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# Relevance of sampling schemes in light of Ruelle's linear response theory

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#### Abstract

We reconsider the theory of the linear response of non-equilibrium steady states to perturbations. We first show that by using a general functional decomposition for space-time dependent forcings, we can define elementary susceptibilities that allow to construct the response of the system to general perturbations. Starting from the definition of SRB measure, we then study the consequence of taking different sampling schemes for analysing the response of the system. We show that only a specific choice of the time horizon for evaluating the response of the system to a general time-dependent perturbation allows to obtain the formula first presented by Ruelle. We also discuss the special case of periodic perturbations, showing that when they are taken into consideration the sampling can be fine-tuned to make the definition of the correct time horizon immaterial. Finally, we discuss the implications of our results in terms of strategies for analyzing the outputs of numerical experiments by providing a critical review of a formula proposed by Reick.

#### 1 Introduction

The study of how the properties of general non-equilibrium statistical mechanical systems change when considering a generic perturbation, usually related to variations either in the value of some internal parameters or in the external forcing, is of great relevance, both in purely

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mathematical terms and with regards to applications to the natural and social sciences. Whereas in quasi-equilibrium statistical mechanics it is possible to link the response of a system to perturbations to its unforced fluctuations thanks to the fluctuation-dissipation theorem [9, 28], in the general non-equilibrium case it is not possible to frame rigorously an equivalence between internal fluctuations and forcings. At a fundamental level, this is closely related to the fact that forced and dissipative systems feature a singular invariant measure. Whereas natural fluctuations of the system are restricted to the unstable manifold, because, by definition, asymptotically there is no dynamics along the stable manifold, perturbations will induce motions - of exponentially decaying amplitude - out of the attractor with probability one, as discussed in, e.g., [24, 22, 13]. It is worth noting that Lorenz anticipated some of these ideas when studying the difference between free and forced variability of the climate system [12]. This crucial difficulty inherent to out-of-equilibrium systems is lifted if the external perturbation is, rather artificially, everywhere tangent to the unstable manifold, or if the system includes some stochastic forcing, which smooths out the resulting the invariant measure [10].

Recently, Ruelle [21, 22, 24] paved the way to the study of the response of general non-equilibrium systems to perturbations by presenting rigorous results leading to the formulation of a response theory for Axiom A dynamical systems [20], which possess a Sinai-Ruelle-Bowen invariant measure [26]. Given a measurable observable of the system, the change in its expectation value due to an  $\epsilon$ -perturbation in the flow (or in the map, in the case of discrete dynamics) can be written as a perturbative series of terms proportional to  $\epsilon^n$ , where each term of the series can be written as the expectation value of some well-defined observable over the unperturbed state. Ruelle's formula is identical to Kubo's classical formula [8] when a Hamiltonian system is considered [13].

Whereas Axiom A systems are mathematically non-generic, the applicability of the Ruelle theory to a variety of actual models is supported by the so-called *chaotic hypothesis* [6], which states that systems with many degrees of freedom behave as if they were Axiom A systems when macroscopic statistical properties are considered. The chaotic hypothesis has been interpreted as the natural extension of the classic ergodic hypothesis to non-Hamiltonian systems [5].

In the last decade great efforts have been directed at extending and clarifying the degree of applicability of the response theory for non-equilibrium systems along five main lines:

- extension of the theory for more general classes of dynamical systems [4, 2];
- introduction of effective algorithms for computing the response in dynamical systems with many degrees of freedom [1], in order to support the numerical analyses pioneered by [19, 3];
- investigation of the frequency-dependent response the susceptibility for the linear and nonlinear cases, with the ensuing introduction of a new theory of Kramers-Kronig relations and sum

rules for non-equilibrium systems [13, 25] supported by numerical experiments [14];

- study of the response to external perturbations of non-equilibrium systems undergoing stochastic dynamics [17, 27];
- use of Ruelle's response theory to study the impact of adding stochastic forcing to otherwise deterministic systems [15].

In particular, the response theory seems especially promising for tackling notoriously complex problems such as those related to studying the response of geophysical systems to perturbations, which include the investigation of climate change; see discussions in [1, 14, 16]. In particular, in [16], it is discussed that by deriving from the linear susceptibility the time-dependent Green function, it is possible to devise a strategy to compute climate change for a general observable and for a general time-dependent pattern of forcing. Recently, response theory is becoming of great interest also in social sciences such as economics [7].

When developing a response theory, there are two possible ways to frame the temporal impact of the additional perturbation to the dynamics. Either one considers the impact at a given time t of a perturbation affecting the system since a very distant past, or one considers the impact in the distant future of perturbations starting at the present time. When deriving the response formula, Ruelle takes the first approach and delivers the *correct* formula [21]. Taking a different point of view and considering the specific case of periodic perturbations – which, anyway, tell us the whole story about the response by linearity –, Reick [19] derives a formula that is well suited for analyzing the output of numerical experiments [14, 16].

Given the great relevance and increasing popularity in applications of the response theory introduced by Ruelle, in this paper, we reconsider the theory of the linear response of non-equilibrium steady states to perturbations and try to bridge rigorously the theoretical derivations and the strategies for designing numerical experiments and analyzing efficiently their outputs.

In Section 2, we study the relevance of the choice of the time horizon for evaluating the impact of the perturbation and we demonstrate why the Ruelle approach is the correct one. We clarify some of the assumptions implicitly considered in his derivations. We then discuss the special case of periodic perturbations, showing that using them as basis for a response theory greatly simplifies the formulas and the conditions under which the formulas are derived. In Section 3, we discuss the implications of our results in terms of strategies for improving the quality of numerical simulations and of the analysis of their output signals and reconsider Reick's formula [19]. In Section 4 we present our conclusions and perspectives for future work.

### 2 Linear Response Theory, revised

#### 2.1 Separable perturbations

We study the linear response of a discrete dynamical system to general time-dependent perturbations. All calculations are formal, in the sense that we neglect all higher orders in the perturbation without deriving an estimate for these terms and we assume that all sums converge in all senses necessary.

The unperturbed dynamical system is given by

$$x_{t+1} = f(x_t) \,,$$

with  $t \in \mathbb{Z}$ ,  $x_t \in M$ , M being a smooth manifold and  $f: M \to M$  a differentiable map. For simplicity we consider a time-independent unperturbed dynamics, although the following can be extended to a time-dependent case in a straightforward manner. Moreover, the analysis of the case of a continuous time flow  $\dot{x} = f(x)$  is perfectly analogous to what is presented in the following and the main corresponding results will be mentioned in Appendix A.

The dynamical system is perturbed by a time-dependent forcing X(t,x) as follows:

$$\tilde{x}_{t+1} = \tilde{f}_{t+1}(\tilde{x}_t) := f(\tilde{x}_t) + X(t+1, f(\tilde{x}_t)).$$
 (1)

The effect of the perturbation on individual trajectories is in general difficult to describe. More can however be said about the statistical properties of the system. One can look at the expectation values of observables under invariant states of the dynamical system:

$$\rho(A) := \int \rho(dx) A(x) \,,$$

where  $\rho(dx)$  is an invariant measure of the unperturbed dynamics i.e.

$$\rho(A \circ f) = \rho(A)$$
.

for any observable A. In general, a dynamical system can possess many invariant measures. The physically relevant measure for dynamical systems is the SRB measure [26]. This measure is physical in the sense that for a set of initial conditions of full Lebesgue measure the time averages  $\lim_{t\to\infty}\frac{1}{t}\sum_{k=1}^t A(f^k(x))$  converge to the expectation value under  $\rho$ . In other words, for any measure l(dx) that is absolutely continuous w.r.t. Lebesgue, we have that

$$\rho(A) = \lim_{t \to \infty} \frac{1}{t} \sum_{k=1}^{t} \int l(dx) A(f^{k}(x)).$$
 (2)

We want to determine the linear response of expectation values under the SRB measure to perturbations of the dynamical system as in Eq. 1. We denote by  $\delta_T \rho$  the difference in the expectation value between the perturbed and unperturbed system at time T. In [24],

Ruelle presents a formula for the linear response due to perturbations that are separable in time and space:

$$X(t,x) = \phi(t)\chi(x)$$
.

The leading order term of the expansion of  $\delta_T \rho(A)$  in X is given by

$$\delta_T \rho(A) \approx \sum_{j \in \mathbb{Z}} G_A(j) \phi(T-j),$$
 (3)

with

$$G_A(j) = \theta(j) \int \rho(dx) \chi(x) D(A \circ f^j)(x) , \qquad (4)$$

where  $\theta$  is the Heaviside function. Since  $\delta_T \rho(A)$  is expressed as a convolution product of  $G_A$  and  $\phi$ , the Fourier transform of the response  $\delta_\omega \rho(A) = \sum_{T \in \mathbb{Z}} e^{iT\omega} \delta_T \rho(A)$  is given by a product of the Fourier transform  $\hat{\phi}(\omega)$  of the time factor  $\phi(t)$  and a susceptibility function  $\hat{\kappa}_A(\omega)$ :

$$\delta_{\omega}\rho(A) \approx \hat{\kappa}_A(\omega)\hat{\phi}(\omega)$$
. (5)

where

$$\hat{\phi}(\omega) = \sum_{j \in \mathbb{Z}} e^{ij\omega} \phi(j) ,$$

$$\hat{\kappa}_A(\omega) = \sum_{j \in \mathbb{Z}} e^{ij\omega} G_A(j)$$

$$= \sum_{j \geq 0} e^{ij\omega} \int \rho(dx) \chi(x) D(A \circ f^j)(x) .$$
(6)

Due to the causality of the response function  $G_A(j)$  (i.e.  $G_A(j) = 0$ , j < 0)), the susceptibility  $\hat{\kappa}_A(\omega)$  is analytic in the upper complex plane and satisfies Kramers-Kronig relations [22, 13].

#### 2.2 General perturbations

If the perturbation is of a more general nature (i.e. not separable), we can deduce a linear response formula from Eq. 5, solely based on linearity in the following way. Let  $\phi_r(t)$  be a Schauder basis [11] of time-dependent functions and  $\psi_s(x)$  a Schauder basis of space-dependent functions. One can take for example the Fourier basis in time and a wavelet basis in space, or whatever basis may be suitable for the system at hand. The product functions  $\phi_r(t)\psi_s(x)$  then form a basis of the time and space dependent functions as a tensor product [18]. More concretely, we may for an appropriate sense of convergence assume that any function X(t,x) can be decomposed in the product basis  $\phi_r(t)\psi_s(x)$  with coefficients  $a_{r,s}$ :

$$X(t,x) = \sum_{r,s>0} a_{r,s} \phi_r(t) \psi_s(x).$$

Since each of the factors in this sum is separable, we can use Eq. 5 and the linearity of the response to get that the response is given by

$$\delta_{\omega}\rho(A) \approx \sum_{r,s>0} a_{r,s}\hat{\phi}_r(\omega)\hat{\kappa}_{s,A}(\omega),$$
 (7)

where  $\hat{\kappa}_{s,a}$  is the susceptibility function of observable A, corresponding to the forcing pattern given by  $\psi_s(x)$ . Since the vectors  $\psi_s(x)$  constitute a basis, the functions  $\hat{\kappa}_{s,a}$  are elementary linear susceptibilities that allow to construct the response of the system to any pattern of forcing.

By inserting the expression of  $\hat{\kappa}_{s,A}(\omega)$  from Eq. 6 into Eq. 7, it is possible to deduce the frequency-dependent response of the system. It is expressed as an ensemble average of a dot product of Fourier transforms, namely the transforms of the perturbation term and of the linear tangent of the observable,  $\Gamma_A(\omega, x)$ :

$$\delta_{\omega}\rho(A) \approx \sum_{r,s\geq 0} a_{r,s}\hat{\phi}_{r}(\omega) \sum_{j\geq 0} e^{ij\omega} \int \rho(dx)\psi_{s}(x)D(A\circ f^{j})(x)$$

$$= \sum_{j\geq 0} e^{ij\omega} \int \rho(dx)\hat{X}(\omega,x)D(A\circ f^{j})(x)$$

$$= \int \rho(dx)\hat{X}(\omega,x)\Gamma_{A}(\omega,x), \qquad (8)$$

with

$$G_A(\omega, x) = \sum_{j \ge 0} e^{ij\omega} D(A \circ f^j)(x)$$

$$\hat{X}(\omega, x) = \sum_{T \in \mathbb{Z}} X(T, x) e^{i\omega T}$$

$$= \sum_{r,s \ge 0} a_{r,s} \hat{\phi}_r(\omega) \psi_s(x). \tag{9}$$

Instead from Eqs. 3-4 in the time domain

$$\delta_T \rho(A) \approx \int \rho(dx) \sum_{j \ge 0} X(T - j, x) D(A \circ f^j)(x).$$
 (10)

For the case of a periodic perturbation  $X(t + \tau, x) = X(t, x)$ , we get as linear response

$$\delta_{T}\rho(A) \approx \int \rho(dx) \sum_{n=1}^{\tau} \sum_{m=0}^{\infty} X(T+n+m\tau,x) D(A \circ f^{n+m\tau})(x)$$

$$= \int \rho(dx) \sum_{n=1}^{\tau} X(T+n,x) \sum_{m=0}^{\infty} D(A \circ f^{n+m\tau})(x)$$

$$= \int \rho(dx) \sum_{n=1}^{\tau} X(T+n,x) G_{A,n}(x), \qquad (11)$$

with

$$G_{A,n}(x) = \sum_{m=0}^{\infty} D(A \circ f^{n+m\tau})(x).$$

In order to elucidate some crucial aspects of the Ruelle's response theory, we now propose a direct derivation of the linear response to the perturbation X(t,x) by considering the history of the perturbed and unperturbed trajectory of the system and verify under which conditions we find agreement with Eqs. 8-10. Our goal is to derive the leading order term of the expansion of  $\delta_T \rho(A)$  with respect to X from first principle, i.e. without resorting to the Schauder decomposition as above. Such a derivation should of course arrive at the same results as those in Eqs. 8-10.

#### 2.2.1 Response at a moving time horizon

We describe the perturbed measure  $\tilde{\rho}_T(A)$  such that the system is initialized at time T in an initial condition according to the measure l. We move the time horizon at which we observe forward and average the time-evolved measurements. The system is prepared and then observed while it is evolving over a sufficiently long time. The measure  $\tilde{\rho}_T$  is time-dependent as the dynamics  $\tilde{f}$  is also time-dependent. Formally we take  $\tilde{\rho}_T$  to be the ergodic mean of the expectation values of A, starting at time T:

$$\tilde{\rho}_T(A) = \lim_{t \to \infty} \frac{1}{t} \sum_{k=1}^t \int l(dx) A(\tilde{f}_T^k(x)). \tag{12}$$

Here l(dx) is an initial measure that is absolutely continuous with respect to Lebesgue and  $\tilde{f}_T^k$  represents k iterations of the perturbed dynamics from time T to T+k:

$$\tilde{f}_T^k(x) = \tilde{f}_{T+k} \circ \dots \circ \tilde{f}_{T+1}(x). \tag{13}$$

The difference in expectation values  $\delta_T \rho$  is the given by

$$\delta_T \rho(A) = \tilde{\rho}_T(A) - \rho(A). \tag{14}$$

Following the computation presented in [24] for the separable case, we can expand the perturbed dynamics  $\tilde{f}$  around the unperturbed dynamics f. We then try to rewrite the response of the perturbed system in terms of the SRB measure of the unperturbed system by finding an expression for  $A(\tilde{f}_{T}^{k}(x))$  in terms of  $A(f^{k}(x))$ .

We can approximate up to first order in X the two time step future evolution by expanding around the unperturbed dynamics  $f^2(x)$ :

$$\tilde{x}_{T+2} = \tilde{f}_{T+2} \circ \tilde{f}_{T+1}(\tilde{x}_T) \approx f^2(\tilde{x}_T) + X(T+1, f(\tilde{x}_T)) \cdot Df(f(\tilde{x}_T)) + X(T+2, f^2(\tilde{x}_T)) .$$

For k time steps we similarly get:

$$\tilde{x}_{T+k} = \tilde{f}_{T+k} \circ \dots \circ \tilde{f}_{T+1}(\tilde{x}_T)$$

$$\approx f^k(\tilde{x}_T) + \sum_{j=1}^k X(T+j, f^j(\tilde{x}_T)) \cdot (Df^{k-j})(f^j(\tilde{x}_T)).$$

Thus, we can approximate  $A(\tilde{f}_{T+k} \circ \ldots \circ \tilde{f}_{T+1}(x))$  to first order in X as follows:

$$A(\tilde{f}_{T+k} \circ \dots \circ \tilde{f}_{T+1}(x)) \approx A(f^{k}(x))$$

$$+ A'(f^{k}(x)) \left( \sum_{j=1}^{k} X(T+j, f^{j}(x)) \cdot (Df^{k-j})(f^{j}(x)) \right)$$

$$= A(f^{k}(x))$$

$$+ \sum_{j=1}^{k} X(T+j, f^{j}(x)) D(A \circ f^{k-j})(f^{j}(x)) .$$
(15)

The linear response of A is obtained by substituting Eq. 15 into Eq. 14, through Eq. 2 and Eq. 12:

$$\delta_T \rho(A) \approx \lim_{t \to \infty} \frac{1}{t} \sum_{k=1}^t \int l(dx) \sum_{j=1}^k X(T+j, f^j(x)) D(A \circ f^{k-j})(f^j(x))$$

$$= \lim_{t \to \infty} \frac{1}{t} \sum_{j=1}^t \sum_{i=0}^{t-j} \int l(dx) X(T+j, f^j(x)) D(A \circ f^i)(f^j(x))$$

$$= \lim_{t \to \infty} \frac{1}{t} \sum_{i=0}^{t-1} \sum_{j=1}^{t-i} \int l(dx) X(T+j, f^j(x)) D(A \circ f^i)(f^j(x)).$$

Using that for  $i \geq t$  the expression is zero, we have

$$\delta_T \rho(A) \approx \sum_{i \ge 0} \int \left( \lim_{t \to \infty} \frac{1}{t} \sum_{j=1}^{t-i} (f^*)^j l(dx) X(T+j, x) \right) D(A \circ f^i)(x).$$
(16)

Note that it is not possible to rewrite the sum in j as the ergodic time mean of l due to the time dependence of the perturbation X(T+j,x). Therefore, surprisingly, Eq. 16 does not in general agree with Eq. 10. In particular, by taking the limit on the right hand side, we obtain that the T-dependence disappears. Say we shift T to T-T' in the

limit appearing in the above equation:

$$\lim_{t \to \infty} \frac{1}{t} \sum_{j=1}^{t-i} (f^*)^j l(dx) X(T - T' + j, x)$$

$$= \lim_{t \to \infty} \frac{1}{t} \sum_{j'=1-T'}^{t-i-T'} (f^*)^{j'+T'} l(dx) X(T + j', x)$$

$$= \lim_{t \to \infty} \frac{1}{t} \sum_{j'=1}^{t-i} (f^*)^{j'} \left( (f^*)^{T'} l(dx) \right) X(T + j', x) .$$

Taking the reasonable assumption that the result in Eq. 16 does not depend on the initial measure l(dx) (this cannot be obtained from the uniqueness of the SRB measure), the obtained response of the system is time-independent even if the forcing is time-dependent.

Let us compare the result contained in Eq. 16 with Eq. 3 in the special case of a time-independent perturbation  $X(t,x) = \chi(x)$ . Now Eq. 16 and Eq. 3 agree since Eq. 16 simplifies to:

$$\delta_T \rho(A) \approx \sum_{i>0} \int \rho(dx) \chi(x) D(A \circ f^i)(x) ,$$

because  $\lim_{t\to\infty} 1/t \sum_{j=1}^{t-i} (f^*)^j l(dx) = \rho(dx)$ , by the definition of the SRB measure. The formula given by Ruelle [22] is recovered, as can be seen by substituting  $\phi(t) = 1$  into Eq. 3.

However, already in the case of a time-periodic perturbation

$$X(t,x) = X(t+\tau,x)$$

there is no agreement between Eq. 16 and Eq. 10. In this case the sum over j appearing in Eq. 16 can be written as a double sum, one over k periods, indexed by m, and one over the  $\tau$  phases in each period, indexed by n:

$$\lim_{t \to \infty} \frac{1}{t} \sum_{j=1}^{t} (f^*)^j l(dx) X(T+j,x)$$

$$= \lim_{k \to \infty} \frac{1}{k\tau} \sum_{m=1}^{k} \sum_{n=1}^{\tau} (f^*)^{m\tau} (f^*)^n l(dx) X(T+n,x)$$

$$= \frac{1}{\tau} \sum_{n=1}^{\tau} \rho_n(dx) X(t+n,x), \qquad (17)$$

with

$$\rho_n(dx) = \lim_{k \to \infty} \frac{1}{k} \sum_{m=1}^k (f^*)^{m\tau} (f^*)^n l(dx).$$

Under the assumption that  $\rho_n = \rho$  for all  $n \in \{1, ..., \tau\}$ , the response gives a similar result as the Ruelle formula, but with an averaged perturbation. Substituting Eq. 17 into Eq. 16, we obtain a formula of

the form of Eq. 10, with the difference that instead of the true forcing X(t,x) the averaged forcing

$$\frac{1}{\tau} \sum_{n=1}^{\tau} X(t+n, x)$$

appears in the formula. The disagreement is apparent, e.g. when one considers a perturbation of the form  $X(t,x) = \sin(\frac{2\pi l}{\tau}t)\chi(x)$ , which obviously results in a zero response.

One way to obtain agreement with Formula 11 is to choose a specific sampling procedure. We sample with the same periodicity of the forcing, thus altering the definition of the response. We define the measures for the perturbed and unperturbed system as

$$\tilde{\rho}'_{T}(dx) := \lim_{N \to \infty} \frac{1}{N} \sum_{k=1}^{N} (\tilde{f}_{T}^{k\tau+p})^{*} l(dx)$$

$$\rho'(dx) := \lim_{N \to \infty} \frac{1}{N} \sum_{k=1}^{N} (f_{T}^{k\tau+p})^{*} l(dx),$$

for  $p \in \{0, \dots, \tau - 1\}$ . With this definition we obtain Eq. 11:

$$\delta \rho_T' = \tilde{\rho}_T' - \rho'$$

$$\approx \lim_{N \to \infty} \frac{1}{N} \sum_{k=1}^N \int l(dx) \sum_{n=1}^\tau \sum_{m=1}^k X(T + n + m\tau, f^{n+m\tau}(x))$$

$$D(A \circ f^{k\tau - n - m\tau})(f^{n+m\tau}(x))$$

and by the periodicity of X

$$\delta \rho_T' \approx \sum_{n=1}^{\tau} \lim_{N \to \infty} \frac{1}{N} \sum_{m=0}^{N-1} \sum_{i=1}^{N-m} \int (f^{n+m\tau})^* l(dx) X(T+n,x) D(A \circ f^{i\tau-n})(x)$$

$$= \sum_{n=1}^{\tau} \lim_{N \to \infty} \frac{1}{N} \sum_{i \ge 1} \sum_{m=0}^{N-i} \int (f^{n+m\tau})^* l(dx) X(T+n,x) D(A \circ f^{i\tau-n})(x)$$

$$= \sum_{n=1}^{\tau} \int \left( \lim_{N \to \infty} \frac{1}{N} \sum_{m=0}^{N-i} (f^{n+m\tau})^* l(dx) \right) X(T+n,x) \left( \sum_{i \ge 1} D(A \circ f^{i\tau-n})(x) \right)$$

$$= \sum_{n=1}^{\tau} \int \rho_n(dx) X(T+n,x) \left( \sum_{i \ge 1} D(A \circ f^{i\tau-n})(x) \right).$$

Still, this situation is unsatisfactory. Why do we get a time-independent response? Why do we only get the correct result by taking periodic perturbations and fine-tuning the sampling or by taking constant perturbations?

#### 2.2.2 Response at a fixed time horizon

This paradox can be resolved by defining the time-dependent SRB measure in Eq. 12 using a different method of sampling. We now

consider the time evolution  $\tilde{f}_T^k$  in this definition to go from time T-k in the past up to the fixed time horizon T, so instead of Eq. 13, we have:

$$\tilde{f}_T^k = \tilde{f}_T \circ \dots \circ \tilde{f}_{T-k} \,. \tag{18}$$

Note that this approach does not use the reversed time dynamics but rather a different time perspective in which the final time is fixed as the current time and the perturbation starts in the remote past.

The expansion to first order in X around the dynamics of f now becomes:

$$\tilde{x}_T = \tilde{f}_T \circ \dots \circ \tilde{f}_{T-k+1}(\tilde{x}_{T-k})$$

$$\approx f^k(\tilde{x}_{T-k}) + \sum_{j=0}^{k-1} X(T-j, f^{k-j}(\tilde{x}_{T-k})) \cdot (Df^j)(f^{k-j}(x_{T-k})).$$

Hence, the linear response of  $\rho(A)$  at time T is given by

$$\delta_T \rho(A) \approx \lim_{t \to \infty} \frac{1}{t} \sum_{k=1}^t \int l(dx) \sum_{j=0}^{k-1} X(T-j, f^{k-j}(x)) D(A \circ f^j) (f^{k-j}(x))$$

$$= \lim_{t \to \infty} \frac{1}{t} \sum_{j=0}^{t-1} \sum_{i=1}^{t-j} \int l(dx) X(T-j, f^i(x)) D(A \circ f^j) (f^i(x))$$

$$= \lim_{t \to \infty} \frac{1}{t} \sum_{j \ge 0} \int \sum_{i=1}^{t-j} (f^*)^i l(dx) X(T-j, x) D(A \circ f^j) (x).$$

Note that in contrast to Eq. 16 the indices are such that the time average of the measure and the perturbation are decoupled. This crucially depends on the choice of the sampling. This allows us to use the definition of the SRB measure in Eq. 2 and replace the time average in the limit by  $\rho$ :

$$\delta_T \rho(A) \approx \sum_{j \ge 0} \int \rho(dx) X(T - j, x) D(A \circ f^j)(x). \tag{19}$$

This agrees with Eq. 10.

Here we do get the anticipated result. Note that this expression gives also a non-zero response for a perturbation which is non-zero for a finite time as opposed to Eq. 16.

This sampling is the natural one fore deducing the general linear response theory. Doing the calculation for constant forcing does not elucidate the relevance of the choice of sampling. This sampling corresponds to a Gedanken experiment where the system is prepared in the distant past and we observe the difference of the perturbed and unperturbed evolution up to a given instant T.

#### 3 Numerics

For perturbations that are separable  $(X(t,x) = \phi(t)\chi(x))$  and have a single driving frequency  $\Omega$   $(\phi(t) = \epsilon cos(\Omega t))$ , the following sampling scheme for computing the susceptibility for a given observable A has been proposed by Reick [19]:

$$\hat{\kappa}_A(\Omega) = \lim_{\epsilon \to 0} \lim_{N \to \infty} \frac{1}{N\epsilon} \sum_{t=1}^N e^{i\Omega t} \int \rho(dx) \left( A(\tilde{f}_0^t(x)) - A(f^t(x)) \right). \tag{20}$$

This formula has been later adopted to analyze the output of a simple climate model [16] and a generalization has been proposed to study the nonlinear susceptibilities describing harmonic generation [14]. Applicability of this formula depends on performing numerical experiments where the initial samples approximate the unperturbed SRB measure  $\rho$ .

Using our previous calculations, we want to circumstantiate the validity of the formula. We apply Ruelle's response theory to obtain a perturbative expression of Eq. 20 in terms of quantities of the unperturbed dynamics.

$$\begin{split} &\lim_{N\to\infty}\frac{1}{N}\sum_{t=1}^N e^{i\Omega t}\int l(dx)\left(A(\tilde{f}_0^t(x))-A(f^t(x))\right)\\ &\approx \lim_{N\to\infty}\frac{1}{N}\sum_{t=1}^N e^{i\Omega t}\int \sum_{j=1}^t (f^j)^*l(dx)X(j,f^j(x))D(A\circ f^{t-j})(x)\,. \end{split}$$

Here we encounter the same problem as in Eq. 16, namely the coupling of the averages of the measure and the perturbation. Indeed, using Reick's formula sampling from an initial measure different from the unperturbed SRB measure, one does not get a reasonable response, as reported in [16].

By sampling according to the unperturbed SRB measure  $\rho$  instead of l, the above equation becomes

$$\lim_{N \to \infty} \frac{1}{N} \sum_{j=1}^{N} \sum_{k=0}^{N-j} e^{i\Omega k} e^{i\Omega j} \int \rho(dx) X(j,x) D(A \circ f^k)(x)$$

$$= \lim_{N \to \infty} \frac{1}{N} \int \rho(dx) G_A(\omega, x) \sum_{j=1}^{N-k} e^{i\Omega j} X(j, x). \tag{21}$$

We insert the inverse discrete time Fourier transform

$$X(j,x) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \hat{X}(\omega,x) e^{-i\omega j} d\omega$$

into Eq. 21:

$$\int \rho(dx)\Gamma_{A}(\omega,x) \lim_{N \to \infty} \frac{1}{N} \sum_{j=1}^{N} e^{i\Omega j} X(j,x)$$

$$= \int \rho(dx)\Gamma_{A}(\omega,x) \lim_{N \to \infty} \frac{1}{N} \sum_{j=1}^{N} e^{i\Omega j} \frac{1}{2\pi} \int_{-\pi}^{\pi} \hat{X}(\omega,x) e^{-i\omega j} d\omega$$

$$= \int \rho(dx)\Gamma_{A}(\omega,x) \lim_{N \to \infty} \frac{1}{2\pi} \int_{-\pi}^{\pi} \hat{X}(\omega,x) \frac{1}{N} \sum_{j=1}^{N} e^{i(\Omega-\omega)j} d\omega. \tag{22}$$

The sum over j can be rewritten by making use of

$$x + \ldots + x^N = x \frac{1 - x^N}{1 - x}$$
.

as

$$\frac{1}{N}e^{i(\Omega-\omega)j}\frac{1-e^{iN(\Omega-\omega)j}}{1-e^{i(\Omega-\omega)j}},$$

which converges to 0 as N goes to infinity, except for  $\Omega = \omega$ . At  $\Omega = \omega$  the sum over j gives N. Hence, if  $\hat{X}(\omega, x)$  is continuous, we can take the limit in Eq. 22 inside the integral

$$\frac{1}{2\pi} \int_{-\pi}^{\pi} \hat{X}(\omega, x) \mathbf{1}_{\{\Omega\}}(\omega) d\omega = 0,$$

where  $\mathbf{1}_{\{\Omega\}}$  is the indicator function on  $\{\Omega\}$ . We deduce that in the case of a general perturbation with a continuous Fourier spectrum, Reick's numerical approach cannot be applied. Note also that for finite time steps N (as is always the case for numerical experiments), there is an additional broadening of the signal of order 1/N, as is apparent from Eq. 22.

If on the other hand the Fourier transform is singular, for example

$$X(\omega, x) = \delta(\Omega - \omega)\chi(x),$$

which corresponds to the monochromatic signal

$$X(j,x) = \frac{1}{2\pi} e^{-i\Omega j} \chi(x) \,,$$

we have that

$$\lim_{N \to \infty} \frac{1}{N} \sum_{j=1}^{N-k} e^{i\Omega j} X(j, x) = \lim_{N \to \infty} \frac{N-k}{2\pi N} \chi(x)$$
$$= \frac{\chi(x)}{2\pi}.$$

Such that Eq. 21 becomes

$$\int \rho(dx)G_A(\omega,x)\chi(x) = \hat{\kappa}_A(\Omega),$$

as predicted by Reick. The above calculation gives a straightforward derivation of Reicks formula.

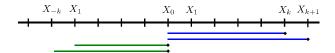


Figure 1: The different methods of calculating the linear response. In blue: the moving time horizon, in green: the fixed time horizon.

#### 4 Conclusion

In this paper we have reconsidered Ruelle's linear response theory by analyzing the impact of different methods of sampling which correspond to different choices of time horizon in our thought experiment.

The general response formula is obtained by choosing a specific sampling where the system is prepared in the distant past and we observe the difference of the perturbed and the unperturbed dynamics up to a given time T.

The other possible sampling strategy, where the time horizon is not fixed, does not give rise to a natural response theory except e.g. for constant perturbations. In the case of periodic forcings one can obtain a meaningful formula by redefining appropriately the response, finely tuned to the forcing under investigation.

Thanks to our approach we get a deeper understanding of the range of applicability of Reicks formula, which has been used as a signal processing tool to study the linear response of numerical models.

By proposing a general decomposition of space-time dependent forcings, we have elucidated that it is possible to define *elementary* linear susceptibilities that allow to construct the response of the system to any pattern of forcing.

We clarify that it is crucial in practical terms to use an ensemble approach where the initial conditions sample approximately the unperturbed SRB measure. Our calculation is explicitly performed for discrete time, but the analogous results for continuous time are presented in Appendix A. Moreover our considerations seem to be appropriate also for the case of nonlinear response [23, 14].

Our approach explains why considering the fixed time horizon it is possible to analyze a response to forcings that have a continuous Fourier spectrum. The clarifications presented in this paper may be of relevance for devising the data processing for actual laboratory experiments on nonlinear systems.

Future work will address the investigation of the response of a non-equilibrium system to a general random field. Moreover, we will analyze the impact of the various sampling schemes described in this paper when studying the output of numerical models.

## A Continuous time response formulas

Here we give the formulas for continuous time systems corresponding to the ones presented in the main text. The time evolution is in this setting given by a differential equation

$$\frac{dx}{dt} = F(x),$$

resulting in a flow  $x(t+s) = f^s(x(t))$ . The SRB measure is given by

$$\rho(A) = \lim_{t \to \infty} \frac{1}{t} \int_0^t ds \int l(dx) A(f^s(x)).$$

For a separable perturbation

$$\frac{dx}{dt} = F(x) + \chi(x)\phi(t)$$

the susceptibility is given by

$$\hat{\kappa}_A(\omega) = \int_0^\infty dt e^{i\omega t} \int \rho(dx) \chi(x) D(A \circ f^t)(x) \,.$$

In case of a general perturbation  $F(x) \to F(x) + X(t,x)$  the linear response becomes:

$$\delta_T \rho(A) \approx \int_0^\infty d\tau \int \rho(dx) X(T-\tau, x) D(A \circ f^\tau)(x)$$

The SRB measure with a moving time horizon is:

$$\tilde{\rho}_T = \lim_{t \to \infty} \frac{1}{t} \int_0^t ds \int l(dx) A(\tilde{f}_T^{T+t}(x))$$

and the SRB measure with a fixed time horizon:

$$\tilde{\rho}_T = \lim_{t \to \infty} \frac{1}{t} \int_0^t ds \int l(dx) A(\tilde{f}_{T-t}^T(x))$$

where  $\tilde{f}_{t_1}^{t_2}(x)$  is a trajectory of the perturbed system, starting at time  $t_1$  in x and evolving up to time  $t_2$ .

Reick's formula now becomes:

$$\hat{\kappa}_{A}(\omega) = \lim_{\epsilon \to 0} \lim_{\nu \to \infty} \frac{1}{\nu \epsilon} \int_{0}^{\nu} dt e^{i\Omega t} \int \rho(dx) \left( A(\tilde{f}_{0}^{t}(x)) - A(f_{0}^{t}(x)) \right)$$

#### References

- [1] R. Abramov and A.J. Majda. Blended response algorithms for linear fluctuation-dissipation for complex nonlinear dynamical systems. *Nonlinearity*, 20:2793–2821, 2007.
- [2] V. Baladi. On the susceptibility function of piecewise expanding interval maps. *Comm. Math. Phys.*, 275:839–859, 2007.
- [3] B. Cessac and J.-A. Sepulchre. Linear response, susceptibility and resonances in chaotic toy models. *Physica D*, 225:13–28, 2007.

- [4] D. Dolgopyat. On differentiability of SRB states for partially hyperbolic systems. *Invent. Math.*, 155:389–449, 2004.
- [5] G. Gallavotti. In J.P. Francoise, G.L. Naber, and T. S. Tsun, editors, Nonequilibrium Statistical Mechanics (Stationary): Overview, volume 3 of Encyclopedia of Mathematical Physics, pages 530–539. Elsevier Amsterdam, 2006.
- [6] G. Gallavotti and E.G.D. Cohen. Dynamical ensembles in stationary states. J. Stat. Phys., 80:931–970, 1995.
- [7] R.S. Hawkins and M. Aoki. Macroeconomic relaxation: Adjustment processes of hierarchical economic structures. *Economics:* The Open-Access, Open-Assessment E-Journal, 3:1–21, 2009.
- [8] R. Kubo. Statistical-mechanical theory of irreversible processes.
   i. J. Phys. Soc. Jpn., 12:570-586, 1957.
- [9] R. Kubo. The fluctuation dissipation theorem. Rep. Prog. Phys., 29:255–284, 1966.
- [10] G. Lacorata and A. Vulpiani. Fluctuation-response relation and modeling in systems with fast and slow dynamics. *Nonlin. Pro*cesses Geophys., 14:681–694, 2007.
- [11] J. Lindenstrauss and L. Tzafriri. Classical Banach spaces. Springer, 1996.
- [12] E.N. Lorenz. Forced and free variations of weather and climate. J. Atmos. Sci., 36:1367-1376, 1979.
- [13] V. Lucarini. Response theory for equilibrium and non-equilibrium statistical mechanics: Causality and generalized Kramers-Kronig relations. *J. Stat. Phys.*, 131:543–558, 2008.
- [14] V. Lucarini. Evidence of dispersion relations for the nonlinear response of Lorenz 63 system. J. Stat. Phys., 134:381–400, 2009.
- [15] V. Lucarini. Stochastic perturbations to dynamical systems: a response theory approach. *Physical Review E. submitted*, 2011.
- [16] V. Lucarini and S. Sarno. A statistical mechanical approach for the computation of the climatic response to general forcings. *Non-lin. Processes Geophys.*, 18:7–28, 2011.
- [17] A.J. Majda and X. Wang. Linear response theory for statistical ensembles in complex systems with time-periodic forcing. *Comm. Math. Sci.*, 8:142–172, 2010.
- [18] M. Reed and B. Simon. Modern Methods of Mathematical Physics. Vol. 1. Functional Analysis. Academic Press, New York, 1972.
- [19] C. H. Reick. Linear response of the Lorenz system. *Physical Review E*, 66(3):036103, 2002.
- [20] D. Ruelle. Chaotic evolution and strange attractors. Cambridge University Press Cambridge, 1989.
- [21] D. Ruelle. Differentiation of SRB states. Communications in Mathematical Physics, 187(1):227–241, 1997.

- [22] D. Ruelle. General linear response formula in statistical mechanics, and the fluctuation-dissipation theorem far from equilibrium. *Phys. Letters A*, 245:220–224, 1998.
- [23] D. Ruelle. Nonequilibrium statistical mechanics near equilibrium: computing higher-order terms. *Nonlinearity*, 11:5–18, 1998.
- [24] D. Ruelle. A review of linear response theory for general differentiable dynamical systems. *Nonlinearity*, 22(4):855–870, 2009.
- [25] A. Shimizu. Universal properties of nonlinear response functions of nonequilibrium steady states. *Journal of the Physical Society of Japan*, 79:113001, 2010.
- [26] L.-S. Young. What are SRB measures, and which dynamical systems have them? *Journal of Statistical Physics*, 108:733–754, 2002. 10.1023/A:1019762724717.
- [27] T. Yuge. Sum rule for response function in nonequilibrium Langevin systems. *Phys. Rev. E*, 82:051130, 2010.
- [28] D.N. Zubarev. *Nonequilibrium Statistical Thermodynamics*. Consultant Bureau New York, 1974.