# Chaos suppression in the parametrically driven Lorenz system

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(Received 2 November 2004; revised manuscript received 9 June 2005; published 8 September 2005)

We predict theoretically and verify experimentally the suppression of chaos in the Lorenz system driven by a high-frequency periodic or stochastic parametric force. We derive the theoretical criteria for chaos suppression and verify that they are in a good agreement with the results of numerical simulations and the experimental data obtained for an analog electronic circuit.

DOI: 10.1103/PhysRevE.72.036206

PACS number(s): 05.45.Gg, 05.45.Ac, 82.40.Bj

### I. INTRODUCTION

Control of the chaotic dynamics of complex nonlinear systems is one of the most important and rapidly developing topics in applied nonlinear science. In particular, the concept of chaos control, first introduced in Ref. [1], has attracted a great deal of attention over the past decade. Among different methods for controlling chaos that have been suggested by now, the so-called nonfeedback control is attractive because of its simplicity: no measurements or extra sensors are required. The idea of this method is to change the complex behavior of a nonlinear stochastic system by applying a properly chosen external force. It is especially advantageous for ultrafast processes, e.g., at the molecular or atomic level, where no possibility to measure the system variables exists.

Many of the suggested nonfeedback methods employ external forces acting at the system frequencies [2], including parametric perturbations with the frequency that is in resonance with the main driving force. In particular, changing the phase and frequency of a parameter perturbation in a bistable mechanical device was shown to either decrease or increase the threshold of chaos [3]. Similarly, unstable periodic orbits are known to be stabilized by a low-frequency modulation of a system control parameter [4], when the control frequency is much lower than the system frequency.

However, the chaos suppression by means of small periodic perturbations [2] will be efficient when the frequency of these periodic perturbations exactly coincides with some well-defined system frequency. For autonomous chaotic systems, such resonances are hard to find and successful suppression of chaos may only be achieved by trial and error. The possibility to change significantly the system dynamics by applying a high- (rather than low-) frequency force is known for almost a century. As a textbook example, we mention the familiar stabilization of a reverse pendulum (known as the Kapitza pendulum) by rapid vertical oscillations of its pivot [5]. This discovery triggered the development of vibrational mechanics [6] where the general analysis of nonlinear dynamics in the presence of rapidly varying forces is based on the Krylov-Bogoljubov averaging method [7]. In the control theory, the high-frequency forces and parameter modulations are usually used for the vibrational control of nonchaotic nonlinear systems [8]. However, as was shown for the example of the Duffing oscillator [9,10], chaos suppression can also be achieved by applying a high-frequency parametric force. Later, chaos suppression in the Belousov-Zhabotinsky reaction was demonstrated numerically by adding white noise [11], and the effect of random parametric noise on a periodically driven damped nonlinear oscillator was studied by the Melnikov analysis [12].

The suppression of chaos by high-frequency parametric modulations has a little in common with other feedback chaos control techniques which recently became popular for stabilizing unstable periodic orbits. Those techniques are either highly sophisticated and can only be applied to very slow systems [1], or they suffer from various other constraints, e.g., missing torsion, control latency, vanishing basins of attraction, etc. [13]. In contrast, the mechanism of high-frequency parametric modulation is very general, and it can be understood in terms of an effective renormalization of the modulated system parameters. What in fact is observed when starting from a chaotic state and increasing the modulation amplitude is a debifurcation route "out of the chaos." So the actual modulation amplitude needed to suppress the chaos depends not only on the specific system, but also on the sensitivity of the chaotic state on the specific parameter modulated and, accordingly, on the distance of its present value from the threshold for chaos and on the final state desired. The advantages of this technique, however, are its simplicity, universality and robustness, no detailed preknowledge, modeling, fine-tuning [2] or highly sophisticated realtime analysis [1] of the system is needed, but the price is generally a larger modulation amplitude.

In this paper we apply the concepts of nonresonant nonfeedback control to the Lorenz system and demonstrate analytically, numerically, and also experimentally that the suppression of the chaotic dynamics can be achieved by applying a high-frequency parametric or random parametric force. The Lorenz system, found in 1963, is known to produce a canonical chaotic attractor in a simple threedimensional autonomous system [14,15], and it can be applied to describe many interesting nonlinear systems, ranging from thermal convection [16] to laser dynamics [17].

The paper is organized as follows. In Sec. II we present our model and outline the theoretical method and results. By applying the averaging method, we derive the effective Lorenz equations with the renormalized control parameter and obtain the conditions for chaos suppression. In Sec. III, we demonstrate the suppression of chaos by means of direct numerical simulations and also present the experimental data obtained for an analog electronic circuit. Finally, Sec. IV concludes the paper.

## **II. THEORETICAL APPROACH**

### A. Model

We consider the familiar Lorenz system [14] driven by a parametric force. In dimensionless variables, the nonlinear dynamics is governed by the equations

$$\dot{x} = \sigma(y - x),$$
  

$$\dot{y} = r[1 + f(t)]x - y - xz,$$
  

$$\dot{z} = xy - bz,$$
(1)

where the dots stand for the derivatives in time,  $\sigma$ , r, and b are the parameters of the Lorenz model, and the function f(t) describes a parametric force. For definiteness but without restrictions of generality, we select the standard set of the parameter values,  $\sigma = 10$  and b = 8/3, whereas the parameter r is assumed to vary. It is well known [14] that, in the absence of the parametric driving force f(t), the original Lorenz equations demonstrate different dynamical regimes on variation of the control parameter r, which are associated with the existence and stability of several equilibrium states. In brief, the system dynamics can be characterized by three regimes.

The  $G_1$  regime, for r < 1: there exists the only stable fixed point at the origin, (x, y, z) = (0, 0, 0).

The  $G_2$  regime, for 1 < r < 24.74: the origin becomes unstable, two new fixed points appear,  $(x, y, z) = (\pm \sqrt{b(r-1)}, \pm \sqrt{b(r-1)}, r-1)$ .

The  $G_3$  regime, for r > 24.74: no stable fixed points exist, chaotic dynamics occur with a strange attractor.

We consider the special case of the general model (1), assuming that the characteristic frequency of the parametric force f(t) is much larger than the characteristic frequency of the unforced Lorenz system, where the frequency (which is the mean-time derivative of the phase) of the Lorenz system can be defined as [18]

$$\omega_0 = \lim_{T \to \infty} \frac{2\pi N(T)}{T},$$
(2)

where N(T) is the number of turns performed in T.

### B. Periodic driving force

First, we consider the case of a periodic driving force,

$$f(t) = k\cos(\omega t). \tag{3}$$

Assuming that the frequency of parametric modulations is large ( $\omega \ge \omega_0$ ), we apply an asymptotic method [5,6] based on a separation of different time scales, and derive the effective equations that describe the slowly varying dynamics. To

do this, we follow Ref. [9] and decompose every variable into a sum of slowly and rapidly varying parts, i.e.,

$$x = X + \xi, \quad y = Y + \eta, \quad z = Z + \zeta, \tag{4}$$

where the functions  $\xi(t)$ ,  $\eta(t)$ , and  $\zeta(t)$  describe fast oscillations around the slowly varying envelope functions X(t), Y(t), and Z(t), respectively. The rapidly oscillating corrections are assumed to be small in comparison with the slowly varying parts, and their mean values during an oscillation period vanish, i.e.,  $\langle x \rangle = X$ ,  $\langle y \rangle = Y$ , and  $\langle z \rangle = Z$ . Substituting the Eq. (4) into Eq. (1) with the force (3), and averaging over the oscillation period, we obtain the equation (cf. Ref. [9]),

$$X = \sigma(Y - X),$$
  
$$\dot{Y} = rX - Y - XZ + rk\langle \xi \cos(\omega t) \rangle,$$
  
$$\dot{Z} = XY - bZ,$$
 (5)

where the terms  $\langle \xi \zeta \rangle$  and  $\langle \xi \eta \rangle$  are neglected because, for large  $\omega$ , they are all of higher orders in  $k\omega^{-1}$ . Using Eq. (5) and keeping only the terms not smaller than those of the order of  $k\omega^{-1}$  and k, respectively, we find the equations,  $\dot{\xi} = \sigma \eta$  and  $\dot{\eta} = rkX \cos(\omega t)$ . Regarding the function X as constant during the period of the function f(t), we obtain the solution  $\xi = -\sigma r(k/\omega^2)X \cos(\omega t)$ , and thus  $\langle \xi \cos(\omega t) \rangle =$  $-\sigma r(k/2\omega^2)X$ . Therefore, the averaged equations (5) are

$$X = \sigma(Y - X),$$
  
$$\dot{Y} = r_{\rm eff}X - Y - XZ,$$
  
$$\dot{Z} = XY - bZ,$$
 (6)

where

$$r_{\rm eff} = r(1 - rK_{\omega}), \quad K_{\omega} = \frac{\sigma k^2}{2\omega^2} > 0.$$
<sup>(7)</sup>

As a result, the averaged dynamics of the Lorenz system in the presence of a rapidly varying parametric force is described by the effective renormalized Lorenz equations (6) with the effective control parameter  $r_{\rm eff}$ , so that all dynamical regimes and the route to chaos discussed earlier can be applicable directly to Eq. (6), assuming the effect of the renormalization.

Therefore, in terms of the averaged system, for  $r_{\rm eff} \leq 1$ , i.e., under the condition

$$K_{\omega} \ge (r-1)/r^2,\tag{8}$$

the fixed point at the origin remains stable and, therefore, in terms of the original Lorenz system, the chaotic dynamics should be suppressed. Next, for  $1 < r_{\text{eff}} \leq 24.74$ , i.e., under the condition

$$(r - 24.74)/r^2 \le K_{\omega} < (r - 1)/r^2,$$
 (9)

the Lorenz chaos is also suppressed and the stable saddlefocus points appear:  $X_0 = Y_0 = \pm \sqrt{b(r_{\text{eff}} - 1)}, Z_0 = r_{\text{eff}} - 1.$  Relations (8) and (9) that follow from the averaged equations, define the conditions for the chaos suppression; they can be expressed as relations between the amplitude and frequency of the rapidly varying oscillations,

$$k \ge \omega \sqrt{2(r-1)/\sigma r^2},\tag{10}$$

$$\omega\sqrt{2(r-24.7)/\sigma r^2} \le k < \omega\sqrt{2(r-1)/\sigma r^2}.$$
 (11)

The dependencies  $K_{\omega} \sim r^{-1}$  [see Eqs. (8) and (9)] and  $k \sim \omega$  [see Eqs. (10) and (11)] are the key characteristics of the chaos suppression effect that we verify numerically and compare with the experimental data obtained with an analog electronic circuit (see Figs. 4 and 5 below).

#### C. Random driving force

Now we turn to the case of a *random force*, and treat the function f(t) in Eq. (1) as random by formally writing  $f(t) = \epsilon(t)$ , where  $\epsilon(t)$  describes a bandwidth-limited noise with a power spectral density

$$P_{\epsilon}(\omega) = \begin{cases} p(\omega), & \text{when } \omega_1 < |\omega| < \omega_2, \\ 0, & \text{when } |\omega| < \omega_1, \ |\omega| > \omega_2, \end{cases}$$
(12)

and the zero mean value,  $\langle \epsilon(t) \rangle = 0$ . In order to apply the analytical method discussed above, the noise  $\epsilon(t)$  is assumed to be composed of high-frequency components only, i.e.,  $\tilde{\omega} \gg \omega_0$ , where

$$\widetilde{\omega} = \int_{-\infty}^{\infty} \omega P_{\epsilon}(\omega) d\omega \left[ \int_{-\infty}^{\infty} P_{\epsilon}(\omega) d\omega \right]^{-1}$$

is the characteristic frequency of the bandwidth-limited parametric fluctuations. Again, decomposing the system variables into sums of slow and rapidly varying functions and averaging over the period  $T_0$ , we obtain

$$\dot{X} = \sigma(Y - X),$$
  
$$\dot{Y} = rX - Y - XZ + rk\langle \xi(t)\epsilon(t) \rangle,$$
  
$$\dot{Z} = XY - bZ,$$
 (13)

where the value of  $T_0$  is much smaller than the characteristic time scale of the Lorenz system oscillations but sufficiently larger than the time scale of fluctuations, i.e.,  $2\pi/\tilde{\omega} < T_0$  $\ll 2\pi/\omega_0$ . Then, the equations for the rapid parts can be reduced to the equations,  $\dot{\xi} = \sigma \eta(t)$  and  $\dot{\eta} = r\epsilon(t)X$ . Thus the key averaged quantity becomes  $\langle \xi(t)\epsilon(t)\rangle = \langle \xi(t)\ddot{\xi}(t)\rangle/(\sigma rX)$ . On the other hand, for the stationary random process  $\xi(t)$ , the following expressions are valid,

$$\langle \xi(t)\ddot{\xi}(t)\rangle = -\langle \dot{\xi}^2(t)\rangle = -\int_{-\infty}^{\infty} \omega^2 P_{\xi}(\omega)d\omega$$

where  $P_{\xi}(\omega) = |H(\omega)|^2 P_{\epsilon}(\omega)$  is the power spectral density of  $\xi(t)$  and  $H(\omega) = -\sigma r X/\omega^2$  is the frequency response function for the equation  $\ddot{\xi} = \sigma r X \epsilon(t)$ . As a result, we obtain the aver-

aged equations in the form (6) where this time [cf. Eq. (7)]

$$r_{\rm eff} = r(1 - rK_{\epsilon}), \quad K_{\epsilon} = 2\sigma \int_{\omega_1}^{\omega_2} \frac{p(\omega)}{\omega^2} d\omega.$$
 (14)

The quantity  $K_{\epsilon}$  is related to the intensity of the parametric fluctuations  $\epsilon(t)$  and it is always positive. Since the averaged equations have the same form as the original unforced Lorenz equations but with the effective control parameter  $r_{\rm eff}$  instead of r, there appears the parameter region where the chaotic dynamics is suppressed. In particular, for  $r_{\rm eff} \leq 1$ , i.e., under the condition

$$K_{\epsilon} \ge (r-1)/r^2,\tag{15}$$

the fixed point at the origin remains stable in the presence of rapidly varying fluctuations. When  $1 < r_{\text{eff}} \le 24.74$ , i.e., for

$$(r-24.74)/r^2 \le K_{\epsilon} < (r-1)/r^2,$$
 (16)

two new saddle-focus fixed points appear, with the coordinates  $X_0 = Y_0 = \pm \sqrt{b(r_{\text{eff}} - 1)}, Z_0 = r_{\text{eff}} - 1.$ 

When the random function  $\epsilon(t)$  describes a bandwidthlimited white noise,  $P_{\epsilon}(\omega) = S_0$ , the renormalization factor  $K_{\epsilon}$  can be written as

$$K_{\epsilon} = \frac{2\sigma(\omega_2 - \omega_1)}{\omega_1 \omega_2} S_0 = \frac{\sigma\langle \epsilon^2(t) \rangle}{\omega_1 \omega_2}, \tag{17}$$

and the effective control parameter is simplified to be

$$r_{\rm eff} = r \bigg( 1 - \frac{\sigma r}{\omega_1 \omega_2} \langle \epsilon^2(t) \rangle \bigg), \tag{18}$$

so that the chaos suppression effect is clearly proportional to the noise intensity. Moreover, in the limit  $\omega_1, \omega_2 \rightarrow \omega$ , we recover the result  $K_{\epsilon} = \sigma k^2 / (2\omega^2)$  obtained earlier for the periodic parametric force [see Eq. (7)].

## III. NUMERICAL SIMULATIONS AND EXPERIMENTAL RESULTS

To verify our theory, first we perform direct numerical simulations of the full model (1). Figures 1 and 2 show the results for the system temporal evolution obtained numerically by solving Eqs. (1) and (3) with the parameters:  $(\sigma, b, r) = (10, 8/3, 28)$ . Since the mean frequency of the unforced oscillations of the Lorenz system is found to be  $\omega_0$ =8.24 [see Eq. (2)], for the above parameters the frequency of the rapid parametric force is chosen as  $\omega = 70$ . As shown in the examples presented in Fig. 1, the fixed point at the origin x=0 is stabilized for k=6.5 whereas a new stationary point  $X_0 = 6.8$  appears for k = 3.5. We can see from Fig. 2 that the system dynamics in phase space after the suppression of chaos is reduced to coexisting limit cycles (marked in solid and dotted) whose averaged behavior is described by Eq. (6), shrinking to the origin as the parametric force amplitude k is increased.

Next, we study experimentally the chaos suppression in the Lorenz system in the framework of an analog electronic

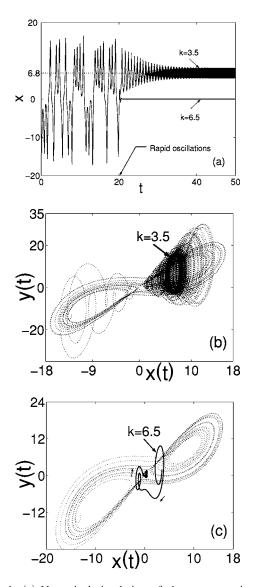


FIG. 1. (a) Numerical simulation of chaos suppression shown for the evolution of the function x(t). The high-frequency parametric force is turned on at the point t=20, and it moves the system to a new fixed point  $\langle x(t) \rangle = 6.8$ , for k=3.5, or it stabilizes the origin x(t)=0, for k=6.5. (b), (c) Phase plane representation of the system dynamics for (b) k=3.5 and (c) k=6.5.

circuit implementation of the system (see, e.g., Refs. [19,20]). Our circuit uses three op-amps (TL084) as the building blocks for the operations of sum, difference, and integration, and three analog multipliers (AD633) for the operations of products.

The variables *x*, *y*, and *z* are represented by the respective op-amp output voltages (see Fig. 3) all normalized to 100 mV. Time is measured in units of  $\tau = R_6C = 0.1$  s. Such a normalization guarantees that all variables fit within the dynamical range of the source (-15 V, 15 V) and that the circuit operates in a frequency range of a few Hz. The parameters of the Lorenz model can be defined through the values of the resistors,  $\sigma = R_6/R_1$ ,  $b = R_6/R_8$ ,  $r = R_6/R_3$ , and the values of the resistors used are  $R_1 = R_2 = 100 \text{ k}\Omega$ ,  $R_3 = 36 \text{ k}\Omega$  (for  $r \approx 28$ ),  $R_4 = R_7 = 10 \text{ k}\Omega$ ,  $R_5 = 5.1 \text{ k}\Omega$ ,  $R_6 = 1 \text{ M}\Omega$ , and  $R_8$ = 374 k $\Omega$  with a tolerance of less than 1%. The value of the

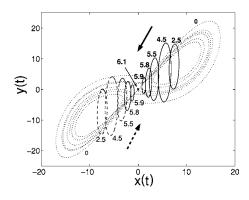


FIG. 2. Phase plane representation of the system dynamics after the suppression of chaos, at different values of the parametric force amplitude (marked in the figure).

capacitors is C=100 nF. For these values we obtain  $\sigma=10$ and b=2.67. The control parameter r of the Lorenz system can be adjusted by varying the resistor  $R_3$ , whereas the force amplitude k is determined by the input signal amplitude normalized to  $(R_5/2R_3)V=71$  mV (for r=28). The frequency of the unforced oscillations for r=28 is found to be  $\omega_0=2\pi$  $\times 13.1$  Hz. So, applying a parametric force with a frequency larger than 65 Hz fulfills the fast modulation condition.

Figure 4 shows the threshold curves described by Eqs. (8) and (9), as well as Eqs. (15) and (16) that define three regions with different nonlinear dynamics, as compared with the numerical and experimental data. The domains  $G_1$  and  $G_2$  indicate the regions for stabilizing the fixed point at the origin and the transition to a new fixed point, respectively, whereas the domain  $G_3$  is the region of the chaotic dynamics. The critical values of  $K_{\omega}$  were experimentally obtained by changing the amplitude k of the parametric driving force at the fixed frequency  $\omega = 2\pi \times 1$  kHz. The corresponding experimental data are presented in Fig. 4 by squares and diamonds. In the experiment with parametric fluctuations (the data are marked by asterisks), we applied bandwidth-limited white noise with the limits  $\omega_1 = 2\pi \times 0.7$  kHz and  $\omega_2 = 2\pi \times 1.3$  kHz by means of an external noise generator.

Figure 5 shows the dependence of the critical amplitude of the parametric force for the chaos suppression as a func-

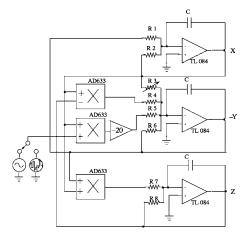


FIG. 3. Schematic of the analog electric circuit representing the Lorenz oscillator, including the parametric driving force and fluctuations.

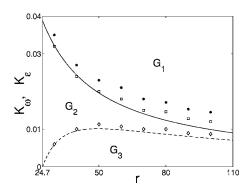


FIG. 4. Experimental vs theoretical results shown as the stability regions in the parameter plane  $(K_{\omega}/K_{\epsilon}, r)$ . Solid: theoretical result [Eq. (8) and Eq. (15)] for stabilization of the fixed point at the origin. Dashed: the theoretical results (9) and (16) for the transition to a new oscillating state. Squares and diamonds: experimental results for the periodic driving force with the frequency  $\omega = 2\pi \times 1$  kHz. Asterisks: the critical values  $K_{\epsilon}$  for the chaos suppression by a bandwidth-limited white noise with  $\omega_1 = 2\pi \times 0.7$  kHz and  $\omega_2 = 2\pi \times 1.3$  kHz.

tion of the frequency. Solid and dashed lines represent the theoretical results described by Eq. (10) and Eq. (11), respectively, whereas the experimental data are plotted as squares and diamonds. As shown in Fig. 4 and Fig. 5, the experimental results are in good agreement with the theoretically calculated threshold dependencies.

### **IV. CONCLUSIONS**

We have studied analytically, numerically, and experimentally the suppression of chaos in the Lorenz system driven by a rapidly oscillating periodic or random parametric force. We

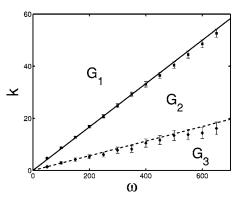


FIG. 5. Experimental vs theoretical results for the critical amplitude of the parametric force vs frequency for r=28. The analytical results (10) and (11) are plotted as solid and dashed lines, respectively. Squares and diamonds show the experimental results for the stabilization of the origin and the suppression of chaos into a fixed point, respectively. Here, the dimensionless units k=1 and  $\omega = 1$  correspond to an input signal amplitude of 71 mV and to a technical frequency of 1.59 Hz, respectively.

have derived theoretical criteria for chaos suppression which indicate that, for a fixed value of the control parameter r, the critical amplitude of the force required for the suppression of chaos is proportional to its frequency. The theoretical criteria for chaos suppression have been found to agree well with both the results of numerical simulations and the experimental data obtained for an analog electronic circuit.

## ACKNOWLEDGMENTS

C.-U.C. and Y.S.K thank the Alexander von Humboldt-Stiftung for support. The work was also supported by the Deutsche Forschungsgemeinschaft.

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