

## Decoding information by following parameter modulation with parameter adaptive control

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It has been proposed to realize secure communication using chaotic synchronization via transmission of a binary message encoded by parameter modulation in the chaotic system. This paper considers the use of parameter adaptive control techniques to extract the message, based on the assumptions that we know the equation form of the chaotic system in the transmitter but do not have access to the precise values of the parameters which are kept secret as a secure set. In the case in which a synchronizing system can be constructed using parameter adaptive control by the transmitted signal and the synchronization is robust to parameter mismatches, the parameter modulation can be revealed and the message decoded without resorting to exact parameter values in the secure set. A practical local Lyapunov function method for designing parameter adaptive control rules based on originally synchronized systems is presented. [S1063-651X(99)07206-2]

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### I. INTRODUCTION

Chaotic dynamics, which have noiselike broadband power spectra, are interesting candidates for encoding and masking information signals in communication. Most approaches proposed to realize this basic idea of using a chaotic signal as broadband carrier are based on the synchronization of coupled chaotic systems [1]. Several schemes have been proposed so far: (I) *chaotic masking* [2,3], where the message is added directly to the chaotic carrier with an amplitude much lower than the chaotic carrier; (II) *chaotic modulation* [4–6], where the message is injected into the chaotic system to modulate the chaotic carrier, and (III) *chaotic switching* [3,7,8], where a binary message is transmitted by switching between two chaotic attractors associated with two sets of parameters of the system.

Intuitively, the communication is expected to be secure based on two considerations: (i) it is difficult to read out the hidden message by any spectral analysis due to the broadband nature of the chaotic carrier, and (ii) exact knowledge of the parameters of the system in the receiver is necessary to recover the message. Thus, a set of the system parameters which serve as the encryption key is not accessible by any third party.

However, recently some researchers have shown that the security may be spoiled, not by access to the secure set of the parameters, but by some other approaches. For the communication schemes (I) and (II), it has been shown that the hidden message may be unmasked with some nonlinear dynamical forecasting methods [9,10]. It is believed that this weakness in security is due to low dimension and a single positive Lyapunov exponent of the chaotic carrier, and the suggestion is to employ hyperchaotic systems, such as coupled chaotic arrays [6,11] or time-delay systems [12,13] in communication. This, however, may not produce drastic improvement in the security, as shown in a recent report [14] that messages masked by hyperchaos of a six-dimensional system can be unmasked only with a three-dimensional reconstruction, and in our recent work [15] demonstrating that messages masked by chaos of time-delay systems with very high dimension and many positive Lyapunov exponents can

also be extracted successfully.

Another work [16] has also shown that hidden messages can be extracted from a chaotic carrier without reconstructing the full dynamics, but using some suitable return maps, which is successfully applied to the Lorenz system for communication schemes I and III.

The idea of encoding by parameter modulation is to use two chaotic attractors  $\mathcal{A}$  and  $\mathcal{B}$  to represent the two symbols of the digital signals [3,7,8]. Since  $\mathcal{A} \neq \mathcal{B}$ , it is possible to construct some suitable return maps which can distinguish the differences between the two attractors, thus reading out the message, just as shown in Ref. [16]. However, if the two attractors are rather complex or the differences between them are very subtle, it may be very difficult to find such distinguishable return maps.

It is natural to ask if it is possible for a motivated intruder to follow the parameter modulation in the transmitter using some parameter adaptive control by the transmitted signal. This paper carries out security analysis of communication systems using the encoding scheme III. Our analysis is based on the following assumptions.

(a) The intruder does not have access to the precise value of any system parameters in the secure set.

(b) The intruder does know the functional form of the chaotic system in the transmitter.

Our results will show that if a synchronizing system can be constructed using parameter adaptive control by the transmitted signal and the synchronization is robust to parameter mismatches, the messages may be decoded without resorting to the exact parameter values. Since it is practically possible for a motivated intruder to locate a region close to the exact parameters based on the knowledge of the system and the character of the transmitted signal, the security may be spoiled. Robustness of the synchronization to parameter mismatches is an advantage for implementation of the communication scheme but may be a weakness in the security.

### II. DECODING BY PARAMETER ADAPTIVE CONTROL

Let us consider the following transmitter system:

$$\frac{d\mathbf{x}}{dt} = \mathbf{F}(\mathbf{p}, \mathbf{q}, \mathbf{x}), \quad (1)$$

where  $\mathbf{p}$  and  $\mathbf{q}$  are system parameters. Binary messages are encoded by switching  $\mathbf{q}$  between  $\mathbf{q}_1$  and  $\mathbf{q}_2$ . Based on our assumption, a motivated intruder has known the equations of the system and that  $\mathbf{q}$  are modulated for encoding. He tries to construct a decoding system using parameter adaptive control by the transmitted signal  $s = h(\mathbf{x})$ , as

$$\frac{d\mathbf{y}}{dt} = \mathbf{H}(\mathbf{p}_y, \mathbf{q}_y, \mathbf{y}, s), \quad (2)$$

$$\frac{d\mathbf{q}_y}{dt} = \mathbf{G}(\mathbf{y}, \mathbf{q}_y, s)[s - h(\mathbf{y})]. \quad (3)$$

Suppose that this system (called the *intruder* system from now on) is synchronizable with some suitable coupling function  $\mathbf{G}$  if  $\mathbf{p}_y = \mathbf{p}$ .

In general, it is not always possible that one can find a synchronizable intruder system for any transmitted signal  $s$  and any subset  $\mathbf{q}$  of the system parameters. However, if with this transmitted signal  $s$  a synchronizable system  $\mathbf{H}(\mathbf{p}, \mathbf{q}, \mathbf{y}, s)$  can be found by some synchronization schemes, such as Pecora-Carroll decomposition [1], active-passive decomposition [4], or feedback control [17] *without* parameter adaptive control, then we can expect that additional parameter adaptive control loops for parameters  $\mathbf{q}_y$  exist if the synchronization is robust to parameter mismatches to some extent, because the system  $\mathbf{H}$  driven by  $s$  is still stable (the largest conditional Lyapunov exponent is negative) for parameters  $\mathbf{q}_y$  in the vicinity of the point  $\mathbf{q}_y = \mathbf{q}$ , although exact synchronization is spoiled by parameter mismatches, and the exact synchronization can be restored by bringing the parameters back to the point  $\mathbf{q}_y = \mathbf{q}$  using some appropriate control methods. For some systems, such parameter control rules can be found by an analysis based on a global Lyapunov function. In general, however, such an analytical treatment may not be possible. In this case, we employ the idea of designing a control rule using the information of a control surface [18] constructed by perturbing the parameters  $\mathbf{q}_y$ . The essence of the idea is that since the synchronization between the systems  $\mathbf{H}$  and  $\mathbf{F}$  before incorporating parameter adaptive control is robust to parameter mismatches, the synchronization behavior changes smoothly when  $\mathbf{q}_y$  deviates slightly from  $\mathbf{q}$ . And there exists a local Lyapunov function with respect to the parameters  $\mathbf{q}_y$  near  $\mathbf{q}$ , with the form

$$E(\mathbf{q}_y) = \mathbf{U}^T \mathbf{U}, \quad (4)$$

where  $\mathbf{U} = (U_1, U_2, \dots, U_k)^T$  are time averages of some functions  $\mathbf{u} = (u_1, u_2, \dots, u_k)^T$  ( $k$  is the number of the components in  $\mathbf{q}_y$ ), i.e.,

$$U_i = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T u_i dt \quad (i = 1, 2, \dots, k). \quad (5)$$

The function  $u_i$  has the form  $u_i = \hat{u}_i(s, \mathbf{y}, \mathbf{q}_y)[s - h(\mathbf{y})]$  so that  $\mathbf{U} = \mathbf{0}$  if  $\mathbf{q}_y = \mathbf{q}$ . In order that  $E(\mathbf{q}_y)$  is a local Lyapunov function, it is required that  $\mathbf{U}$  is smooth with respect to  $\mathbf{q}_y$  near  $\mathbf{q}$ . With this local Lyapunov function, the evolution system

$$\frac{d\mathbf{U}}{dt} = -\alpha \mathbf{U} \quad (6)$$

is stable at  $\mathbf{U} = \mathbf{0}$ .  $\alpha$  is a convergence parameter. Noting that the convergence of  $\mathbf{U}$  is induced by control of the parameters, we have

$$\frac{d\mathbf{q}_y}{dt} = \frac{\partial \mathbf{q}_y}{\partial \mathbf{U}} \frac{d\mathbf{U}}{dt} = -\alpha \frac{\partial \mathbf{q}_y}{\partial \mathbf{U}} \mathbf{U}. \quad (7)$$

In general, it is impossible to obtain an explicit form of  $\mathbf{U}$  for the control rule in Eq. (7). To solve this problem, we use a control surface obtained in simulation or experiment. First we record a time series  $s$  from system  $\mathbf{F}(\mathbf{p}, \mathbf{q}, \mathbf{x})$  with a known set of parameter  $(\mathbf{p}, \mathbf{q})$  in the chaotic regime. Then we perturb the parameter  $\mathbf{q}_y$  in the driven system  $\mathbf{H}(\mathbf{p}, \mathbf{q}_y, \mathbf{y}, s)$  slightly from  $\mathbf{q}$  to some values in its vicinity, and compute  $\mathbf{U}(\mathbf{q}_y)$  as a function of  $\mathbf{q}_y$ . For appropriately chosen function  $\mathbf{u}$ ,  $\mathbf{U}(\mathbf{q}_y)$  are smooth with respect to  $\mathbf{q}_y$  close to the point  $\mathbf{q}$ , and in the vicinity, it can be approximated by

$$\mathbf{U}(\mathbf{q}_y) = M(\mathbf{q}_y - \mathbf{q}), \quad (8)$$

where the constant  $k \times k$  matrix  $M$  is obtained by a local linear fitting of the numerically or experimentally obtained control surface  $\mathbf{U}(\mathbf{q}_y)$ . Now if the initial value of  $\mathbf{q}_y$  is close to  $\mathbf{q}$ , we can replace the Jacobian matrix  $\partial \mathbf{q}_y / \partial \mathbf{U}$  in Eq. (7) by the matrix  $M^{-1}$ , i.e.,

$$\frac{d\mathbf{q}_y}{dt} = -\alpha M^{-1} \mathbf{U}. \quad (9)$$

In practice, one can implement the above control rule by replacing  $\mathbf{U}$  with a time average over a period of time  $\tau$ , e.g.,

$$U_i(t) = \frac{1}{\tau} \int_{t-\tau}^t u_i dt. \quad (10)$$

Often, the parameter modulation in the transmitter is much slower than the time scale of the chaotic system, and one can simplify the control rule by replacing the time average  $U_i$  with  $u_i$ , and finally obtain

$$\frac{d\mathbf{q}_y}{dt} = -\alpha M^{-1} \mathbf{u}. \quad (11)$$

We can expect that with the above additional parameter adaptive control, the synchronization between the systems is maintained with small enough coupling strength  $\alpha$  for  $\mathbf{q}_y$  initially close to  $\mathbf{q}$ , so that the parameters  $\mathbf{q}$  can be recovered. We can also expect that the synchronization is also robust to mismatches of the rest parameters  $\mathbf{p}$ . The function  $u_i$  can be chosen somewhat arbitrarily, as long as  $U_i$  [and thus  $E(\mathbf{q}_y)$ ] is a smooth function of  $\mathbf{q}_y$  in the neighborhood of  $\mathbf{q}$ . This scheme provides a general and practical yet simple way to build additional parameter adaptive control loops for originally coupled and synchronized systems, even though a proper choice of the functions  $u_i$  may still be nontrivial. In this way, the intruder can design systematically an attacking system for the communication scheme based on the knowledge of the system, although such a designed control scheme using local information may not be successful when applied to the signal from the transmitter whose parameters the intruder does not know, especially when the transmitter is operating in a parameter region far away from the one that the intruder uses to build the control rule. However, it is possible for the intruder to get into a neighborhood of the transmitter

TABLE I. Values of the parameters of Chua's circuit in the transmitter.

$C_1$ (nF)	$C_2$ (nF)	$R$ ( $\Omega$ )	$L$ (mH)	$G_0$ (ms)	$G_1$ (ms)	$1/r$ ( $\mu$ s)	$B_p$ (V)
10	100	1680	18	-0.753	-0.396	6	1

parameters using some system identification methods based on the knowledge of the system.

Since the intruder system is quite robust to parameter mismatches, the parameter modulation in the transmitter may be revealed and the message decoded without resorting to the exact values of the transmitter parameters  $\mathbf{p}$ , but within some tolerable neighborhood. In certain cases, it might also be possible to recover all the transmitter parameters ( $\mathbf{p}, \mathbf{q}$ ) by designing adaptive control loops for them all with the above scheme.

In the next sections, we present examples of message decoding based on the above idea. In the first example of Chua's circuit, the control rule is obtained with a global function analysis, and in the second example of the Lorenz system, the control rule is designed with the local Lyapunov function scheme.

### III. EXAMPLE OF CHUA'S CIRCUIT

As an illustration, we carry out analysis on a specific communication system proposed in Ref. [8]. We first give a brief description of the encoding scheme, and then construct a robust intruder system.

#### A. The transmitter

The communication system employs the well-known Chua circuit as the chaotic system. The evolution equations for Chua's circuit are given by

$$C_1 \frac{dx_1}{dt} = \frac{1}{R}(x_2 - x_1) - h(x_1), \quad (12)$$

$$C_2 \frac{dx_2}{dt} = \frac{1}{R}(x_1 - x_2) + x_3, \quad (13)$$

$$L \frac{dx_3}{dt} = -x_3. \quad (14)$$

The nonlinear characteristic of Chua's diode  $h(x_1)$  is given by

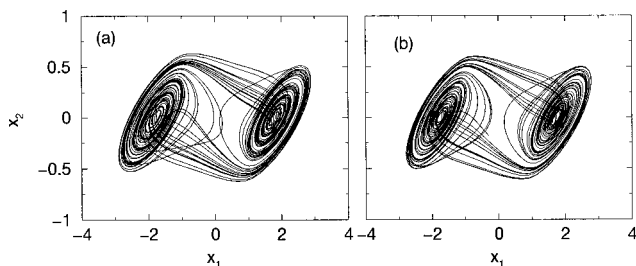


FIG. 1. Two attractors used to encode binary information, with (a) corresponding to bit +1 and (b) to -1.

$$h(x_1) = G_1 x_1 + \frac{1}{2}(G_0 - G_1)[|x_1 + B_p| - |x_1 - B_p|], \quad (15)$$

which is a three-segment piecewise-linear function.

A binary message stream  $I_{\text{in}}$  is encoded by switching between parameters  $G_0$ ,  $G_1$ , and  $G'_0 = G_0 + 1/r$ ,  $G'_1 = G_1 + 1/r$  when the stream switches between +1 and -1, where  $r$  is a resistor in parallel with Chua's diode. The parameters used are shown in Table I. Since  $1/r$  is small (about 1% with respect to  $G_0$  and  $G_1$ ), the two chaotic attractors are very similar, as shown in Fig. 1. To examine the similarity, we construct return maps using the consecutive maxima  $x_{\text{max}}(n)$  and  $x_{\text{max}}(n+1)$  from the transmitter signal  $x_1$ , as done in Ref. [16]. The results are shown in Fig. 2, with circles for  $I_{\text{in}} = 1$  and crosses for  $I_{\text{in}} = -1$ , respectively. It is seen that the maps are quite complicated, and most of the points of the two maps coincide and entangle with each other. A distinguishable difference between the two maps is only seen for  $x_{\text{max}}(n)$  around -0.5, which consists of only a small fraction of the maxima. Extracting the message, although not totally impossible, can be done only for a small portion of the message bits.

#### B. The intruder system

Based on our assumption, the intruder knows that the chaotic system of the transmitter is Chua's circuit and that the message is encoded by the modulation of  $G_0$  and  $G_1$ , but does not know the value of any of the parameters  $C_1, C_2, R, L, B_p, G_0, G_1, r$ . Based on the available information, the intruder constructs a receiver system using the idea of parameter adaptive control as follows:

$$C_1 \frac{dy_1}{dt} = \frac{1}{R}(y_2 - y_1) - h(x_1), \quad (16)$$

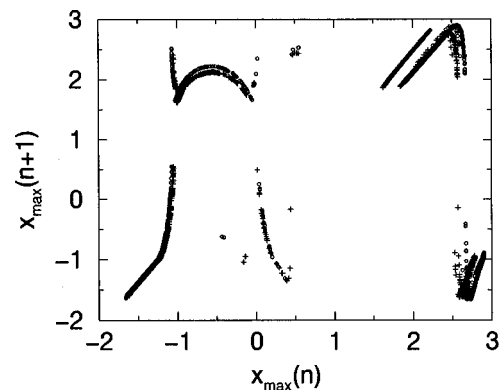


FIG. 2. Return maps of the consecutive maxima of the transmitted signal  $x_1$ . Most of the points for the attractor in (a) (circles) and those for the attractor in (b) (crosses) do not have distinguishable separation.

$$C_2 \frac{dy_2}{dt} = \frac{1}{R}(y_1 - y_2) + y_3, \quad (17)$$

$$L \frac{dy_3}{dt} = -y_2, \quad (18)$$

$$\frac{dQ_0}{dt} = \frac{1}{2}x_1[y_1 - x_1][1 - \text{sgn}(|x_1| - B_p)], \quad (19)$$

$$\frac{dQ_1}{dt} = \frac{1}{2}x_1[y_1 - x_1][1 + \text{sgn}(|x_1| - B_p)], \quad (20)$$

where  $Q_0$  and  $Q_1$  are controllable parameters of  $h(x_1)$ , which is now

$$h(x_1) = Q_1 x_1 + \frac{1}{2}(Q_0 - Q_1)[|x_1 + B_p| - |x_1 - B_p|]. \quad (21)$$

Equations (19) and (20) mean that  $Q_0$  is modified when  $|x_1| \leq B_p$  and  $Q_1$  is modified when  $|x_1| > B_p$ .

The intruder system of Eqs. (16)–(21) will synchronize with the transmitter Eqs. (12)–(15) if the parameters  $C_1, C_2, R, L, B_p$  are identical for the two systems. To prove it, let us examine the dynamics of the difference  $e_i = y_i - x_i$  ( $i = 1, 2, 3$ ),  $e_4 = Q_0 - G_0$ , and  $e_5 = Q_1 - G_1$  given by

$$C_1 \frac{de_1}{dt} = \frac{1}{R}(e_2 - e_1) - \frac{1}{2}x_1 e_4 [1 - \text{sgn}(|x_1| - B_p)] - \frac{1}{2}x_1 e_5 [1 + \text{sgn}(|x_1| - B_p)], \quad (22)$$

$$C_2 \frac{de_2}{dt} = \frac{1}{R}(e_1 - e_2) + e_3, \quad (23)$$

$$L \frac{de_3}{dt} = -e_2, \quad (24)$$

$$\frac{de_4}{dt} = \frac{1}{2}x_1 e_1 [1 - \text{sgn}(|x_1| - B_p)], \quad (25)$$

$$\frac{de_5}{dt} = \frac{1}{2}x_1 e_1 [1 + \text{sgn}(|x_1| - B_p)]. \quad (26)$$

The global Lyapunov function

$$E = C_1 e_1^2 + C_2 e_2^2 + L e_3^2 + e_4^2 + e_5^2, \quad (27)$$

$$\frac{dE}{dt} = -\frac{2}{R}(e_1^2 + e_2^2) \leq 0, \quad (28)$$

suggests that the state and parameters of the intruder system will converge to those of the transmitter. Figure 3 illustrates the synchronization process of the system with the attractor in Fig. 1(a) ( $I_{in} = 1$ ). The synchronization error decreases exponentially with time, with fluctuations only within small time scales, and we can expect that the synchronization is robust to parameter mismatches. Note that the stable values  $Q_0 = -0.753$  and  $Q_1 = -0.396$  are just the values of  $G_0$  and  $G_1$  in the transmitter.

When the information stream enters the transmitter, lasting a time interval  $T$  for each bit, the transmitted signal is a sequence switching between the two chaotic attractors in Fig.

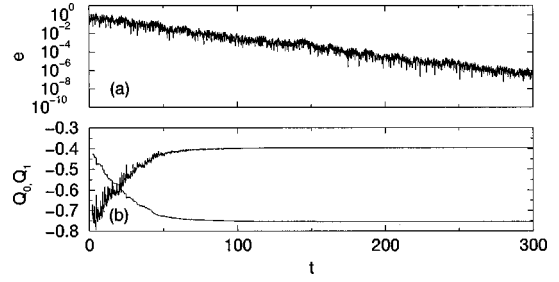


FIG. 3. Synchronization process of the intruder to the attractor in Fig. 2(a). (a) Synchronization error  $\sqrt{\sum e_i^2}$ . (b) Convergence of parameters  $Q_0$  and  $Q_1$  to those of the transmitter  $G_0 = -0.753$  and  $G_1 = -0.396$ , respectively. In all the figures in this paper, the unit of time is ms.

1. We take  $T = 4.65$  ms as in Ref. [8]. With the transmitted signal  $s = x_1$  [Fig. 4(a)] carrying a random message stream, the intruder observed the change of parameter  $Q_0$  and  $Q_1$  in Figs. 4(b) and 4(c), respectively. Switching between the two chaotic attractors results in only small fluctuations of  $Q_0$ , but sudden jumps of  $Q_1$ , because  $|x_1| > B_p$  most of the time so that  $Q_1$  is modified more frequently. After a transient of about 50 ms,  $Q_1$  comes to oscillate slightly about  $-0.395$  for bit +1 and  $-0.385$  for bit -1. A comparison between the evolution of  $Q_1$  and the parameter modulation in the transmitter shows clearly that the message can be decoded correctly except for a few bits during the synchronization transient. An interesting thing is that, since  $T$  is much smaller than the relaxation time of synchronization (about 50 ms, see Fig. 3), the intruder operates in a regime of synchronization transient after the message stream switches from one value to the other. As a result, the oscillation amplitude of  $Q_1$  (about  $10 \mu\text{s}$ ) is larger than the parameter modulation  $1/r = 6 \mu\text{s}$  in the transmitter, which can be an advantage for message decoding.

So far, we use the exact values of the transmitter parameters  $C_1, C_2, R, L, B_p$  in the intruder system to demonstrate that it is able to follow the parameter modulation in the trans-

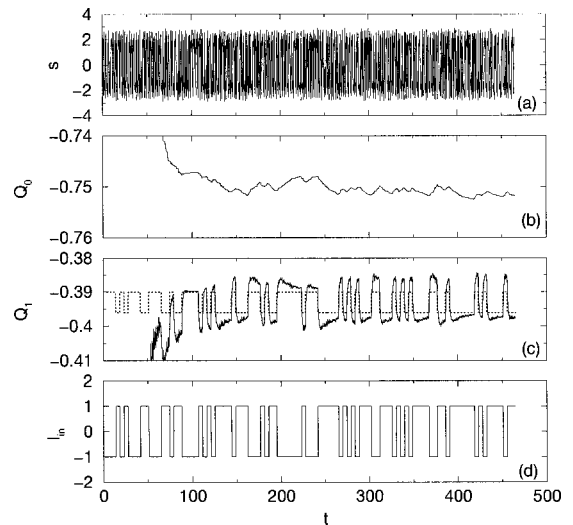


FIG. 4. The process of following the parameter modulation in the transmitter. (a) The transmitted signal  $s = x_1$ . (b) Change of  $Q_0$ . (c) Change of  $Q_1$ . (d) The input message. The dotted line shows the parameter modulation in the transmitter.

TABLE II. Parameters of the transmitter and the intruder and their relative differences.

Transmitter	$C_1$	$C_2$	$R$	$L$	$B_p$
Intruder	$0.817C_1$	$1.163C_2$	$1.072R$	$0.897L$	$0.849B_p$
Differences (%)	-18.3	16.3	7.2	-10.3	15.1

mitter by adaptive control. By assumption, the intruder does not have access to these values. However, it is possible to locate an approximate region in the parameter space using some characteristic quantities for system identification based on the knowledge of the chaotic system and at the same time to monitor the synchronization error during the scanning of the parameter space. In the following simulation, we suppose that the intruder is able to locate a region within 20% deviation from the precise values of the parameters. We choose five random values in  $[-0.2, 0.2]$  as the relative difference of the above parameters between the intruder and the transmitter, as displayed in Table II as an example. For the same information stream in Fig. 4(d), the evolution of  $Q_1$  is now presented in Fig. 5(a), which is shifted to oscillate quite noisily around  $-0.36$  due to the parameter mismatches. After smoothing the fluctuations with a moving average filter with a length of 4 ms, the oscillation of  $Q_1$  reveals most of the parameter modulation in the transmitter correctly, as seen in Fig. 5(b). We use a simple threshold testing to recover the message, as shown in Fig. 5(c) with a threshold  $Q_{th} = -0.365$ . A comparison between the recovered and the original messages has clearly shown that the security is compromised. The results are also robust to external noise, as seen in Fig. 6 for the same parameters as in Fig. 5, but with  $x_1$  containing noise between  $[-0.2, 0.2]$ .

#### IV. EXAMPLE OF THE LORENZ SYSTEM

In the above example, we are able to write down the parameter adaptive control rules based on an analytical treat-

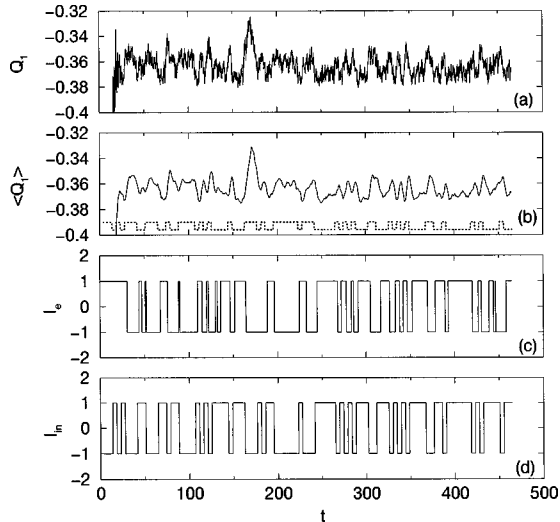


FIG. 5. An example of decoding process. (a) Evolution of  $Q_1$ . It oscillates noisily due to quite large parameter mismatches. (b) Smoothed  $Q_1$  by moving average with length of 4 ms. The dotted line shows the parameter modulation in the transmitter. (c) The decoded message by threshold testing with a threshold  $Q_{th} = -0.365$ . (d) The original message stream.

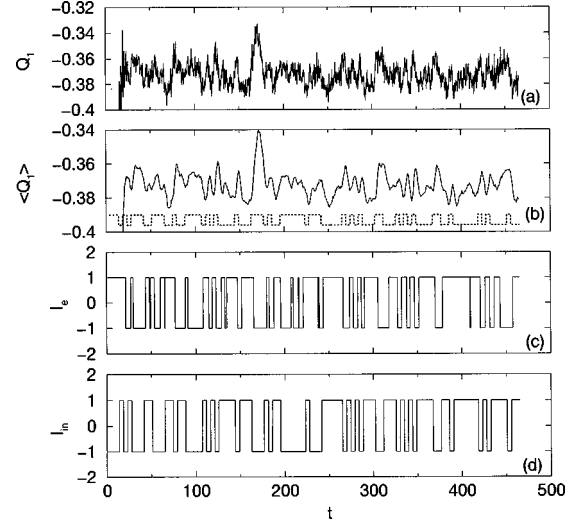


FIG. 6. Analogous to Fig. 5, but with  $s$  containing noise between  $[-0.2, 0.2]$ .

ment by a global Lyapunov function. In the following example, we revisit the communication system in Refs. [3,16] to illustrate the idea of designing an intruder system using the above local Lyapunov function method, although it has been shown that the message can be extracted using some suitable return maps [16].

The communication system using the Lorenz system is

$$\frac{dx_1}{d\tau} = \sigma(x_2 - x_1), \quad (29)$$

$$\frac{dx_2}{d\tau} = rx_1 - x_2 - x_1x_3, \quad (30)$$

$$\frac{dx_3}{d\tau} = x_1x_2 - bx_3, \quad (31)$$

where  $\sigma = 16.0$ ,  $r = 45.6$ , and  $b$  is modulated between  $b = 4.0$  and  $b = 4.4$ .  $s = x_1$  is the transmitted signal.

We can design an attacking system with parameter adaptive control for parameter  $b$  based on the following system coupled by feedback [17]:

$$\frac{dy_1}{d\tau} = \sigma(y_2 - y_1) + \epsilon(s - y_1), \quad (32)$$

$$\frac{dy_2}{d\tau} = ry_1 - y_2 - y_1y_3, \quad (33)$$

$$\frac{dy_3}{d\tau} = y_1y_2 - by_3, \quad (34)$$

which will be synchronized with the system  $x$  for large enough coupling strength  $\epsilon$ . The synchronization is also quite robust to parameter mismatches for large  $\epsilon$ . Since by assumption we do not know the parameter values in the transmitter, we use  $(\sigma, r, b) = (10, 28, 8/3)$  from a chaotic region in experiment or simulation. With  $\epsilon = 40$ , for example, the two systems are synchronized. Now let systems  $x$  and  $y$  have the same  $\sigma$  and  $r$ , but perturb the parameter  $b$  in the system  $y$  around  $b = 8/3$ , e.g.,  $b_y = b(1 + \Delta)$ . We calculate  $U(\Delta)$  as a function of  $\Delta$  by trying the following three sim-

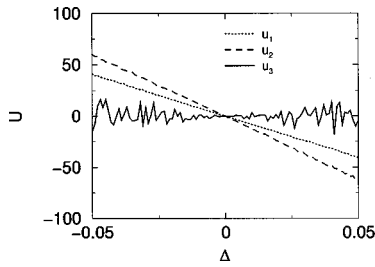


FIG. 7.  $U$  as a function of the relative deviation of the parameter  $b$  for different choice of function  $u$ . The smooth functions can be used to design the parameter adaptive control loop.

plest functions  $u_1=(s-y_1)y_1$ ,  $u_2=(s-y_1)y_2$ , and  $u_3=(s-y_1)y_3$ . The results of  $U$  are shown in Fig. 7. It is seen that  $U$  is a smooth function for  $\Delta$  close to zero for the functions  $u_1$  and  $u_2$ , but not for  $u_3$ . And it is obvious that  $U$  is also a smooth function for any linear combination of  $u_1$  and  $u_2$ . Some other choices of  $u$  are possible, for example  $u=(s-y_1)y_1y_3$ . Now we can introduce the additional parameter control loop

$$\frac{db}{d\tau} = \alpha u, \tag{35}$$

where  $u$  can be  $u_1$ ,  $u_2$ , any of their linear combinations, or many other possible choices. The sign of  $\alpha$  is determined by the sign of  $dU/d\Delta$ . Simulations have demonstrated that such designed control rules maintain the synchronization for  $\alpha$  small enough.  $\alpha$  is allowed to be larger for larger  $\epsilon$ . The control is still stable if the system parameters are shifted from  $(\sigma, r, b) = (10, 28, 8/3)$  to those of the transmitter, and the initial values of the parameter  $b$  in the system  $y$  do not need to be close to that of the system  $x$ . Since our purpose is to illustrate the designing idea, we do not go into great detail on the synchronization behavior in the parameter space  $(\epsilon, \alpha)$ . The fact is, in a large region of this parameter space, the synchronization is very robust to mismatches of the rest parameters  $\sigma$  and  $r$ . An example of the synchronization without and with the additional parameter control loop is shown in Fig. 8 for  $u = u_1 = (s - y_1)y_1$ ,  $(\sigma, r, b) = (16.0, 45.6, 4.0)$ , and  $(\epsilon, \alpha) = (40, 0.1)$ . The synchronization is a little slower when parameter adaptive control is introduced, and it is robust to parameter mismatches because the synchronization error decreases exponentially, with fluctuations only within very small time scales. The parameter  $b$  comes very close to

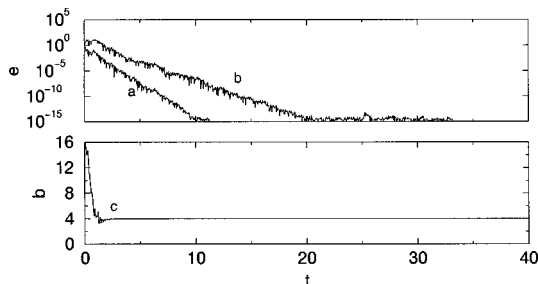


FIG. 8. Synchronization and parameter recovery with the additional parameter adaptive control loop of Eq. (35). (a) and (b) are synchronization errors without and with this control loop, respectively. (c) is the convergence of the parameter  $b$ .

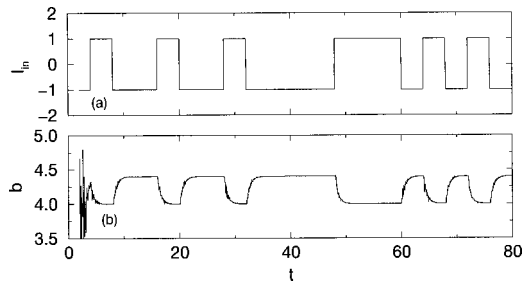


FIG. 9. The process of following the parameter modulation in the case in which the rest of the parameters  $\sigma$  and  $r$  are identical.

$b = 4.0$  in the transmitter from a large initial value  $b = 17.0$  within only a few ms (in the new time scale below).

Now let us use the system to attack the secure communication. In order that the time scale agrees with the system in [3,16], we introduce a new time scale  $t = \tau/K$ , where  $K = 2505$  [16] is a scale factor. In the transmitter, the bit duration is 4 ms. As in the above section, we first demonstrate the parameter recovery for identical parameters  $\sigma$  and  $r$  in the transmitter and intruder systems, as seen in Fig. 9. Then, we examine the robustness to the parameter mismatches of  $\sigma$  and  $r$ . The message can be recovered quite reliably if  $(\sigma_y, r_y)$  in the intruder system is within a relatively close neighborhood of the transmitter, say within a 20% deviation. Message decoding is generally extremely robust for  $\sigma_y < \sigma$  and  $r_y > r$ . An example for  $\sigma_y = 0.37\sigma$  and  $r_y = 1.72r$  is shown in Fig. 10. It is seen that  $b$  has been made to oscillate around  $b = 2.2$  rather than  $b = 4.2$  in the transmitter due to very large parameter mismatches, however the message is recovered correctly with a moving average filter with a length of 2 ms and a simple threshold test.

In the following, we go further to design adaptive control loops for all three parameters  $(\sigma, r, b)$  in the Lorenz system. We find that  $U$  changes smoothly when the parameters in the originally synchronized systems change slightly if we choose the following three functions:

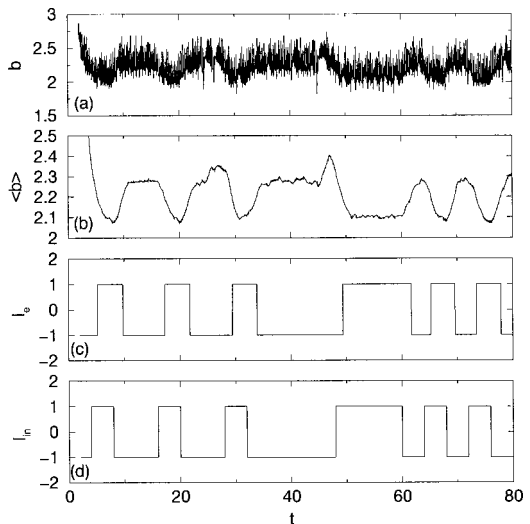


FIG. 10. Illustration of message decoding when the parameters  $\sigma$  and  $r$  have large mismatches between the systems. (a) Evolution of  $b$ . (b) Smoothed  $b$  by moving average filter with length of 2 ms. (c) The decoded message by threshold testing with a threshold  $b_{th} = 2.2$ . (d) The original message stream.

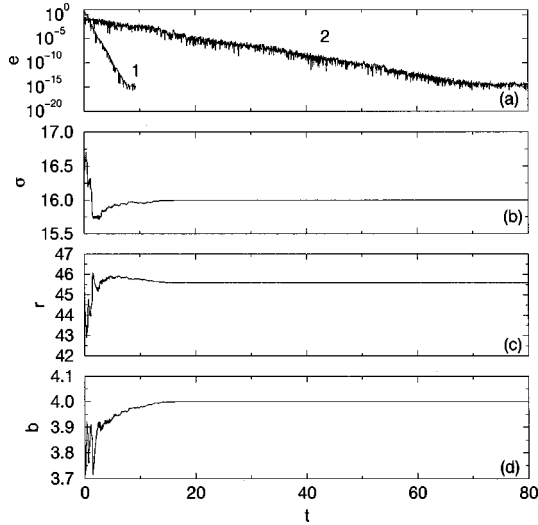


FIG. 11. Synchronization and parameter recovery with the additional parameter adaptive control loops for all three parameters. (a) Plots 1 and 2 are synchronization errors without and with these control loops, respectively. (b), (c), and (d) are the evolutions of the parameters  $\sigma$ ,  $r$ , and  $b$ , respectively.

$$\begin{aligned} u_1 &= (s - y_1)y_1y_3, & u_2 &= (s - y_1)y_2, \\ u_3 &= (s - y_1)(y_1 + y_2). \end{aligned} \quad (36)$$

The control surface is obtained by perturbing the parameters in the driven system within a 2% vicinity of  $(\sigma, r, b) = (10, 28, 8/3)$ .  $U_i$  ( $i=1,2,3$ ) is the time average of  $u_i$  in a period of 0.1 sec in the time scale  $t$ . After evaluating the matrix  $M^{-1}$ , we obtained the following attacking system with parameter adaptive control loops:

$$\frac{dy_1}{d\tau} = \sigma(y_2 - y_1) + \epsilon(s - y_1), \quad (37)$$

$$\frac{dy_2}{d\tau} = ry_1 - y_2 - y_1y_3, \quad (38)$$

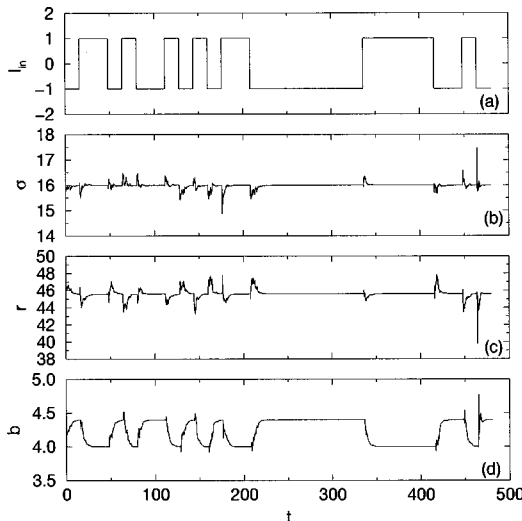


FIG. 12. The process of following the parameter modulation in the transmitter. (a) is an input message stream. (b), (c), and (d) are the evolutions of the parameters  $\sigma$ ,  $r$ , and  $b$ , respectively. The bit duration is 16 ms.

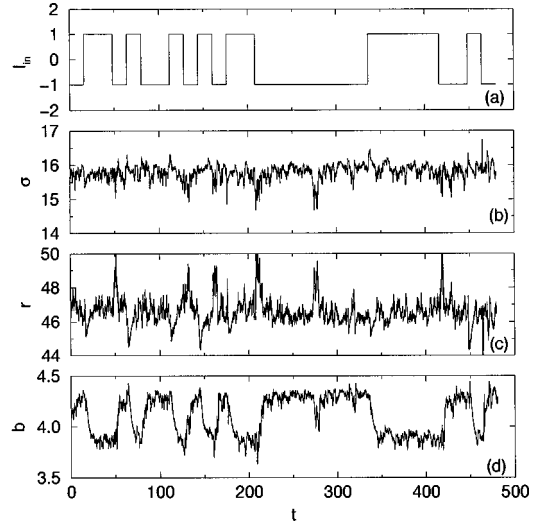


FIG. 13. Robustness of the message decoding in the presence of channel noise within  $[-1, 1]$ .

$$\frac{dy_3}{d\tau} = y_1y_2 - by_3, \quad (39)$$

$$\frac{d\sigma}{d\tau} = \alpha(-0.293u_1 - 18.5u_2 + 15.7u_3), \quad (40)$$

$$\frac{dr}{d\tau} = \alpha(1.18u_1 + 95.4u_2 - 75.4u_3), \quad (41)$$

$$\frac{db}{d\tau} = \alpha(-0.123u_1 - 10.2u_2 + 8.10u_3). \quad (42)$$

This system is synchronized for small enough  $\alpha$  even if the parameters have been shifted to those in the transmitter, i.e.,  $(\sigma, r, b) = (16.0, 45.6, 4.0)$ . An example of the synchronization and parameter recovery process is shown in Fig. 11 for  $(\epsilon, \alpha) = (100, 0.2)$ . The convergence rate with the additional parameter adaptive control is slower than that without these control loops. Now if the bit duration in the transmitter is longer than the synchronization transient, the attacking system can follow the parameter modulation in the transmitter and thus decode the message. An example is shown in Fig. 12, where the bit duration is 16 ms. While the parameter  $b$  clearly follows the modulation, the other two parameters also reflect the switch of the message from one value to the other. It is seen again in Fig. 11 that the synchronization error decreases exponentially, with fluctuations only within very small time scale, so that the message decoding is also very robust to channel noise. Figure 13 shows the recovered parameters when the transmitted signal  $s = x_1$  contains an additive noise in  $[-1, 1]$ .

## V. DISCUSSION

Based on the assumption that the chaotic system structure is in the public domain and the system parameters are kept in secret as the encryption key in a secure communication system encoding a digital message by parameter modulation, we have shown how an intruder might decode the message using an appropriate attacking system with parameter adaptive control by the transmitted signal, even though the intruder does not have access to the exact parameter values in the

transmitter. A requirement for the success of this attacking approach is that the intruder can design a synchronizing parameter adaptive control system which is quite robust to mismatches between the parameters of the two systems, so that the message can be recovered without resorting to the exact parameters in the transmitter, but within some neighborhood. Based on the knowledge of the system, it is practically possible for the intruder to get into such a neighborhood using some system identification methods.

For some systems, such a robust synchronizing intruder system with parameter adaptive control can be constructed based on an analysis of a global Lyapunov function. Generally, such an analytical treatment is impossible, and we proposed a quite general and practical local Lyapunov function method to design parameter adaptive control rules based on a system which has been synchronized by the transmitted signal. Such a synchronizing system is often obtainable using many possible approaches for constructing synchronization chaotic systems, such as Pecora-Carroll decomposition, active-passive decomposition, or feedback control. In many systems, the synchronization is robust to parameter mismatches if the coupling is not close to the synchronization threshold. The parameter control rules are designed by seeking appropriate functions of the synchronization error whose time average changes smoothly when the parameters in the originally synchronized systems deviate slightly from each other. The smooth control surface is obtained in simulation or experiment by perturbing the parameters that will be involved in adaptive control. Although this scheme is quite general, in practice finding a set of appropriate functions may not be a trivial task when many parameters are involved in modulation. In some cases, the originally synchronized system may be very sensitive to parameter mismatches due to unstable invariant sets embedded in the synchronization manifold [19,20], and the proposed local Lyapunov function

may not be successful in designing additional parameter adaptive control loops for such systems. However, such systems will not be used in the communication systems because the authorized receiver cannot decode the message correctly in practical environment with unavoidable perturbations.

Employing some system identification methods, the intruder may be able to identify a region near the transmitter parameters in the parameter space in order to design a stable intruder system. Furthermore, the intruder may be able to get close enough to the transmitter parameters by monitoring the synchronization error while scanning the parameter space, so that the message can be decoded with a very low rate of errors. During the decoding process, the intruder can improve the decoding by comparison of the results using different parameters in the identified region. In some systems, it is also possible to design adaptive control loops for all the system parameters, so that the message can be decoded even more reliably.

Based on our investigation, we would like to point out an interesting paradox between robustness and security in chaotic communications. Since, in practice, parameter mismatches and external noise are unavoidable, we would require the synchronization systems to be robust to these perturbations, so that high-quality synchronization can be established between the transmitter and the authorized receiver to recover the message correctly in practical implementation. On the other hand, this robustness may be employed by a third party to compromise the security. How to improve the security while maintaining the robustness is an interesting and meaningful research topic for chaotic communications.

#### ACKNOWLEDGMENTS

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