of the exponential.

As can be seen from Eqs. (3) and (4), the differential operator $\hat{L}(\mathbf{x}, t)$ acts on \mathbf{x} , which is the value of the stochastic variable at a later time *t*, rather than on \mathbf{x}' at earlier time *t'*. This representation is therefore known as the forward Kramers-Moyal expansion. In a similar way, it is possible to derive a backward representation of Eq. (3), with the operator \hat{L}^+ acting on \mathbf{x}' rather than \mathbf{x} ,

$$\frac{\partial P(\mathbf{x},t|\mathbf{x}',t')}{\partial t} = \hat{L}^{+}(\mathbf{x}',t')P(\mathbf{x},t|\mathbf{x}',t').$$
(5)

It has been shown¹⁰ that \hat{L}^+ is actually the adjoint of the Fokker-Planck operator \hat{L} and has the following form:

$$\hat{L}^{+}(\mathbf{x}',t') = -D_{i}(\mathbf{x}',t')\frac{\partial}{\partial x_{i}'} + D_{ij}(\mathbf{x}',t')\frac{\partial^{2}}{\partial x_{i}'\partial x_{j}'}.$$
 (6)

The solution of Eq. (6) will then be formally similar to Eq. (4),

$$P(\mathbf{x},t|\mathbf{x}',t') = T_{-}\exp\left[\int_{t'}^{t} \hat{L}^{+}(\mathbf{x}',\tau')d\tau'\right]\delta(\mathbf{x}-\mathbf{x}'),$$
(7)

the difference is that the time ordering T_{-} is now reversed and that the operator \hat{L}^{+} acts on \mathbf{x}' instead of \mathbf{x} . Dual representation (4) and (7) of the same quantity $P(\mathbf{x}, t | \mathbf{x}', t')$ will be useful in the following derivations since it allows to switch between the old and new variables \mathbf{x}' and \mathbf{x} .

It is also useful to define the propagators $\hat{U}(x,t;t')$ and $\hat{U}^+(x',t;t')$,

$$\hat{U}(\mathbf{x};t,t') = T_{+} \exp\left[\int_{t'}^{t} \hat{L}(\mathbf{x},\tau) d\tau\right],$$
(8)

$$\hat{U}^{+}(\mathbf{x}';t,t') = T_{-} \exp\left[\int_{t'}^{t} \hat{L}^{+}(\mathbf{x}',\tau')d\tau'\right].$$
(9)

The operator \hat{U}^+ is the adjoint of \hat{U} .¹⁰ Yet, one should keep in mind that unlike ordinary quantum or classical dynamics, the operator \hat{U} is non-Hermitian and thus $\hat{U}^+ \neq \hat{U}$.

Equations (4) and (7) can be recast in terms of these propagators as

$$P(\mathbf{x},t|\mathbf{x}',t') = \hat{U}(\mathbf{x};t,t')\delta(\mathbf{x}-\mathbf{x}'), \tag{10}$$

$$P(\mathbf{x},t|\mathbf{x}',t') = \hat{U}^{+}(\mathbf{x}';t,t')\delta(\mathbf{x}-\mathbf{x}').$$
(11)

If we measure a dynamical variable A, we can calculate the outcome either by propagating P forward with \hat{U} or by propagating A backward with \hat{U}^+ . So, physically we are moving forward in either case, but calculate a different intermediate object.

A. Properties of the operator \hat{U}

The operator $\hat{U}(\mathbf{x}; t, t')$ satisfies the following property:

$$\hat{U}(\mathbf{x}; t_3, t_1) = \hat{U}(\mathbf{x}; t_3, t_2)\hat{U}(\mathbf{x}; t_2, t_1).$$
(12)

Indeed, since

$$\int P(\mathbf{x}_3, t_3 | \mathbf{x}_2, t_2) P(\mathbf{x}_2, t_2 | \mathbf{x}_1, t_1) d\mathbf{x}_2 = P(\mathbf{x}_3, t_3 | \mathbf{x}_1, t_1),$$
(13)

then, substituting Eqs. (10) and (11) into Eq. (13), one gets

$$\int \left[\hat{U}(\mathbf{x}_3; t_3, t_2) \delta(\mathbf{x}_3 - \mathbf{x}_2) \right] \left[\hat{U}^+(\mathbf{x}_1; t_2, t_1) \delta(\mathbf{x}_2 - \mathbf{x}_1) \right] d\mathbf{x}_2$$

= $\hat{U}(\mathbf{x}_3; t_3, t_1) \delta(\mathbf{x}_3 - \mathbf{x}_1).$ (14)

Since in Eq. (14) $\hat{U}(\mathbf{x}_3; t_3, t_2)$ acts only on \mathbf{x}_3 and $\hat{U}^+(\mathbf{x}_1; t_2, t_1)$ acts only on \mathbf{x}_1 , one can rearrange the order of operators,

$$\hat{U}(\mathbf{x}_3; t_3, t_2) \left[\hat{U}^+(\mathbf{x}_1; t_2, t_1) \left[\int \delta(\mathbf{x}_3 - \mathbf{x}_2) \delta(\mathbf{x}_2 - \mathbf{x}_1) d\mathbf{x}_2 \right] \right] \\
= \hat{U}(\mathbf{x}_3; t_3, t_1) \delta(\mathbf{x}_3 - \mathbf{x}_1),$$
(15)

which reduces to

$$\hat{U}(\mathbf{x}_3; t_3, t_2) \left[\hat{U}^+(\mathbf{x}_1; t_2, t_1) \delta(\mathbf{x}_3 - \mathbf{x}_1) \right] = \hat{U}(\mathbf{x}_3; t_3, t_1) \delta(\mathbf{x}_3 - \mathbf{x}_1).$$
(16)

From Eqs. (10) and (11), we know that Eq. (16) is equivalent to the following expression:

$$\hat{U}(\mathbf{x}_3; t_3, t_2) \left[\hat{U}(\mathbf{x}_3; t_2, t_1) \delta(\mathbf{x}_3 - \mathbf{x}_1) \right] = \hat{U}(\mathbf{x}_3; t_3, t_1) \delta(\mathbf{x}_3 - \mathbf{x}_1),$$
(17)

which proves Eq. (12).

By plugging Eq. (10) into Eq. (3), we find that the operator $\hat{U}(\mathbf{x}; t, t')$ satisfies the following differential equation:

$$\frac{\partial \hat{U}(\mathbf{x};t,t')}{\partial t} = \hat{L}(\mathbf{x},t)\hat{U}(\mathbf{x};t,t').$$
(18)

For further discussion, we also define an inverse operator $\hat{U}^{-1}(\mathbf{x}; t, t')$,

$$1 \equiv \hat{U}^{-1}(\mathbf{x}; t_2, t_1) \hat{U}(\mathbf{x}; t_2, t_1),$$
(19)

which stands for the backward propagator. Explicit form of the inverse operator \hat{U}^{-1} is not needed for our purposes, yet it can be found as a Taylor expansion of the operator $1/\hat{U} = 1/(1 + \int \hat{L}d\tau + \cdots)$. Clearly, the operators \hat{U}^{-1} and \hat{U} commute, i.e., $\hat{U}^{-1}\hat{U} = \hat{U}\hat{U}^{-1}$.

III. PERTURBATION IN THE INTERACTION PICTURE

Consider a stochastic process described with a Fokker-Planck operator $\hat{L}_0(\mathbf{x},t)$. We now impose a perturbation $\delta \hat{L}(\mathbf{x},t)$ on our stochastic process (small, time-dependent perturbation of control parameter, such as, for instance, variations of external temperature, or perturbation of kinetic rate coefficients), setting 212430-3 M. Kryvohuz and S. Mukamel

$$\hat{L}(\mathbf{x},t) = \hat{L}_0(\mathbf{x},t) + \delta \hat{L}(\mathbf{x},t).$$
(20)

We wish to represent time evolution of the probability density of perturbed stochastic process in terms of the unperturbed operator $\hat{L}_0(\mathbf{x},t)$ and the perturbation operator $\delta \hat{L}(\mathbf{x},t)$. To that end, we introduce interaction operator \hat{U}_I defined as

$$\hat{U}(\mathbf{x};t,t') \equiv \hat{U}_0(\mathbf{x};t,t')\hat{U}_I(\mathbf{x};t,t'), \qquad (21)$$

where \hat{U}_0 describes unperturbed stochastic process,

$$\hat{U}_0(\mathbf{x};t,t') = T_+ \exp\left[\int_{t'}^t \hat{L}_0(\mathbf{x},\tau)d\tau\right].$$
(22)

Substituting Eqs. (20) and (21) into Eq. (18), we obtain

$$\left(\frac{\partial \hat{U}_{0}(\mathbf{x};t,t')}{\partial t}\right)\hat{U}_{I}(\mathbf{x};t,t') + \hat{U}_{0}(\mathbf{x};t,t')\left(\frac{\partial \hat{U}_{I}(\mathbf{x};t,t')}{\partial t}\right) \\
= \left(\hat{L}_{0}(\mathbf{x},t) + \delta \hat{L}(\mathbf{x},t)\right)\hat{U}_{0}(\mathbf{x};t,t')\hat{U}_{I}(\mathbf{x};t,t').$$
(23)

Since \hat{U}_0 satisfies Eq. (18) for the unperturbed stochastic process, the first terms on each side of Eq. (23) cancel out, leading to

$$\hat{U}_0(\mathbf{x};t,t')\frac{\partial\hat{U}_I(\mathbf{x};t,t')}{\partial t} = \delta\hat{L}(\mathbf{x},t)\hat{U}_0(\mathbf{x};t,t')\hat{U}_I(\mathbf{x};t,t'), \qquad (24)$$

which can be rewritten as

$$\frac{\partial U_I(\mathbf{x};t,t')}{\partial t} = \hat{U}_0^{-1}(\mathbf{x};t,t')\delta\hat{L}(\mathbf{x},t)\hat{U}_0(\mathbf{x};t,t')\hat{U}_I(\mathbf{x};t,t'), \quad (25)$$

using the inverse operator \hat{U}^{-1} defined in Eq. (19). Equation (25) can be solved recursively for \hat{U}_I in exactly the same way as Eq. (4). Introducing compact notation

$$\delta \hat{L}_{I}(\mathbf{x};t,t') \equiv \hat{U}_{0}^{-1}(\mathbf{x};t,t')\delta \hat{L}(\mathbf{x},t)\hat{U}_{0}(\mathbf{x};t,t'), \qquad (26)$$

which stands for the representation of operator $\delta \hat{L}(\mathbf{x},t)$ in the interaction picture, the solution of Eq. (25) takes the following form:

$$\hat{U}_{I}(\mathbf{x};t,t') = 1 + \int_{t'}^{t} \hat{L}_{I}(\mathbf{x};\tau,t')d\tau + \int_{t'}^{t} d\tau_{2} \int_{t'}^{\tau_{2}} d\tau_{1} \hat{L}_{I}(\mathbf{x};\tau_{2},t') \hat{L}_{I}(\mathbf{x};\tau_{1},t') + \cdots$$
(27)

The final step is to plug Eq. (27) back to Eq. (21), which provides the desired expansion of the perturbed operator \hat{U} in terms of the unperturbed operator \hat{U}_0 and perturbation $\delta \hat{L}$. Substituting Eqs. (27) and (26) into Eq. (21), we obtain

$$\hat{U}(\mathbf{x};t,t') = \hat{U}_{0}(\mathbf{x};t,t') + \int_{t'}^{t} \hat{U}_{0}(\mathbf{x};t,t') \hat{U}_{0}^{-1}(\mathbf{x};\tau,t') \delta \hat{L}(\mathbf{x},\tau) \hat{U}_{0}(\mathbf{x};\tau,t') d\tau + \int_{t'}^{t} d\tau_{2} \int_{t'}^{\tau_{2}} d\tau_{1} \hat{U}_{0}(\mathbf{x};t,t') \hat{U}_{0}^{-1}(\mathbf{x};\tau_{2},t') \delta \hat{L}(\mathbf{x},\tau_{2}) \hat{U}_{0}(\mathbf{x};\tau_{2},t') \hat{U}_{0}^{-1}(\mathbf{x};\tau_{1},t') \delta \hat{L}(\mathbf{x},\tau_{1}) \hat{U}_{0}(\mathbf{x};\tau_{1},t') + \cdots$$
(28)

Using the properties of the operator \hat{U} , given by Eqs. (12) and (19), one can further simplify Eq. (28),

$$\hat{U}(\mathbf{x};t,t') = \hat{U}_{0}(\mathbf{x};t,t') + \int_{t'}^{\tau} \hat{U}_{0}(\mathbf{x};t,\tau)\delta\hat{L}(\mathbf{x},\tau)\hat{U}_{0}(\mathbf{x};\tau,t')d\tau + \int_{t'}^{t} d\tau_{2} \int_{t'}^{\tau_{2}} d\tau_{1}\hat{U}_{0}(\mathbf{x};t,\tau_{2})\delta\hat{L}(\mathbf{x},\tau_{2})\hat{U}_{0}(\mathbf{x};\tau_{2},\tau_{1})\delta\hat{L}(\mathbf{x},\tau_{1})\hat{U}_{0}(\mathbf{x};\tau_{1},t') + \cdots$$
(29)

Equation (29) can be written in the final compact form

$$\hat{U}(\mathbf{x};t,t') = \hat{U}_0(\mathbf{x};t,t') + \sum_{n=1}^{\infty} \int_{t'}^t d\tau_n \int_{t'}^{\tau_n} d\tau_{n-1} \dots \int_{t'}^{\tau_2} d\tau_1$$
$$\times \hat{U}_0(\mathbf{x};t,\tau_n) \delta \hat{L}(\mathbf{x},\tau_n) \hat{U}_0(\mathbf{x};\tau_n,\tau_{n-1})$$
$$\times \delta \hat{L}(\mathbf{x},\tau_{n-1}) \dots \hat{U}_0(\mathbf{x};\tau_1,t').$$
(30)

IV. RESPONSE TO IMPULSIVE PERTURBATIONS

We wish to express the system's response to external perturbations. First, we consider the linear response of a stochastic process $\mathbf{x}(t)$ described by the Fokker-Planck operator $\hat{L}_0(\mathbf{x},t)$ and which, at time t = 0, is prepared in a steady state characterized by distribution function $\rho_{ss}(\mathbf{x})$. One can consider an experiment, in which the above stochastic system is subjected to a small perturbation at time t_1 : $\delta \hat{L}(\mathbf{x},t) = \epsilon \hat{V}(\mathbf{x},t)\delta(t-t_1)$ (where $\epsilon \ll 1$ and operator $\hat{V}(\mathbf{x},t)$ depends on the problem of interest), and the value of some observable

 $f(\mathbf{x})$, which is a function of stochastic variable \mathbf{x} , is then measured at a later time t_2 . If such perturbation, measurement process is repeated several times, one can calculate the value of $f(\mathbf{x}(t_2))$ averaged over several realizations of the perturbed stochastic trajectories. The average value $\langle f(\mathbf{x}(t_2)) \rangle$ reads

$$\langle f(\mathbf{x}(t_2))\rangle = \int d\mathbf{x}_2 \int d\mathbf{x}_1 f(\mathbf{x}_2) P(\mathbf{x}_2, t_2 | \mathbf{x}_1, 0) \rho_{ss}(\mathbf{x}_1), \quad (31)$$

where $P(\mathbf{x}_2, t_2 | \mathbf{x}_1, 0)$ is the perturbed transition probability. Substituting Eq. (10) into Eq. (31) results in

$$\langle f(\mathbf{x}(t_2)) \rangle = \int d\mathbf{x}_2 \int d\mathbf{x}_1 f(\mathbf{x}_2) \\ \times \left[\hat{U}(\mathbf{x}_2; t_2, 0) \delta(\mathbf{x}_2 - \mathbf{x}_1) \right] \rho_{ss}(\mathbf{x}_1) \\ = \int f(\mathbf{x}_2) \hat{U}(\mathbf{x}_2; t_2, 0) \rho_{ss}(\mathbf{x}_2) d\mathbf{x}_2.$$
(32)

By inserting Eq. (30) into Eq. (32), we then express the perturbed operator \hat{U} in Eq. (32) using the properties of the

unperturbed stochastic process,

$$\langle f(\mathbf{x}(t_2)) \rangle = \langle f(\mathbf{x}(t_2)) \rangle_0 + \epsilon \int f(\mathbf{x}_2) \hat{U}_0(\mathbf{x}_2; t_2, t_1) \\ \times \hat{V}(\mathbf{x}_2, t_1) \rho_{ss}(\mathbf{x}_2) d\mathbf{x}_2 + O(\epsilon^2),$$
(33)

where we have used steady-state property $\hat{U}_0(\mathbf{x}_2; t_1, 0) \rho_{ss}(\mathbf{x}_2) = \rho_{ss}(\mathbf{x}_2)$.

We next define the impulsive linear response function

$$R^{(1)}(t) = \frac{\partial}{\partial \epsilon} \langle f(\mathbf{x}(t)) \rangle |_{\epsilon=0}, \qquad (34)$$

and making use of Eq. (33), we obtain

$$R^{(1)}(t) = \int f(\mathbf{x})\hat{U}_0(\mathbf{x};t,0)\hat{V}(\mathbf{x},0)\rho_{ss}(\mathbf{x})d\mathbf{x}$$
$$\equiv \left\langle f(\mathbf{x})\hat{U}_0(t,0)\hat{V}(0)\right\rangle_0, \tag{35}$$

where we have defined the average of operator over unperturbed trajectories $\langle \hat{A} \rangle_0 \equiv \int \hat{A}(\mathbf{x}) \rho_{ss}(\mathbf{x}) d\mathbf{x}$ and for brevity omitted **x** in the arguments of the operators \hat{U}_0 and \hat{V} .

In a similar way, one can calculate the non-linear response functions of a stochastic process. The non-linear response function can be defined as the second derivative of response with respect to the perturbation. In case of impulsive perturbations, the second order term in $\delta \hat{L}$ in Eq. (30) survives only if we perturb the stochastic process twice (since impulsive perturbation cannot interact with itself and thus time-ordered integrals $\tau_n > \tau_{n-1} > \cdots > \tau_1$ vanish). We thus consider an experiment with two impulsive perturbations: $\delta \hat{L}(\mathbf{x},t) = \epsilon_1 \hat{V}(\mathbf{x},t) \delta(t-t_1) + \epsilon_2 \hat{V}(\mathbf{x},t) \delta(t-t_2)$, $t_2 > t_1$. By substituting the latter into Eq. (30), we obtain the perturbative expansion of the operator \hat{U} ,

$$\hat{U}(\mathbf{x};t,t') = \hat{U}_{0}(\mathbf{x};t,t') + \epsilon_{1}\hat{U}_{0}(\mathbf{x};t,t_{1})\hat{V}(\mathbf{x},t_{1})\hat{U}_{0}(\mathbf{x};t_{1},t')
+ \epsilon_{2}\hat{U}_{0}(\mathbf{x};t,t_{2})\hat{V}(\mathbf{x},t_{2})\hat{U}_{0}(\mathbf{x};t_{2},t')
+ \epsilon_{1}\epsilon_{2}\hat{U}_{0}(\mathbf{x};t,t_{2})\hat{V}(\mathbf{x},t_{2})\hat{U}_{0}(\mathbf{x};t_{2},t_{1})
\times \hat{V}(\mathbf{x},t_{1})\hat{U}_{0}(\mathbf{x};t_{1},t') + \cdots.$$
(36)

Defining the second-order response function as

$$R^{(2)}(t - t_2, t_2 - t_1) = \frac{\partial^2}{\partial \epsilon_1 \partial \epsilon_2} \langle f(\mathbf{x}(t)) \rangle |_{\epsilon_1 = \epsilon_2 = 0}, \quad (37)$$

and substituting Eq. (36) into Eq. (32), we obtain the compact final expression

$$R^{(2)}(\tau_{2},\tau_{1}) = \int f(\mathbf{x})\hat{U}_{0}(\mathbf{x};t,t_{2})\hat{V}(\mathbf{x},t_{2})\hat{U}_{0}(\mathbf{x};t_{2},t_{1}) \times \hat{V}(\mathbf{x},t_{1})\hat{U}_{0}(\mathbf{x};t_{1},0)\rho_{ss}(\mathbf{x})d\mathbf{x} \equiv \langle f(\mathbf{x})\hat{U}_{0}(t,t_{2})\hat{V}(t_{2})\hat{U}_{0}(t_{2},t_{1})\hat{V}(t_{1})\rangle_{0}, \quad (38)$$

where $\tau_2 = t - t_2$ and $\tau_1 = t_2 - t_1$.

Expressions (35) and (38) are very similar to the response functions in classical or quantum dynamics and have a clear physical meaning. For instance, from Eq. (38), it follows that the second-order response function is an average of the product of operators which act in the following sequence: perturbation operator \hat{V} at time t_1 , evolution operator \hat{U}_0 from time t_1 to time t_2 , perturbation operator \hat{V} at time t_2 , evolution operator \hat{U}_0 from time t_2 to time t, and a measurement of \mathbf{x} at time t. The perturbation operator \hat{V} is known and often can be controlled externally, while the evolution operator \hat{U}_0 depends entirely on the unperturbed Fokker-Planck operator \hat{L}_0 , and therefore, various (measurable) correlation functions which contain different occurrences of \hat{U}_0 operators should contain valuable information on the dynamics of the unperturbed stochastic process governed by the Fokker-Planck operator \hat{L}_0 . The linear and second-order response functions, given by Eqs. (35) and (38), respectively, thus serve as time-dependent *experimentally* accessible measures of the unperturbed stochastic process $\mathbf{x}(t)$.

In the same way, one can obtain the *n*th order nonlinear response functions which represents experiments with *n* impulsive perturbations $\delta \hat{L}(\mathbf{x},t) = \sum_{j=1}^{n} \epsilon_j \hat{V}(\mathbf{x},t) \delta(t-t_j)$,

$$R^{(n)}(\tau_n, \tau_{n-1}, \dots, \tau_1) = \left\langle f(\mathbf{x}) \hat{U}_0(t, t_n) \hat{V}(t_n) \right. \\ \left. \times \hat{U}_0(t_n, t_{n-1}) \dots \hat{V}(t_1) \right\rangle_0, \quad (39)$$

where $\tau_j = t - t_j$ for j = n and $\tau_j = t_{j+1} - t_j$ for $j = n - 1, \dots, 1$.

V. GENERALIZED RESPONSE FUNCTIONS

In Sec. IV, we have derived expressions for nonlinear response functions, which can serve as multi-dimensional (i.e., multi-time) measures of the stochastic process $\mathbf{x}(t)$. These measures correspond to the standard protocol involving nperturbations followed by a single measurement. We now generalize these experiments to include m perturbations and k measurements. Consider, for instance, an experiment, in which we measure the value of an observable $f(\mathbf{x})$ at time t_1 , then perturb our stochastic system with $\delta L = \epsilon \hat{V}(\mathbf{x}, t) \delta(t - t_2)$ at time t_2 , and finally measure the value of observable $g(\mathbf{x})$ at time t_3 . Having the values $f(\mathbf{x}(t_1))$ and $g(\mathbf{x}(t_3))$ of the stochastic variable x, one can construct the correlation function $\langle g(\mathbf{x}(t_3)) f(\mathbf{x}(t_1)) \rangle$, in which the average is taken over various stochastic realizations of the same perturbation protocol (i.e., measurement at t_1 , perturbation at t_2 , and another measurement at t_3). This correlation function should depend on the type of perturbation and the time t_2 of the perturbation. Such an experiment should thus provide non-vanishing (twodimensional) quantities

$$R_{+-+}(t_3 - t_2, t_2 - t_1) \equiv \frac{\partial}{\partial \epsilon} \langle g(\mathbf{x}(t_3)) f(\mathbf{x}(t_1)) \rangle |_{\epsilon=0}.$$
 (40)

The "+" index denotes measurement and "–" denotes perturbation. We shall now calculate the new measure R_{+-+} . The correlation function $\langle g(\mathbf{x}(t_3)) f(\mathbf{x}(t_1)) \rangle$ reads

$$\langle g(\mathbf{x}(t_3))f(\mathbf{x}(t_1))\rangle$$

$$= \int d\mathbf{x}_3 d\mathbf{x}_1 g(\mathbf{x}_3) P(\mathbf{x}_3, t_3 | \mathbf{x}_1, t_1) f(\mathbf{x}_1) \rho_{ss}(\mathbf{x}_1)$$

$$= \int d\mathbf{x}_3 d\mathbf{x}_1 g(\mathbf{x}_3) \left[\hat{U}(\mathbf{x}_3; t_3, t_1) \delta(\mathbf{x}_3 - \mathbf{x}_1) \right] f(\mathbf{x}_1) \rho_{ss}(\mathbf{x}_1)$$

$$= \int d\mathbf{x}_3 g(\mathbf{x}_3) \left[\hat{U}(\mathbf{x}_3; t_3, t_1) f(\mathbf{x}_3) \rho_{ss}(\mathbf{x}_3) \right]. \quad (41)$$

Substituting Eq. (29) into Eq. (41), we get

$$\langle g(\mathbf{x}(t_3)) f(\mathbf{x}(t_1)) \rangle = \langle g(\mathbf{x}(t_3)) f(\mathbf{x}(t_1)) \rangle_0 + \epsilon \int dx g(\mathbf{x}) \hat{U}_0(\mathbf{x}; t_3, t_2) \hat{V}(\mathbf{x}, t_2) \times \hat{U}_0(\mathbf{x}; t_2, t_1) f(\mathbf{x}) \rho_{ss}(\mathbf{x}) + O(\epsilon^2).$$
(42)

Inserting Eq. (42) into Eq. (40), we finally obtain

$$R_{+-+}(t_{3} - t_{2}, t_{2} - t_{1}) = \int d\mathbf{x}g(\mathbf{x})\hat{U}_{0}(\mathbf{x}; t_{3}, t_{2})\hat{V}(\mathbf{x}, t_{2})$$

$$\times \hat{U}_{0}(\mathbf{x}; t_{2}, t_{1})f(\mathbf{x})\rho_{ss}(\mathbf{x})$$

$$= \langle g(\mathbf{x})\hat{U}_{0}(t_{3}, t_{2})\hat{V}(t_{2})\hat{U}_{0}(t_{2}, t_{1})f(\mathbf{x})\rangle_{0}.$$
(43)

Equation (43) contains the following sequence of operations: measurement of $f(\mathbf{x})$ at time t_1 , time evolution $\hat{U}_0(t_2,t_1)$ of unperturbed stochastic system from t_1 to t_2 , perturbation \hat{V} at t_2 , time evolution of $\hat{U}_0(t_3,t_2)$ of unperturbed stochastic system from t_2 to t_3 , and a measurement $g(\mathbf{x})$ at t_3 . This is a completely new (experimentally accessible) correlation function, which depends on two time intervals $t_3 - t_2$ and $t_2 - t_1$, and thus should carry additional information about the unperturbed stochastic process.

In a similar way, we can consider various experiments with k perturbations and m = n - k - 1 measurements which are described by a new class of multidimensional measures,

$$R_{\{+\dots+,-\dots-\}}(\tau_n,\dots,\tau_1) = \frac{\partial^k}{\partial\epsilon_1\dots\partial\epsilon_k} \langle f_m(\mathbf{x}(t_m))\dots f_1(\mathbf{x}(t_1))\rangle|_{\epsilon_i=0}, \quad (44)$$

in which $\{+\dots+,-\dots-\}$ denote various permutations of *m* "+" (measurements) and *k* "–" (perturbations). Note that the leftmost symbol being always "+" due to causality. We call these new quantities GRFs.^{8,9} GRFs simply extend the class of correlation and response functions. For instance, in the notation of Eq. (44), an ordinary two-time correlation function corresponds to R_{++} , while the second-order nonlinear response function corresponds to R_{+--} .

The GRF expression has a simple form and is a correlation function of appropriate ordering of measurements $f(\mathbf{x})$ and perturbations \hat{V} with time evolution operators $\hat{U}_0(t_j, t_{j-1})$ in between. It is also useful to recast the expressions for GRFs using Eq. (10) in terms of unperturbed transition probabilities $P_0(\mathbf{x}, t | \mathbf{x}', t')$. We demonstrate this for two-dimensional GRFs:

(a) 3-point correlation function

$$R_{+++}(\tau_{2},\tau_{1}) = \int d\mathbf{x}_{1} d\mathbf{x}_{2} d\mathbf{x}_{3}$$

× $f_{3}(\mathbf{x}_{3}) P_{0}(\mathbf{x}_{3},\tau_{2}+\tau_{1}|\mathbf{x}_{2},\tau_{1}) f_{2}(\mathbf{x}_{2})$
× $P_{0}(\mathbf{x}_{2},\tau_{1}|\mathbf{x}_{1},0) f_{1}(\mathbf{x}_{1}).\rho_{ss}(\mathbf{x}_{1}),$ (45)

(b) second-order response function

$$R_{+--}(\tau_2, \tau_1) = \int d\mathbf{x}_1 d\mathbf{x}_2 d\mathbf{x}_3$$

$$\times f_3(\mathbf{x}_3) P_0(\mathbf{x}_3, \tau_2 + \tau_1 | \mathbf{x}_2, \tau_1) \hat{V}(\mathbf{x}_2)$$

$$\times P_0(\mathbf{x}_2, \tau_1 | \mathbf{x}_1, 0) \hat{V}(\mathbf{x}_1) \rho_{ss}(\mathbf{x}_1), \qquad (46)$$

(c) GRF corresponding to 1 perturbation followed by 2 measurements,

$$R_{++-}(\tau_2, \tau_1) = \int d\mathbf{x}_1 d\mathbf{x}_2 d\mathbf{x}_3$$

× $f_3(\mathbf{x}_3) P_0(\mathbf{x}_3, \tau_2 + \tau_1 | \mathbf{x}_2, \tau_1) f_2(\mathbf{x}_2)$
× $P_0(\mathbf{x}_2, \tau_1 | \mathbf{x}_1, 0) \hat{V}(\mathbf{x}_1) \rho_{ss}(\mathbf{x}_1),$ (47)

(d) GRF corresponding to a measurement followed by 1 perturbation and then 1 measurement,

$$R_{+-+}(\tau_{2},\tau_{1}) = \int d\mathbf{x}_{1} d\mathbf{x}_{2} d\mathbf{x}_{3}$$

 $\times f_{3}(\mathbf{x}_{3}) P_{0}(\mathbf{x}_{3},\tau_{2}+\tau_{1}|\mathbf{x}_{2},\tau_{1}) \hat{V}(\mathbf{x}_{2})$
 $\times P_{0}(\mathbf{x}_{2},\tau_{1}|\mathbf{x}_{1},0) f_{1}(\mathbf{x}_{1}) \rho_{ss}(\mathbf{x}_{1}).$ (48)

In (45)–(48), $\hat{V}(\mathbf{x})$ is perturbation operator specific to the type of perturbation used to disturb the stochastic system. Examples will be given in Secs. VII–IX.

VI. RESPONSE TO PERIODIC PERTURBATIONS; FREQUENCY-DOMAIN SIGNALS

Periodic perturbations are an alternative to impulsive perturbations that can be imposed on a stochastic system¹² via an oscillatory variation of one of its parameters, such as temperature, see also Sec. VII B. Rather than varying time-ordered interactions, we now scan frequencies with no control over time ordering.

We consider the perturbed Fokker-Planck operator $\hat{L}(\mathbf{x},t) = \hat{L}_0(\mathbf{x},t) + \delta \hat{L}(\mathbf{x},t)$, where

$$\delta \hat{L}(\mathbf{x},t) = \frac{1}{2\pi} \sum_{n=1} \epsilon_n e^{-\iota \omega_n t} \hat{V}(\mathbf{x}).$$
(49)

We first consider the linear response, n = 1. By substituting Eq. (49) into Eq. (30) and then into Eq. (32), we obtain the mean of observable $f(\mathbf{x}(t))$,

$$\langle f(\mathbf{x}(t)) \rangle = \langle f(\mathbf{x}(t)) \rangle_0 + \frac{\epsilon_1}{2\pi} \int_0^t d\tau \int f(\mathbf{x}) \hat{U}_0(\mathbf{x}; t, \tau) \times \hat{V}(\mathbf{x}) e^{-i\omega_1 \tau} \rho_{ss}(\mathbf{x}) d\mathbf{x}.$$
 (50)

In case of a time-independent Fokker-Planck operator $L_0(\mathbf{x})$, the operator $\hat{U}_0(\mathbf{x}; t_2, t_1) = \hat{U}_0(\mathbf{x}; t_2 - t_1)$ solely depends on the time difference $t_2 - t_1$, so that this leads to a simple form of the Fourier transform of Eq. (50),

$$\left\langle \widetilde{f}(\omega) \right\rangle = \left\langle \widetilde{f}(\omega) \right\rangle_0 + \epsilon_1 \left\langle f(\mathbf{x}) \widehat{U}_0(\mathbf{x};\omega) \widehat{V}(\mathbf{x}) \right\rangle \delta(\omega - \omega_1).$$
(51)

The coefficient in front of the delta-function in Eq. (51), which we denote the susceptibility $\chi^{(1)}(\omega)$, is simply given by the Fourier transform of the time-domain linear response function in Eq. (35),

$$\chi^{(1)}(\omega) = \widetilde{R^{(1)}}(\omega).$$
 (52)

From Eq. (51), one sees that the susceptibility can be measured as the spectral amplitude of the derivative $\partial \langle \tilde{f}(\omega) \rangle / \partial \epsilon_1$ at frequency ω_1 , i.e.,

$$\chi^{(1)}(\omega_1) = \frac{\partial}{\partial \epsilon_1} \int_{\omega_1 - \Delta \omega}^{\omega_1 + \Delta \omega} \left\langle \widetilde{f}(\omega) \right\rangle \Big|_{\epsilon_1 = 0} d\omega.$$
(53)

In case of the second-order response, n = 2, one considers two simultaneous perturbations with two different frequencies $\delta \hat{L}(\mathbf{x},t) = [\epsilon_1 e^{-\iota \omega_1 t} \hat{V}(\mathbf{x}) + \epsilon_2 e^{-\iota \omega_2 t} \hat{V}(\mathbf{x})]/2\pi$. Again, substituting it to Eq. (30), we obtain

$$\frac{\partial^2}{\partial \epsilon_1 \partial \epsilon_2} \left\langle \widetilde{f}(\omega) \right\rangle = \delta(\omega_1 + \omega_2 - \omega) \\ \times \left[\left\langle f(\mathbf{x}) \hat{U}_0(\mathbf{x}; \omega_1 + \omega_2) \hat{V}(\mathbf{x}) \hat{U}_0(\mathbf{x}; \omega_1) \hat{V}(\mathbf{x}) \right\rangle \\ + \left\langle f(\mathbf{x}) \hat{U}_0(\mathbf{x}; \omega_1 + \omega_2) \hat{V}(\mathbf{x}) \hat{U}_0(\mathbf{x}; \omega_2) \hat{V}(\mathbf{x}) \right\rangle \right].$$
(54)

The second-order susceptibility can then be defined as

$$\chi^{(2)}(\omega_2,\omega_1) = \frac{1}{2} \sum_{p} \left\langle f(\mathbf{x}) \hat{U}_0(\mathbf{x};\omega_1 + \omega_2) \hat{V}(\mathbf{x}) \hat{U}_0(\mathbf{x};\omega_1) \hat{V}(\mathbf{x}) \right\rangle, \quad (55)$$

where the summation p is over the two permutations of frequencies ω_1 and ω_2 .

The higher-order nonlinear susceptibilities can be obtained similarly,

$$\chi^{(n)}(\omega_n, \omega_{n-1}, \dots, \omega_1) = \frac{1}{p} \sum_p \langle f(\mathbf{x}) \hat{U}_0(\mathbf{x}; \omega_n + \omega_{n-1} + \dots + \omega_1) \hat{V} \dots (\mathbf{x}) \hat{U}_0(\mathbf{x}; \omega_1) \hat{V}(\mathbf{x}) \rangle.$$
(56)

In analogy to Sec. V, one can also extend the class of nonlinear susceptibilities $\chi^{(m,k)}$ to a general class of experiments in which a stochastic system is perturbed on k frequencies $\omega'_1, \ldots, \omega'_k$, and *m*-point correlation function $\langle f_1(\omega_1) \ldots f_m(\omega_m) \rangle$ is measured as a response to these perturbations,

$$\left\langle \widetilde{f}_{m}(\omega_{m})\dots\widetilde{f}_{1}(\omega_{1})\right\rangle = \left\langle \widetilde{f}_{m}(\omega_{m})\dots\widetilde{f}_{1}(\omega_{1})\right\rangle_{0} + \sum_{n}\int d\omega_{1}'\dots\int d\omega_{k}'\delta(\omega_{1}+\dots+\omega_{m}-\omega_{1}'-\dots-\omega_{k}')$$
$$\times\chi^{(m,k)}(\omega_{1},\dots,\omega_{m},\omega_{1}',\dots,\omega_{k}')\widetilde{F}(\omega_{1}')\dots\widetilde{F}(\omega_{k}').$$
(57)

It follows from Eq. (57) that the m + k frequencies ω'_i, ω_j are constrained via a delta-function, such that only m + k - 1of them are independent. We therefore omit one frequency ω_m from the arguments of $\chi^{(m,k)}$ and consider (k + m - 1)-th order generalized susceptibility as a function of k + m - 1frequencies $\chi^{(m,k)}(\omega_m, \ldots, \omega_2, \omega'_k, \ldots, \omega'_1)$.

The generalized nonlinear susceptibility can be expressed as a linear combination of Fourier transforms of the *n*th order (n = k + m - 1) GRFs $R_{+...,-...}$ of Sec. V. Below, we present the frequency-domain versions of Eqs. (45)–(48), i.e., generalized 2D susceptibilities corresponding to various combinations of three perturbations and measurements:

(a) 3-point correlation function

$$\chi^{(3,0)}(\omega_2,\omega_1) = \frac{1}{2} \sum_p \int d\mathbf{x}_1 d\mathbf{x}_2 d\mathbf{x}_3$$
$$\times f_3(\mathbf{x}_3) \widetilde{P}_0(\mathbf{x}_3,\omega_2|\mathbf{x}_2) f_2(\mathbf{x}_2)$$
$$\times \widetilde{P}_0(\mathbf{x}_2,\omega_2+\omega_1|\mathbf{x}_1) f_1(\mathbf{x}_1) \rho_{ss}(\mathbf{x}_1), \qquad (58)$$

 (b) second-order susceptibility for periodic perturbations on frequencies ω'₁ and ω'₂,

$$\chi^{(1,2)}(\omega_2',\omega_1') = \frac{1}{2} \sum_p \int d\mathbf{x}_1 d\mathbf{x}_2 d\mathbf{x}_3$$
$$\times f_3(\mathbf{x}_3) \widetilde{P}_0(\mathbf{x}_3,\omega_2' + \omega_1' | \mathbf{x}_2) \hat{V}(\mathbf{x}_2)$$
$$\times \widetilde{P}_0(\mathbf{x}_2,\omega_1' | \mathbf{x}_1) \hat{V}(\mathbf{x}_1) \rho_{ss}(\mathbf{x}_1), \qquad (59)$$

(c) generalized susceptibility corresponding to perturbation on one frequency ω'_1 and measurements on 2 frequencies ω_1 and ω_2 such that $\omega'_1 = \omega_1 + \omega_2$,

$$\chi^{(2,1)}(\omega_2,\omega_1) = \frac{1}{2} \sum_p \int d\mathbf{x}_1 d\mathbf{x}_2 d\mathbf{x}_3$$

$$\times \left[f_3(\mathbf{x}_3) \widetilde{P}_0(\mathbf{x}_3,\omega_2 | \mathbf{x}_2) f_2(\mathbf{x}_2) \widetilde{P}_0(\mathbf{x}_2,\omega_2 + \omega_1 | \mathbf{x}_1) \right. \\ \times \hat{V}(\mathbf{x}_1) \rho_{ss}(\mathbf{x}_1) + f_3(\mathbf{x}_3) \widetilde{P}_0(\mathbf{x}_3,\omega_2 | \mathbf{x}_2) \\ \times \hat{V}(\mathbf{x}_2) \widetilde{P}_0(\mathbf{x}_2,-\omega_1 | \mathbf{x}_1) f_1(\mathbf{x}_1) \rho_{ss}(\mathbf{x}_1) \right].$$
(60)

Note that while in the time domain we had 4 two-time GRFs (Eqs. (45) and (46)), in the frequency domain we have only 3 two-frequency generalized susceptibilities (Eqs. (58)–(60)). This is because we are losing time ordering when switching to the frequency domain, i.e., in the frequency domain, there is no difference between the + + - and + - + techniques since all processes involving of 1 perturbation and 2 measurements are occurring simultaneously.

VII. EXAMPLE 1: THE ORNSTEIN-UHLENBECK PROCESS

The OUP represents stochastic diffusion in a harmonic potential. It is frequently used as a simplified model of diffusion in external force fields, since it can be solved analytically. Its dynamics is governed by the stochastic Langevin equation

$$\frac{dx}{dt} = -kx + \sqrt{D}\xi(t), \tag{62}$$

where $\xi(t)$ is a delta-correlated noise $\langle \xi(t)\xi(s) \rangle = \delta(t-s)$. The corresponding Fokker-Planck equation reads¹³

$$\frac{\partial}{\partial t}\rho(x,t) = \frac{\partial}{\partial x}\left(kx\rho(x,t)\right) + \frac{D}{2}\frac{\partial^2}{\partial x^2}\rho(x,t), \quad (63)$$

with the Fokker-Planck operator

$$\hat{L}_0(x) = \frac{\partial}{\partial x} kx + \frac{D}{2} \frac{\partial^2}{\partial x^2}.$$
(64)

The transition probability of this model reads¹⁰

$$P_{0}(x,t|x',t') = \sqrt{\frac{k}{\pi D\left(1 - e^{-2k(t-t')}\right)}} \times \exp\left[-\frac{k(x - e^{-k(t-t')}x')^{2}}{D\left(1 - e^{-2k(t-t')}\right)}\right], \quad (65)$$

which in the long time $t - t' \rightarrow \infty$ limit leads to the stationary distribution

$$\rho_s(x) = \sqrt{\frac{k}{\pi D}} \exp(-kx^2/D).$$
(66)

A. Impulsive perturbation of the coordinate

We now subject the OUP to some external perturbation and examine its response. We consider perturbation in a form of sudden displacement of the coordinate x by small ϵ at time t_1 . This perturbation can be described by an additional δ -function term in stochastic differential equation (62),

$$\frac{dx}{dt} = -kx + \sqrt{D}\xi(t) + \epsilon\delta(t - t_1).$$
(67)

Indeed, if we integrate both parts of equation (67) in the vicinity of t_1 , we get

$$x(t_1 + \Delta t) - x(t_1) = -kx(t_1)\Delta t + \sqrt{D\Delta W} + \epsilon$$
$$= \epsilon, \ \Delta t \to 0, \tag{68}$$

where $\int_{t_1-\Delta t/2}^{t_1+\Delta t/2} \xi(t) dt = \Delta W \sim \sqrt{\Delta t} \to 0$ for $\Delta t \to 0$. This also implies that in discrete stochastic simulations of Eq. (68), one needs to take such intermediate mesoscopic²³ values of ϵ so that ϵ is small compared to the macroscopic variation of x and at the same time $\epsilon \gg \sqrt{D\Delta t}$.

The Fokker-Planck equation with the perturbation in Eq. (68) reads

$$\frac{\partial}{\partial t}\rho(x,t) = \frac{\partial}{\partial x}\left(kx\rho(x,t)\right) + \frac{D}{2}\frac{\partial^2}{\partial x^2}\rho(x,t) -\epsilon\delta(t-t_1)\frac{\partial}{\partial x}\rho.$$
(69)

This implies that the perturbation operator is

$$\delta \hat{L}(x,t) = -\epsilon \delta(t-t_1) \frac{\partial}{\partial x}.$$
(70)

Correspondingly, the infinitesimal perturbation operator $\hat{V}(x)$ (defined in the beginning of section V as $\delta \hat{L} = \epsilon \hat{V}(x)\delta(t - t_1)$) appearing in Eqs. (45)-(48) reads

$$\hat{V}(x) = -\frac{\partial}{\partial x}.$$
(71)

Using Eqs. (65), (66), and (71), it is now possible to calculate the generalized response functions in Eqs. (45)-(48)analytically for the OUP. 2D generalized response functions in harmonic potential will vanish for perturbations of the type in Eq. (71) in case of $f_1(x) = f_2(x) = f_3(x) = x$. We therefore consider the following simple observables $f_1(x_1)$ $= x_1, f_2(x_2) = x_2$, and $f_3(x_3) = x_3^2$. We note, however, that other types of perturbations, like the temperature perturbation discussed in Sec. VII B, can result in non-zero 2D measures for any kind of observable. The exact results for 2D measures $R_{+++}(\tau_2,\tau_1)$ and $R_{++-}(\tau_2,\tau_1)$ and $R_{+-+}(\tau_2,\tau_1)$ and $R_{+--}(\tau_2,\tau_1)$ calculated with Eqs. (65), (66), and (71) are shown in Fig. 1. They compare well with the numerical results of Fig. 2, which were calculated using the non-equilibrium approach, Eq. (44). Both approaches are equivalent. We note that some signals look similar. This is to be expected for Gaussian processes, where these measures are related to each other via the generalized fluctuation-dissipation relations.⁹

B. Periodic temperature modulation of the OUP

Harmonic perturbations can be realized by periodically modulating the strength of stochastic noise $\xi(t)$ in Eq. (62), for instance, by changing the temperature. For a general stochastic process,

$$\frac{dx}{dt} = -\frac{dU(x)}{dx} + \sqrt{D}\xi(t),\tag{72}$$

of diffusion in potential U(x), periodic temperature modulation will result in periodic modulation of diffusion constant $D = D_0 + \epsilon \cos(\omega t)$, which depends linearly on temperature. The perturbation operator then reads

$$\delta \hat{L}(x,t) = \frac{\epsilon}{4} \left(e^{\iota \omega t} + e^{-\iota \omega t} \right) \frac{\partial^2}{\partial x^2}.$$
 (73)

The operator of infinitesimal perturbation appearing in Eqs. (58)–(60) is then given by

$$\hat{V}(x) = \frac{\pi}{2} \frac{\partial^2}{\partial x^2}.$$
(74)

By substituting Eq. (74) in Eqs. (70)–(60), one can calculate the exact 2D susceptibilities. Alternatively, these can be calculated using the non-equilibrium approach, Eq. (57), by taking derivatives of correlation functions with respect to the amplitudes of oscillatory perturbations. This mimics the experimental protocol. Results of both calculations are shown in Fig. 3. Since temperature perturbations correspond to the second-order (even) derivative, then $\chi^{(3,0)}$ and $\chi^{(1,2)}$ must vanish for symmetric potentials since they effectively involve averaging over the product of odd number of random variables. This can be one of the merits of the generalized 2D measures as compared to the regular 2D correlation and response functions for studying stochastic processes by temperature perturbations.



FIG. 1. Exact generalized response functions of the OU process Eq. (63): (a) $R_{+++}(\tau_2, \tau_1)$, (b) $R_{++-}(\tau_2, \tau_1)$, (c) $R_{+-+}(\tau_2, \tau_1)$, and (d) $R_{+--}(\tau_2, \tau_1)$.

FIG. 2. Generalized response functions of the OU process calculated numerically: (a) $R_{+++}(\tau_2,\tau_1)$, (b) $R_{++-}(\tau_2,\tau_1)$, (c) $R_{+-+}(\tau_2,\tau_1)$, and (d) $R_{+--}(\tau_2,\tau_1)$.



FIG. 3. Generalized susceptibility $\chi^{(2,1)}(\omega_2,\omega_1)$ of the OU process: (a) exact result and (b) numerically calculated result.

VIII. EXAMPLE 2: IMPULSIVE TEMPERATURE PERTURBATIONS OF DIFFUSION PROCESS IN ANHARMONIC AND DOUBLE-WELL POTENTIALS

Diffusion in anharmonic, such as double-well, potentials is an important process, which is often used to model biological processes.¹⁴ For instance, conformational dynamics of proteins is often described as a diffusion in double well potential along a reaction coordinate.¹⁵ One can study such stochastic processes using temperature perturbations since it is hard to control perturbations of stochastic variables and often the relevant stochastic variable (reaction coordinate) is not known. By applying temperature perturbations at different times, one can measure various multidimensional GRFs as described in Secs. IV-VII and obtain, e.g., information on conformational dynamics of single proteins. It is interesting to examine the GRFs for different potentials and what information they can provide. Such experiment can be done, for instance, by measuring the electron-transfer (ET) rate between donor and acceptor on different ends of a protein at different times.¹⁶ The ET rate constant depends exponentially on the distance between donor and acceptor $k(t) = k_0 \exp(-\beta x(t));$ thus, the stochastic variable $X(t) \equiv -\log[k(t)]$ can be a measure of conformational dynamics of the protein.

Temperature perturbations can be either impulsive $\Delta T \delta(t-\tau)$ (such as pulsed heating^{17,18}) or a step T-jump perturbation $\Delta T \theta(t-\tau)$, which in some cases may be easier to implement experimentally. For the purpose of consistency with the discussion in Secs. IV and V, here we consider impulsive temperature perturbations. Since $\delta(t-\tau) = -d\theta(t-\tau)/d\tau$, GRFs of T-jump perturbations can be easily obtained from impulsive GRFs by integration over the corresponding perturbation times.

In Fig. 4, we compare 2D generalized response functions obtained from numerical simulations of an overdamped diffusion process x(t) in non-equilibrium experiments with impulsive perturbations of temperature. We consider symmetric double well $U(x) = x^2 + 3e^{-x^2}$ and asymmetric double well $U(x) = x^2 + 3e^{-x^2} - 0.5x$ potentials and also harmonic $U(x) = x^2/2$ and Morse $U(x) = 10(1 - \exp(-0.3x))^2$ potentials for comparison. The observable in the GRFs was taken to be the stochastic variable itself $f_1 = f_2 = f_3 = x - \langle x \rangle$. First, Fig. 4 shows that 2D GRFs are qualitatively different for

harmonic, Morse, and double-well potentials. In particular, the harmonic system only has a single non-vanishing GRF, $R_{++-}(\tau_2,\tau_1)$. In fact, $R_{+++}(\tau_2,\tau_1)$ and $R_{+--}(\tau_2,\tau_1)$ vanish for both the harmonic and symmetric double-well potentials which is the consequence of symmetry. Second, Fig. 4 shows that both Morse and double well potentials have a strong signal $R_{+-+}(\tau_2,\tau_1)$, which should be an indicator of anharmonicity. Third, the decay of $R_{+-+}(\tau_2, \tau_1)$ along τ_1 in the symmetric and asymmetric double-well potential is significantly slower than the decay of $R_{++-}(\tau_2, \tau_1)$. This is related to the fact that the lowest non-zero eigenvalue of the Fokker-Planck operator (proportional to the Kramers over-the-barrier escape rate) is significantly smaller than the second non-zero eigenvalue in double-well potentials (see also Sec. VIII A). Fourth, 2D $R_{+-+}(\tau_2,\tau_1)$ of a symmetric potential at long times decays with $\tau_1 + \tau_2$, while in asymmetric double-well potentials, the symmetry along the $\tau_1 = \tau_2$ direction is broken. This should provide an experimental measure of the asymmetry of the potential. In Sec. VIII A, we discuss quantitative aspects of these observations.

A. Quantitative information contained in 2D GRFs

The impulsive 2D GRFs given in Eqs. (45)–(48) have the common form

$$R_{+,\alpha,\gamma}(\tau_{2},\tau_{1}) = \int dx_{1}dx_{2}dx_{3}$$

$$\times f(x_{3})P_{0}(x_{3},\tau_{2}+\tau_{1}|x_{2},\tau_{1})$$

$$\times V_{\alpha}(x_{2})P_{0}(x_{2},\tau_{1}|x_{1},0)V_{\gamma}(x_{1})\rho_{ss}(x_{1}), \quad (75)$$

where $\alpha, \gamma = +, -, V_+(x) = f(x)$, and $V_-(x) = \hat{V}(x)$. The transition probabilities $P_0(x,t|x',t')$ can be expressed in terms of eigenvalues λ_n and eigenfunctions $\varphi_n(x)$ of the Fokker-Planck operator,¹⁰

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$$P_0(x,t|x',t') = e^{U(x')/D} \sum_n \varphi_n(x)\varphi_n(x')e^{-\lambda_n(t-t')}$$
$$= \sum_n \varphi_n(x)\psi_n(x')e^{-\lambda_n(t-t')},$$
(76)

where $\psi_n(x') \equiv e^{U(x')/D}\varphi_n(x')$. It should be noted that for symmetric potentials U(x), eigenfunctions $\varphi_n(x)$ are either symmetric or antisymmetric, thus functions $\varphi_{2n}(x)$, $\psi_{2n}(x)$ are even, while $\varphi_{2n+1}(x)$, $\psi_{2n+1}(x)$ are odd.

We simplify Eq. (76) by keeping the three slowest-decaying terms (with the lowest eigenvalues λ_0, λ_1 , and λ_2),

$$P_0(x,t|x',t') = \varphi_0(x)\psi_0(x') + \varphi_1(x)\psi_1(x')e^{-\lambda_1(t-t')} + \varphi_2(x)\psi_2(x')e^{-\lambda_2(t-t')},$$
(77)

since the diffusion process eventually reaches equilibrium, then $\lambda_0 = 0$. Substituting Eq. (77) into Eq. (75), we obtain

$$\begin{aligned} R_{+,\alpha,\gamma}(\tau_{2},\tau_{1}) &= \int f(x_{3})\varphi_{0}(x_{3})dx_{3} \int \psi_{0}(x_{2})V_{\alpha}(x_{2})\varphi_{0}(x_{2})dx_{2} \int \psi_{0}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{1}\tau_{1}} \int f(x_{3})\varphi_{0}(x_{3})dx_{3} \int \psi_{0}(x_{2})V_{\alpha}(x_{2})\varphi_{1}(x_{2})dx_{2} \int \psi_{1}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{2}\tau_{1}} \int f(x_{3})\varphi_{0}(x_{3})dx_{3} \int \psi_{0}(x_{2})V_{\alpha}(x_{2})\varphi_{2}(x_{2})dx_{2} \int \psi_{2}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{1}\tau_{2}} \int f(x_{3})\varphi_{1}(x_{3})dx_{3} \int \psi_{1}(x_{2})V_{\alpha}(x_{2})\varphi_{0}(x_{2})dx_{2} \int \psi_{0}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{1}(\tau_{1}+\tau_{2})} \int f(x_{3})\varphi_{1}(x_{3})dx_{3} \int \psi_{1}(x_{2})V_{\alpha}(x_{2})\varphi_{1}(x_{2})dx_{2} \int \psi_{1}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{1}\tau_{2}-\lambda_{2}\tau_{1}} \int f(x_{3})\varphi_{1}(x_{3})dx_{3} \int \psi_{1}(x_{2})V_{\alpha}(x_{2})\varphi_{2}(x_{2})dx_{2} \int \psi_{2}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{2}\tau_{2}} \int f(x_{3})\varphi_{2}(x_{3})dx_{3} \int \psi_{2}(x_{2})V_{\alpha}(x_{2})\varphi_{0}(x_{2})dx_{2} \int \psi_{0}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{2}\tau_{2}+\lambda_{1}\tau_{1}} \int f(x_{3})\varphi_{2}(x_{3})dx_{3} \int \psi_{2}(x_{2})V_{\alpha}(x_{2})\varphi_{1}(x_{2})dx_{2} \int \psi_{1}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{2}(\tau_{2}+\tau_{1})} \int f(x_{3})\varphi_{2}(x_{3})dx_{3} \int \psi_{2}(x_{2})V_{\alpha}(x_{2})\varphi_{2}(x_{2})dx_{2} \int \psi_{2}(x_{1})V_{\gamma}(x_{1})\rho_{ss}(x_{1})dx_{1} \end{aligned}$$

We first show how Eq. (78) simplifies for symmetric potentials. For an odd observable f = x, only the integrals $\int f(x_3)\varphi_1(x_3)dx_3$ will survive. Integrals over x_1 and x_2 will be non-zero depending on whether $V_{\gamma}(x)$ is symmetric or antisymmetric. For temperature perturbations, $\hat{V}(x) = \frac{\partial^2}{\partial x^2}$ and is therefore symmetric. These considerations reduce Eq. (78) to the following expressions:

$$\begin{aligned} R_{+++}(\tau_{2},\tau_{1}) &= 0, \\ R_{++-}(\tau_{2},\tau_{1}) &= e^{-\lambda_{1}\tau_{2}} \int x_{3}\varphi_{1}(x_{3})dx_{3} \int x_{2}\psi_{1}(x_{2})\varphi_{0}(x_{2})dx_{2} \int \psi_{0}(x_{1})\frac{\partial^{2}}{\partial x_{1}^{2}}\rho_{ss}(x_{1})dx_{1} \\ &+ e^{-\lambda_{1}\tau_{2}-\lambda_{2}\tau_{1}} \int x_{3}\varphi_{1}(x_{3})dx_{3} \int x_{2}\psi_{1}(x_{2})\varphi_{2}(x_{2})dx_{2} \int \psi_{2}(x_{1})\frac{\partial^{2}}{\partial x_{1}^{2}}\rho_{ss}(x_{1})dx_{1}, \\ R_{+-+}(\tau_{2},\tau_{1}) &= e^{-\lambda_{1}(\tau_{1}+\tau_{2})} \int x_{3}\varphi_{1}(x_{3})dx_{3} \int \psi_{1}(x_{2})\frac{\partial^{2}}{\partial x_{2}^{2}}\varphi_{1}(x_{2})dx_{2} \int x_{1}\psi_{1}(x_{1})\rho_{ss}(x_{1})dx_{1}, \\ R_{+--}(\tau_{2},\tau_{1}) &= 0, \end{aligned}$$
(79)

We can now discuss Eqs. (79). First, we have proven that $R_{+++}(\tau_2, \tau_1)$ and $R_{+--}(\tau_2, \tau_1)$ indeed vanish in symmetric potentials. Second, in harmonic potentials, the integral over x_2 in $R_{+-+}(\tau_2, \tau_1)$ reads $\int \psi_1(x_2) \frac{\partial^2}{\partial x_2^2} \varphi_1(x_2) dx_2$ $= \int H_1(x_2) \frac{\partial^2}{\partial x_2^2} H_1(x_2) \exp(-x_2^2/2)$, where $H_1(x)$ is the firstorder Hermite polynomial, and therefore, the whole integral is exactly 0. Thus, $R_{+-+}(\tau_2, \tau_1)$ indeed serves as indicator of anharmonicity, as it was observed in Sec. VIII.

Third, from Eqs. (79), one can see that $R_{++-}(\tau_2, \tau_1)$ decays along τ_1 as $\exp(-\lambda_2\tau_1)$, while $R_{+-+}(\tau_2, \tau_1)$ decays as $\exp(-\lambda_1\tau_1)$. In double-well potentials, especially with high barriers, the first non-zero eigenvalue λ_1 is *much* smaller than the second non-zero eigenvalue λ_2 , which explains the difference in decays of $R_{++-}(\tau_2, \tau_1)$ and $R_{+-+}(\tau_2, \tau_1)$ observed numerically in Sec. VIII. λ_1 is due to the barrier interaction (or possibly "tunneling") splitting of the ground state in

symmetric double well potential (see Ref. 10 on similarity of Schrödinger and Fokker-Planck equations) and is proportional to the barrier height, while the second non-zero eigenvalue λ_2 is the second "energy" level in the local well and is proportional to the width of the well. Thus, in double-well potentials with high energy barriers, $\lambda_1 \ll \lambda_2$. Numerically, $\lambda_1 = B \exp(-E_b/\kappa T)$ (when multiplied by probability to stay in the well it is exactly the Kramers thermally activated escape rate^{14,19}) and one can determine the barrier height E_b by measuring the decay rate λ_1 . We also note that the different decay of $R_{++-}(\tau_2, \tau_1)$ and $R_{+-+}(\tau_2, \tau_1)$ is due to the fact that different GRFs preserve different terms in expression (76). Selectivity property of GRFs was also observed in nondissipative systems.⁸

In weakly asymmetric potentials, we expect the longtime behavior of $R_{++-}(\tau_2, \tau_1)$ and $R_{+-+}(\tau_2, \tau_1)$ to be dominated by the terms in Eqs. (79). Thus, for instance, by measuring



FIG. 4. Numerical 2D GRFs of the overdamped diffusion process x(t) in various typical 1D potentials subjected to impulsive perturbations of temperature.

the eigenvalues λ_1 and λ_2 from the τ_1 -decays of $R_{++-}(\tau_2, \tau_1)$ and $R_{+-+}(\tau_2, \tau_1)$ in Morse-like potentials, one can estimate its anharmonicity.

Fourth, as can be seen from Eq. (79), the GRF $R_{+-+}(\tau_2, \tau_1)$ in symmetric potentials decays as $\exp(-\lambda_1(\tau_1 + \tau_2))$; however, in asymmetric potentials, other terms of Eq. (78) will survive leading to the symmetry breaking along the $\tau_1 = \tau_2$ direction. This was also observed in the numerical results of Sec. VIII.

IX. EXAMPLE 3: CONNECTIVITY OF STOCHASTIC CHEMICAL NETWORKS

GRFs may also be used to deduce the connectivity of stochastic chemical reaction networks in macroscopic chemical kinetics. This can be determined by the arrival time of a perturbation of concentration along the network.²⁰ However, in microscopic systems, such as biological cells, the number of molecules is small and can fluctuate significantly. In this case, the description of reacting species in terms of concentrations is not adequate, and the propagation of chemical perturbation may be obscured by stochastic fluctuations of the number of molecules. Time dependent concentrations are no longer a good measure of chemical dynamics, and other measures, such as GRFs, may help. In Ref. 9, we proposed to use $R_{++-}(\tau_2, \tau_1)$ to check the order of chemical species in the chemical reaction network described by macroscopic rate equations. We now consider the same example but for microscopic kinetics of the model reaction network,

$$A \underset{k_{-1}}{\overset{k_1}{\leftrightarrow}} X_1 \underset{k_{-2}}{\overset{k_2}{\leftrightarrow}} X_2 \underset{k_{-3}}{\overset{k_3}{\leftrightarrow}} X_3 \underset{k_{-4}}{\overset{k_4}{\leftrightarrow}} X_4 \underset{k_{-5}}{\overset{k_5}{\leftrightarrow}} B.$$
(80)

The concentrations of species A and B are held constant by buffering, and each elementary reaction step can be either of linear (first order) or nonlinear (e.g., Michaelis-Menten) kinetics. Suppose that system (80) is initially in a steady state. We then impulsively change the number of molecules of species X_i by ΔN_i at time t_1 and measure the number of molecules N_j , N_l of species X_j and X_l , respectively, at times t_2 and t_3 . In this way, we can construct 2D GRF $R_{++-}(\tau_2, \tau_1)$ $= \frac{\partial}{\partial \Delta N_i} \langle N_l(t_3)N_j(t_2) \rangle$, where $t_1 = 0$, $t_2 = \tau_1$, and $t_3 = \tau_1 + \tau_2$, and averaging is done over repetitions of the experiment.

One can numerically simulate the latter experiment by using the chemical Langevin equations (CLEs), which are often used to simulate microscopic chemical kinetics.^{21,22} The major difference of the CLE approach as compared to the macroscopic chemical kinetics is the appearance of stochastic noise terms in kinetic equations. The CLEs corresponding to reaction network in Eq. (80) have the following form:

$$\frac{dX_1}{dt} = k_1 A - (k_{-1} + k_2) X_1 + k_{-2} X_2 - \sqrt{(k_{-1} + k_2) X_1} \xi_1(t) + \sqrt{k_{-2} X_2} \xi_2(t),$$

$$\frac{dX_2}{dt} = k_2 X_1 - (k_{-2} + k_3) X_2 + k_{-3} X_3 + \sqrt{k_2 X_1} \xi_1(t) - \sqrt{(k_{-2} + k_3) X_2} \xi_2(t) + \sqrt{k_{-3} X_3} \xi_3(t),$$

$$\frac{dX_3}{dt} = k_3 X_2 - (k_{-3} + k_4) X_3 + k_{-4} X_4 + \sqrt{k_3 X_2} \xi_2(t) - \sqrt{(k_{-3} + k_4) X_3} \xi_3(t) + \sqrt{k_{-4} X_4} \xi_4(t),$$

$$\frac{dX_4}{dt} = k_4 X_3 - (k_{-4} + k_5) X_4 + k_{-5} B + \sqrt{k_4 X_3} \xi_3(t) - \sqrt{(k_{-4} + k_5) X_4} \xi_4(t),$$
(81)

where $\langle \xi_i(t)\xi_j(t') \rangle = \sigma \delta_{ij}\delta(t-t')$, and numerical values of parameters are given in Ref. 9. We numerically solve Eqs. (81) and calculate $R_{l+,j+,i-}(\tau_2,\tau_1) = \frac{\partial}{\partial\Delta N_i} \langle N_l(t_3)N_j(t_2) \rangle$ via the non-equilibrium approach for different i, j, l. The typical behavior of $X_1(t), X_2(t), X_3(t)$, and $X_4(t)$ is shown in Fig. 5(a), while the calculated GRFs $R_{l+,j+,i-}(\tau_2,\tau_1)$ are shown in Fig. 5(b). The 2D plots in Fig. 5(b) look the same as in Fig. 2 of Ref. 9. As discussed in that reference, these 2D plots can suggest the order of X_i, X_l , and X_j species in the chemical network, i.e., out of two permutations $X_i - X_l - X_j$ and $X_i - X_j - X_l$, the one with longer correlation tail along the τ_2 axis corresponds to the correct order of species X_i, X_l , and X_j . In particular, from Fig. 5(b), it follows that species X_1, X_2, X_3 , and X_4 are ordered as $X_1 - X_2 - X_3 - X_4$.

GRFs provide constraints with characteristic signatures that may allow to distinguish between possible reaction paths. An open question for future studies is the possibility to invert data to get the reaction topology. The same problem arises in multidimensional NMR (inverting distances between protons obtained from 2D data to get molecular structure), multidimensional IR to get secondary structure of proteins, and multidimensional electronic spectroscopy to get energy transfer paths in chromophore aggregates. Research on relation of GRFs to reaction topology is currently underway. Our preliminary results suggest that GRFs can be used to distinguish kinetic orders of chemical reactions, to determine kinetic parameters of elementary steps in complex chemical reactions, to identify feedback loops in reaction networks, etc.

X. DISCUSSION

Multidimensional techniques have now become standard tools in spectroscopy. However, these are not limited to optical pulses and electronic or vibrational transitions. One can use similar multidimensional ideas with different types of perturbations, to study arbitrary dynamical systems. In this paper, we considered stochastic dynamical systems and showed that their dynamics can be studied via perturbations of their coordinates or parameters, such as temperature. Moreover, perturbations and measurements can be mixed leading to new types of multidimensional measures of stochastic dynamics, which we call generalized response functions. Generalized response functions can be measured experimentally either in time or the frequency domains. Information contained in GRFs can provide additional insights on stochastic microscopic processes. In particular, the examples given in this paper showed that GRFs can reveal the symmetry or bistability of



FIG. 5. Simulations for the stochastic chemical reaction network of Eq. (80). (a) Stochastic concentrations, i.e., solutions of Eq. (81). (b) Generalized response functions $R_{l+,j+,i-}(\tau_2,\tau_1)$.

the mean field potential of stochastic process and can provide information on its anharmonicity and internal barrier height.

We also propose to use multiple impulsive perturbations of temperature to study conformational dynamics of single proteins. Conformational dynamics of proteins plays an important role in regulating their biological functions, but is not well understood. Temperature perturbations provide an accurate and controllable way to probe conformational dynamics. Multiple perturbations and measurements provide a wide range of new tools—experimentally measured GRFs, which entirely depend on equilibrium fluctuations of protein, and thus serve as multidimensional measures of its stochastic conformational dynamics.

GRFs may be also used to study chemical dynamics of small chemical systems such as biocells. Such finite chemical systems contain a small number of molecules of reacting species and thus cannot be described via macroscopic chemical kinetics. A stochastic description is more appropriate. Because of the small number of molecules, fluctuations will be significant and could hide any response to external perturbations. Time-dependent averaged quantities, such as correlation and response functions, are thus required to extract information about the underlying microscopic chemical dynamics. The proposed GRF approach of mixed perturbations and measurements provides a wide range of new multi-point correlation functions, or measures, characterizing microscopic chemical dynamics. Information contained in these new measures should be explored for specific systems. In particular, in Sec. IX, we showed how 2D GRFs $R_{++-}(\tau_2, \tau_1)$ can be used to study connectivity of stochastic reaction networks.

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