# 國立交通大學

# 機械工程學系 碩士論文

新 Duffing-Van der Pol 系統的渾沌現象,實用渾沌廣義同步,辛同步,應用 GYC 部分區域穩定理論的渾沌同步及控制

Chaos, Pragmatical Chaotic Generalized
Synchronization, Symplectic Synchronization, and
Chaos Synchronization and Control by GYC Partial
Region Stability Theory of a New Duffing-Van der Pol
System

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中華民國九十七年六月

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新 Duffing-Van der Pol 系統的渾沌現象,實用渾沌廣義同步,辛同步, 應用 GYC 部分區域穩定理論的渾沌同步及控制

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#### 摘要

本篇論文以相圖、龐卡萊映射圖、Lyapunov 指數、分歧圖以及參數圖等數值方法研究新 Duffing-Van der Pol 系統的渾沌現象。對此系統研究應用實用漸進穩定性理論和適應性控制法則來達成實用正反投影渾沌廣義同步;應用新動態表面控制和實用漸進穩定理論來獲得不同階系統之實用渾沌交織同步。更進一步使用 GYC 部分區域穩定理論來研究系統的廣義渾沌同步、渾沌控制和渾沌反控制。此外,將探討新 Duffing-Van der Pol 系統以 Legendre function 為參數的渾沌行為及同步。在以上研究中,皆可由相圖和時間歷程圖得到驗證。

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Chaos, Pragmatical Chaotic Generalized Synchronization, Symplectic

Synchronization, and Chaos Synchronization and Control by GYC Partial

Region Stability Theory of a New Duffing-Van der Pol System

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**Abstract** 

In this thesis, the chaotic behavior in new Duffing-van der Pol system is studied

by phase portraits, time history, Poincaré maps, Lyapunov exponent, bifurcation

diagrams, and parametric diagram. A new kind of chaotic generalized synchronization

system, pragmatical hybrid projective chaotic generalized synchronization

(PHPCGS), is obtained by pragmatical asymptotical stability theorem and adaptive

control law. Second new type for chaotic synchronization, pragmatical chaotic

symplectic synchronization (PCSS), is obtained by new dynamic surface control and

pragmatical asymptotical stability theorem. A new method, using GYC partial region

stability theory, is studied for chaos synchronization, chaos control, and chaos

anti-control. Moreover, the new Duffing-van der Pol system with Legendre function

parameters is studied for chaos and synchronization. Numerical analyses, such as

phase portraits and time histories can be provided to verify the effectiveness in all

above studies.

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### Chapter 1

#### Introduction

In the phenomenon of chaos, the chaotic state is very sensitive to its initial condition and chaos causes often irregular behavior in practical systems. Slight errors occurring in initial states of two identical oscillators will lead to completely different trajectories after enough time. Research efforts have studied control and synchronization of chaos in many chaotic systems [1-4].

Many methods have been applied theoretically and experimentally to synchronize chaotic system [5-7]. Chaos synchronization has been widely investigated in a variety of fields such as secure communication [8], biological science [9], chemical reaction [10], physical science [11], and many other fields. So far, there exist many types of synchronization such as complete synchronization [4,12,13], phase synchronization [14,15], lag synchronization [16,17], projective synchronization [18-22], and generalized synchronization [23], etc. Complete synchronizations and antisynchronizations are the special cases of generalized projective synchronization where the scaling factor  $\alpha = 1$  and  $\alpha = -1$ , respectively.

Many approaches for the synchronization of chaotic systems are based on the exact knowledge of the system structure and parameters. But in practice, almost all parameters of the system are uncertain. In current study, the traditional Lyapunov stability theorem and Babalat lemma are used to prove that the error vector approaches zero in adaptive synchronization [24-28]. But the question of that why the estimated parameters also approach uncertain parameters remains unanswered. By the pragmatical asymptotical stability theorem, the question can be answered strictly [29-31].

In symplectic synchronization, there exists a functional relationship between

"master" system and "slave" system. The final state y of "slave" system not only depends upon the state x of "master" system but also depends upon itself. In other words, the "slave" system does not completely obey the "master" system but plays a role to determine the final state y of "slave" system [32]. The generalized synchronization is a special case of the symplectic synchronization. Besides, the state variables of another different order system as a constituent of functional relation between "master" and "slave" are used.

Chaos control has attracted a great deal of attention from various fields. There are many control techniques for chaos control, such as active control [33], observer-based control [34], feedback and non-feedback control [35-38], inverse optimal control [39], adaptive control [40,41], etc. A new chaos synchronization and chaos control strategy by GYC partial region stability theory are proposed [42,43]. By using the GYC partial region stability theory, the controllers are of lower degree than that of controllers by using traditional Lyapunov asymptotical stability theorem [21-28]. The simple linear homogeneous Lyapunov function of error states makes the controllers introducing less simulation error.

This thesis is organized as follows. Chapter 2 gives the dynamic equation of new Duffing-Van der Pol system. The chaotic behaviors are studied. In Chapter 3 and Chapter 4, numerical simulations of generalized and symplectic synchronization scheme based on the pragmatical asymptotical stability theorem with adaptive control and new dynamic surface control are presented. In Chapter 5, numerical simulations of chaos synchronization scheme based on GYC partial region stability theory are presented. In Chapter 6, numerical simulations of chaos control and anti-control scheme based on GYC partial region stability theory are presented. In Chapter 7, numerical simulations of chaos synchronization scheme using chaotic system with Legendre function parameters are presented. In Chapter 8, conclusions are drawn.

# Chapter 2

### Chaos of a New Duffing-Van der Pol System

In this Chapter, the chaotic behaviors of a new Duffing-Van der Pol system is studied numerically by phase portraits, time histories, Poincaré maps, Lyapunov exponents, bifurcation diagrams, and parametric diagram.

### 2.1 Description of New Duffing-Van der Pol System

Duffing equation and Van der Pol equation are two typical nonlinear nonautonomous systems:

$$\begin{cases} \frac{dx_1}{dt} = x_2 \\ \frac{dx_2}{dt} = -x_1 - x_1^3 - ax_2 + d\sin\omega t \end{cases}$$
 (2.1)

$$\begin{cases} \frac{dx_{1}}{dt} = x_{2} \\ \frac{dx_{2}}{dt} = -x_{1} - x_{1}^{3} - ax_{2} + d \sin \omega t \\ \frac{dx_{3}}{dt} = x_{4} \\ \frac{dx_{4}}{dt} = -bx_{3} + c(1 - x_{3}^{2})x_{4} + f \sin \omega t \end{cases}$$
(2.1)

A new autonomous Duffing-Van der Pol system is proposed by coupling of these two typical nonautonomous systems. Exchanging  $sin\omega t$  in Eq. (2.1) with  $x_3$  and  $\sin \omega t$  in Eq. (2.2) with  $x_1$ , we obtain the new autonomous Duffing-van der Pol system:

$$\begin{cases} \frac{dx_1}{dt} = x_2 \\ \frac{dx_2}{dt} = -x_1 - x_1^3 - ax_2 + dx_3 \\ \frac{dx_3}{dt} = x_4 \\ \frac{dx_4}{dt} = -bx_3 + c\left(1 - x_3^2\right)x_4 + fx_1 \end{cases}$$
(2.3)

where a, b, c, d, f are parameters. Computational analysis of Eq. (2.3) is studied as

follows.

# 2.2 Computational Analysis of a New Duffing-Van der Pol System

For numerical analysis of computation, this system exhibits chaos when the parameters of system are a=0.01, b=1, c=5, d=0.67, f=0.05 and the initial states of system are  $x_1(0)=2, x_2(0)=2.4, x_3(0)=5, x_4(0)=6$ . The bifurcation diagram by changing damping parameter a is shown in Fig. 2.1. Its corresponding Lyapunov exponents are shown in Fig. 2.2. The phase portraits, time histories, and Poincaré maps of the systems are showed in Fig. 2.3~Fig. 2.5. When a=0.012, period 2 phenomena are shown in Fig. 2.3. When a=0.01 and a=0.0006, the chaotic behaviors are given in Fig. 2.4 and Fig. 2.5, respectively. In addition, the parametric diagram is obtained in Fig. 2.6.

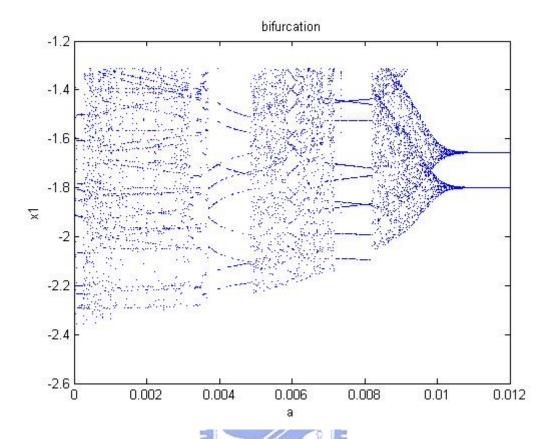


Fig. 2.1 The bifurcation diagram for new Duffing-Van der Pol system.

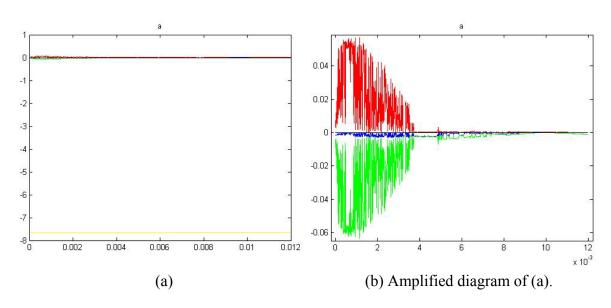


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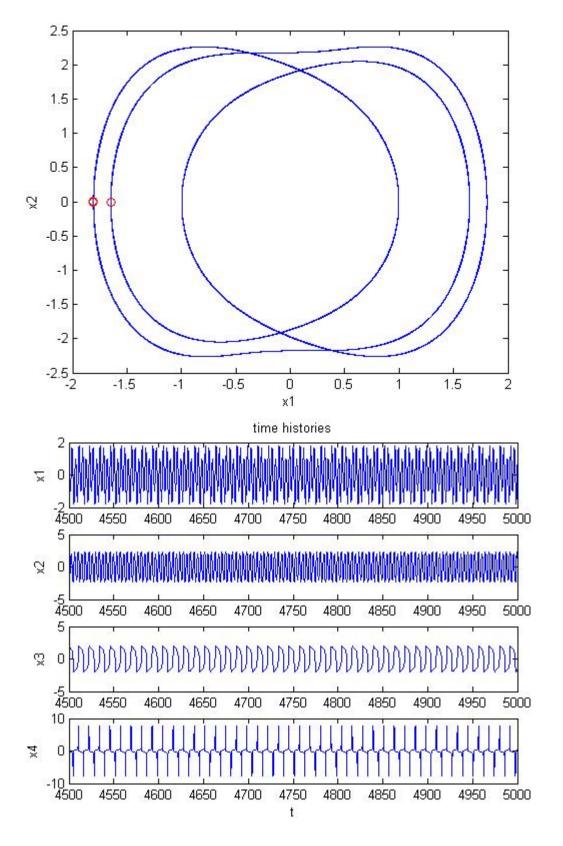


Fig. 2.3 Phase portrait, Poincaré maps, and time histories for new Duffing-Van der Pol system with a=0.012 (period 2).

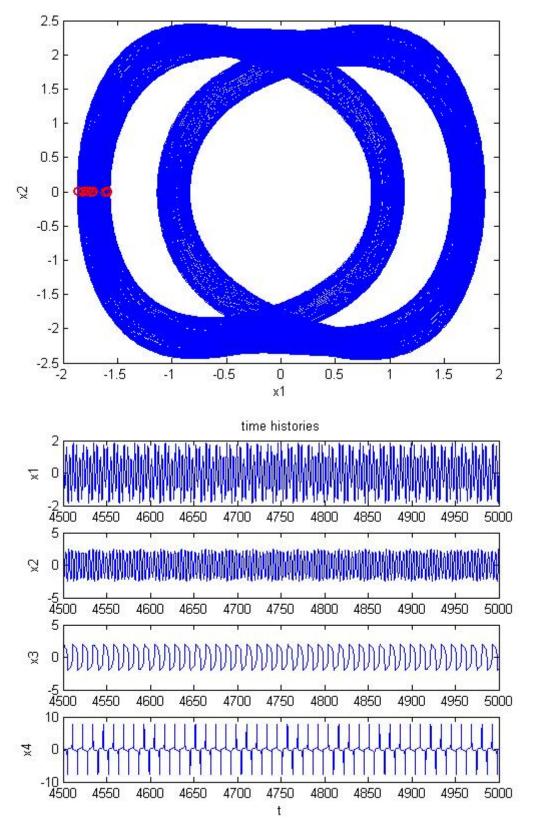


Fig. 2.4 Phase portrait, Poincaré maps, and time histories for new Duffing-Van der Pol system with a=0.01 (chaos).

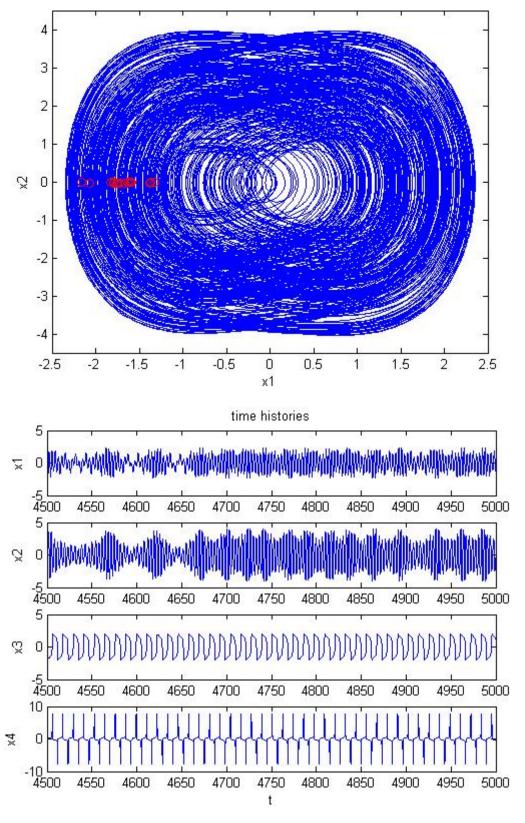


Fig. 2.5 Phase portrait, Poincaré maps, and time histories for new Duffing-Van der Pol system with a=0.0006 (chaos).

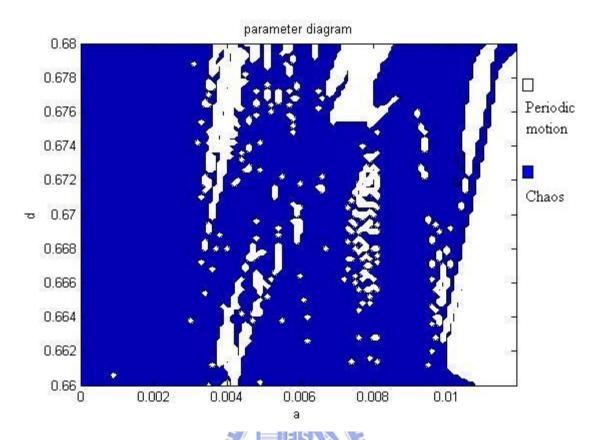


Fig. 2.6 The parametric diagram for new Duffing-Van der Pol system with b=1,

c=5, f=0.05

# **Chapter 3**

# Pragmatical Hybrid Projective Chaotic Generalized Synchronization of Chaotic System with Uncertain Parameters by Adaptive Control

A new kind of generalized synchronization, *pragmatical hybrid projective* chaotic generalized synchronization (PHPCGS) of two identical chaotic systems of which one has uncertain parameters, by pragmatical adaptive control, is achieved with the state vector of another hyperchaotic chaotic system as a constituent of the functional relation between master and slave. The PHPCGS is as follows:

$$y = G(x) = zH x \tag{3.1}$$

where  $x = (x_1, x_2, ..., x_n)^T$ ,  $y = (y_1, y_2, ..., y_n)^T \in \mathbb{R}^n$  are the n-dimensional state vectors of master system and slave system, respectively,  $H = diag[h_1, h_2, ..., h_n] \in \mathbb{R}^{(n \times n)}$  is a constant scaling diagonal matrix where  $h_i$  are called scaling factors, which may either positive or negative to form the hybrid projective synchronization.  $z = [z_1(t), z_2(t), ..., z_n(t)]$  is a given state vector with chaotic time function components. The existence of z causes so called chaotic synchronization. Based on the pragmatical asymptotical stability theorem, adaptive control law is used. Furthermore, the scheme can achieve not only projective synchronization, but also projective anti-synchronization. With all above consideration, PHPCGS is achieved. Numerical simulations are provided to verify the effectiveness of the proposed scheme.

# 3.1 The PHPCGS Scheme of Chaotic Systems by Adaptive Control

There are two identical nonlinear dynamical systems, and the master system

controls the slave system. The master system is given by

$$\dot{x} = Ax + f(x, B) \tag{3.2}$$

where  $x = [x_1, x_2, \dots x_n]^T \in \mathbb{R}^n$  denotes a state vector, A is an  $n \times n$  uncertain constant coefficient matrix, f is a nonlinear vector function, and B is a vector of uncertain constant coefficients in f. The slave system is given by

$$\dot{y} = \hat{A}y + f(y, \hat{B}) + u(t)$$
 (3.3)

where  $y = [y_1, y_2, \cdots y_n]^T \in \mathbb{R}^n$  denotes a state vector,  $\hat{A}$  is an  $n \times n$  estimated coefficient matrix,  $\hat{B}$  is a vector of estimated coefficients in f, and  $u(t) = [u_1(t), u_2(t), \cdots u_n(t)]^T \in \mathbb{R}^n$  is a control input vector. The chaotic system which affords chaotic z matrix, is called functional system because in Eq. (3.1), it is a constituent of function G. The functional system is given by

$$\dot{z} = Cz + g(z) \tag{3.4}$$

where  $z = [z_1, z_2, \dots z_n]^T \in \mathbb{R}^n$  denotes a state vector, C is an  $n \times n$  constant coefficient matrix, g is a nonlinear vector function. PHPCGS demands:

$$y = G(x, z) = zH x \tag{3.5}$$

Our goal is to accomplish Eq. (3.5) via controller u(t) and parameter update dynamics. Define the error vector e:

$$e = zH x - y \tag{3.6}$$

The synchronization is achieved when

$$\lim_{t \to \infty} e_i = 0 \qquad (i = 1, 2, ..., n)$$
(3.7)

By (3.2), (3.3), and (3.4), the error dynamics is

$$\dot{e} = zH\dot{x} + \dot{z}Hx - \dot{y}$$

$$= zHAx - \hat{A}y + zHf(x,B) - f(y,\hat{B}) + CzHx + g(z)Hx - u(t)$$
(3.8)

A Lyapunov function  $V(e, \tilde{A}_c, \tilde{B}_c)$  is chosen as a positive definite function of  $e, \ \tilde{A}_c, \ \tilde{B}_c$ :

$$V(e, \widetilde{A}_c, \widetilde{B}_c) = \frac{1}{2}e^T e + \frac{1}{2}\widetilde{A}_c^T \widetilde{A}_c + \frac{1}{2}\widetilde{B}_c^T \widetilde{B}_c$$
(3.9)

where  $\tilde{A} = A - \hat{A}$ ,  $\tilde{B} = B - \hat{B}$ ,  $\tilde{A}_c$  and  $\tilde{B}_c$  are two vectors whose elements are all the elements of matrix  $\tilde{A}$  and of matrix  $\tilde{B}$ , respectively. By properly choosing u(t),  $\dot{\tilde{A}}_c$ , and  $\dot{\tilde{B}}_c$ , its time derivative  $\dot{V}$  along any solution of Eq. (3.8) and parameter update differential equations for  $\tilde{A}_c$  and  $\tilde{B}_c$  can be obtained as

$$\dot{V} = e^T C e \tag{3.10}$$

where C is a diagonal negative definite matrix.  $\dot{V}$  is a negative definite function of e but a negative semi-definite function of e,  $\tilde{A}_e$ ,  $\tilde{B}_e$ . In the current scheme of adaptive synchronization [24-28], the traditional Lyapunov stability theorem and Babalat lemma are used to prove that the error vector approaches zero, as time approaches infinity. But the question that why the estimated parameters also approach uncertain parameters remains unanswered. By the GYC pragmatical asymptotical stability theorem, the question can be answered strictly. The equilibrium point  $e = \tilde{A} = \tilde{B} = 0$  is pragmatically asymptotically stable. Under the assumption of equal probability, it is actually asymptotically stable, as shown in Appendix. Hence, PHPCGS can be achieved.

# 3.2 Description of Two New Chaotic 4D Systems, a New Duffing-Van der Pol System and a New Mathieu-Duffing System

This section introduces new Duffing-Van der Pol system and new Mathieu-Duffing system.

#### 3.2.1 New Duffing-Van der Pol system

The master new Duffing-Van der Pol system:

$$\begin{cases} \frac{dx_1}{dt} = x_2 \\ \frac{dx_2}{dt} = -x_1 - x_1^3 - ax_2 + dx_3 \\ \frac{dx_3}{dt} = x_4 \\ \frac{dx_4}{dt} = -bx_3 + c\left(1 - x_3^2\right)x_4 + fx_1 \end{cases}$$
(3.11)

The slave system is

$$\begin{cases} \frac{dy_1}{dt} = y_2 + u_1 \\ \frac{dy_2}{dt} = -y_1 - y_1^3 - \hat{a}y_2 + \hat{d}y_3 + u_2 \\ \frac{dy_3}{dt} = y_4 + u_3 \\ \frac{dy_4}{dt} = -\hat{b}y_3 + \hat{c}(1 - y_3^2)y_4 + \hat{f}y_1 + u_4 \end{cases}$$
where  $u_1, u_2, u_3, u_4$  are control inputs,  $a, b, c, d, f$  are uncertain parameters,

where  $u_1, u_2, u_3, u_4$  are control inputs, a, b, c, d, f are uncertain parameters,  $\hat{a}$ ,  $\hat{b}$ ,  $\hat{c}$ ,  $\hat{d}$  and  $\hat{f}$  are estimated parameters. The master system exhibits chaos when the parameters are a = 0.01, b = 1, c = 5, d = 0.67, f = 0.05 and the initial states are  $x_1(0) = 2, x_2(0) = 2.4, x_3(0) = 5, x_4(0) = 6$ .

#### 3.2.2 New Mathieu-Duffing system

The functional system is a new Mathieu-Duffing system. By coupling two noautonomous nonlinear Mathieu system and Duffing system, a new Mathieu-Duffing system is obtained:

$$\begin{cases}
\frac{dz_1}{dt} = z_2 \\
\frac{dz_2}{dt} = -(a_3 + b_3 z_3) z_1 - (a_3 + b_3 z_3) z_1^3 - c_3 z_2 + d_3 z_3 \\
\frac{dz_3}{dt} = z_4 \\
\frac{dz_4}{dt} = -z_3 - z_3^3 - f_3 z_4 + g_3 z_1
\end{cases}$$
(3.13)

This system exhibits chaos when the parameters of system are  $a_3=20.3$ ,

 $b_3$ =0.597,  $c_3$ =0.005,  $d_3$ =-24.44,  $f_3$ =0.002,  $g_3$ =14.63 and the initial states of system are  $z_1(0)$ =-2,  $z_2(0)$ =10,  $z_3(0)$ =-2,  $z_4(0)$ =10, its phase portraits and time histories as shown in Fig. 3.1.

#### 3.3 Numerical Results for PHPCGS of Two New

#### **Duffing-Van der Pol Systems**

Since the master system, slave system, and functional system are described by Eq. (3.11), Eq. (3.12), and Eq. (3.13), respectively, the error dynamics Eq. (3.8) becomes:

$$\begin{cases} \dot{e}_{1} = h_{1}(x_{2}z_{1} + x_{1}z_{2}) - y_{2} - u_{1} \\ \dot{e}_{2} = h_{2}\left\{z_{2}\left[-x_{1} - x_{1}^{3} - ax_{2} + dx_{3}\right] + x_{2}\left[-(a_{3} + b_{3}z_{3})(z_{1} + z_{1}^{3}) - c_{3}z_{2} + d_{3}z_{3}\right]\right\} \\ + y_{1} + y_{1}^{3} + \hat{a}y_{2} - \hat{d}y_{3} - u_{2} \\ \dot{e}_{3} = h_{3}(x_{4}z_{3} + x_{3}z_{4}) - y_{4} - u_{3} \\ \dot{e}_{4} = h_{4}\left\{z_{4}\left[-bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1}\right] + x_{4}\left[-z_{3} - z_{3}^{3} - f_{3}z_{4} + g_{3}z_{1}\right]\right\} \\ + \hat{b}y_{3} - \hat{c}(1 - y_{3}^{2})y_{4} - \hat{f}y_{1} - u_{4} \end{cases}$$

$$(3.14)$$

Choose a positive definite Lyapunov function for  $e_1$ ,  $e_2$ ,  $e_3$ ,  $e_4$ ,  $\tilde{a}$ ,  $\tilde{b}$ ,  $\tilde{c}$ ,  $\tilde{d}$ ,  $\tilde{f}$ :

$$V = \frac{1}{2} (e_1^2 + e_2^2 + e_3^2 + e_4^2 + \tilde{a}^2 + \tilde{b}^2 + \tilde{c}^2 + \tilde{d}^2 + \tilde{f}^2)$$
(3.15)

where  $\tilde{a}=a-\hat{a}$ ,  $\tilde{b}=b-\hat{b}$ ,  $\tilde{c}=c-\hat{c}$ ,  $\tilde{d}=d-\hat{d}$ , and  $\tilde{f}=f-\hat{f}$ . Controllers and parameter update dynamics are selected as:

$$\begin{cases} u_{1} = h_{1}(x_{2}z_{1} + x_{1}z_{2}) - y_{2} + e_{1} \\ u_{2} = h_{2}\left\{-z_{2}\left[x_{1} + x_{1}^{3} + \hat{a}x_{2} - \hat{d}x_{3}\right] - x_{2}\left[(a_{3} + b_{3}z_{3})(z_{1} + z_{1}^{3}) + c_{3}z_{2} - d_{3}z_{3}\right]\right\} \\ + y_{1} + y_{1}^{3} + \hat{a}y_{2} - \hat{d}y_{3} + e_{2} \\ u_{3} = h_{3}(x_{4}z_{3} + x_{3}z_{4}) - y_{4} + e_{3} \\ u_{4} = h_{4}\left\{z_{4}\left[-bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1}\right] + x_{4}\left[-z_{3} - z_{3}^{3} - f_{3}z_{4} + g_{3}z_{1}\right]\right\} \\ + \hat{b}y_{3} - \hat{c}(1 - y_{3}^{2})y_{4} - \hat{f}y_{1} + e_{4} \end{cases}$$

$$(3.16)$$

$$\begin{cases}
\dot{\hat{a}} = -e_2 h_2 z_2 x_2 \\
\dot{\hat{b}} = -e_4 h_4 z_4 x_3 \\
\dot{\hat{c}} = e_4 h_4 z_4 \left(1 - x_3^2\right) x_4
\end{cases}$$

$$\dot{\hat{d}} = e_2 h_2 z_2 x_3 \\
\dot{\hat{f}} = e_4 h_4 z_4 x_1$$
(3.17)

The time derivative of V is

$$\dot{V} = -e_1^2 - e_2^2 - e_3^2 - e_4^2 \qquad \le 0 \tag{3.18}$$

which is negative semi-definite function for  $e_1$ ,  $e_2$ ,  $e_3$ ,  $e_4$ ,  $\tilde{a}$ ,  $\tilde{b}$ ,  $\tilde{c}$ ,  $\tilde{d}$ ,  $\tilde{f}$ . The Lyapunov asymptotical stability theorem cannot be satisfied in this case. The common origin of error dynamics and parameter update dynamics cannot be concluded to be asymptotically stable. By pragmatical asymptotical theorem, D is a 9-manifold, n=9 and the number of error state variables p=4. When  $e_1=e_2=e_3=e_4=0$  and  $\tilde{a}$ ,  $\tilde{b}$ ,  $\tilde{c}$ ,  $\tilde{d}$ ,  $\tilde{f}$  take arbitrary values,  $\dot{V}=0$ , so X is a 5-manifold, m=n-p=9-4=5. m+1 < n are satisfied. By the pragmatical asymptotical stability theorem, the common origin of error dynamics (3.14) and parameter dynamics (3.17) is asymptotically stable. The equilibrium point  $e_1=e_2=e_3=e_4=\tilde{a}=\tilde{b}=\tilde{c}=\tilde{c}=\tilde{d}=\tilde{f}=0$  is pragmatically asymptotically stable. The PHPCGS is achieved under this scheme.

In this numerical simulation, we select the "unknown" parameter and initial states of the master system and of functional system the same as that in Section 3.2 to ensure the chaotic behavior. The initial states of slave system are  $y_1(0) = y_2(0) = 5$  and  $y_3(0) = y_4(0) = 10$ , the estimated parameters have initial conditions  $\hat{a}(0) = \hat{b}(0) = \hat{c}(0) = \hat{d}(0) = \hat{f}(0) = 0$  and the scaling matrix is  $H = diag(2, \frac{-1}{2}, 2, \frac{-1}{2})$ . The numerical results are shown in Fig. 3.2 and Fig. 3.3.

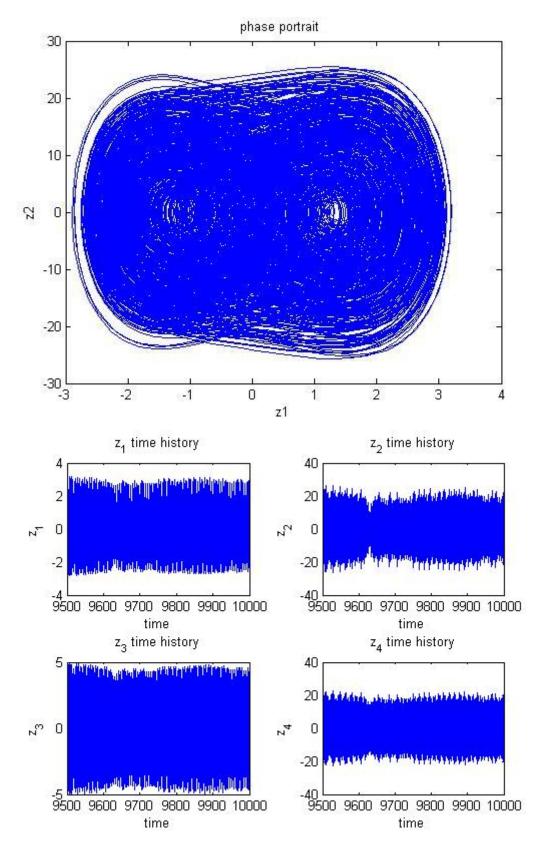


Fig. 3.1 Phase portrait and time histories of chaotic Mathieu-Duffing system.

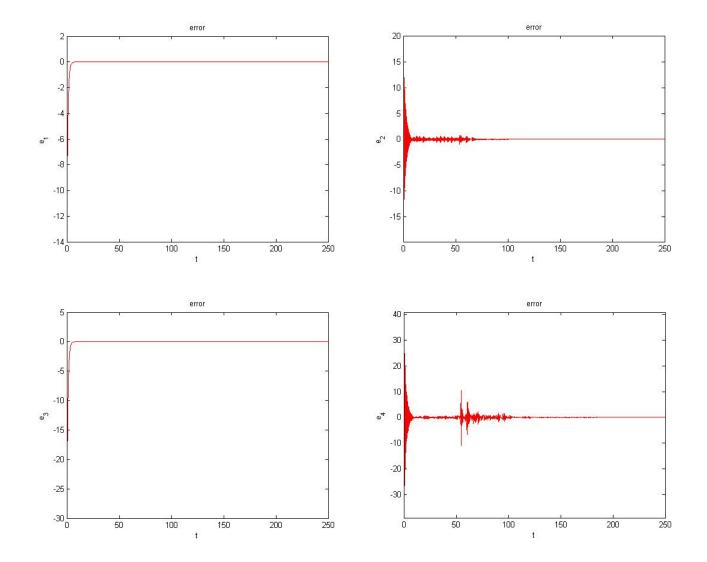


Fig. 3.2 Time histories of errors.

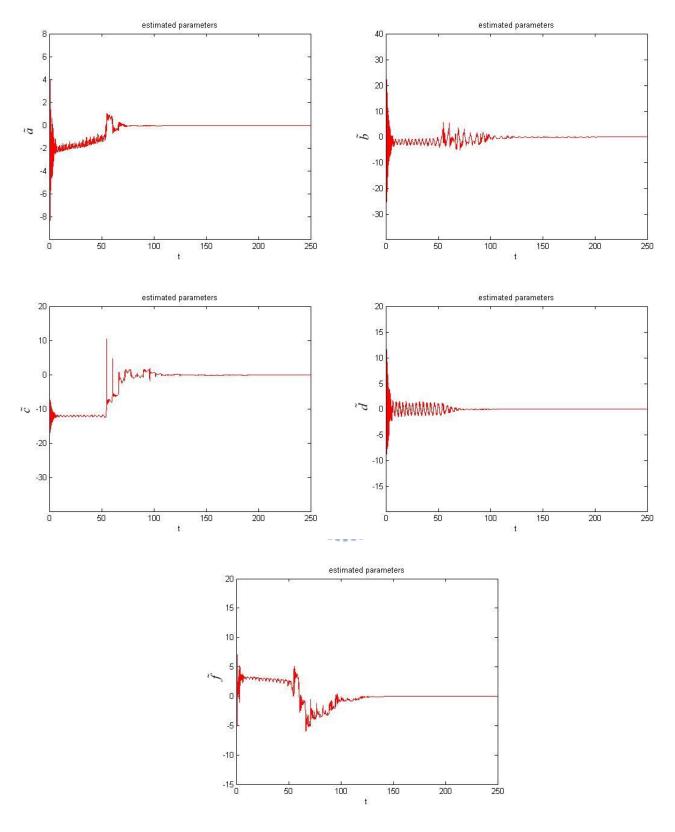


Fig. 3.3 Time histories of the differences of uncertain parameters and estimated parameters.

# Chapter 4

# Pragmatical Chaotic Symplectic Synchronization with Different Order System by New Dynamic Surface Control

A new type of chaotic synchronization, *pragmatical chaotic symplectic synchronization* (PCSS), is obtained with the state variables of another different order system as a constituent of the functional relation between "master" and "slave". The PCSS as follows:

$$y = H(x, y, t) + F(t) \tag{4.1}$$

where x and y are the "master" and the "slave" system, respectively. The final state y of "slave" system not only depends upon the state x of "master" system but also depends upon itself. In other words, the "slave" system does not completely obey the "master" system but plays a role to determine the final state y of "slave" system. This kind of synchronization called "symplectic synchronization"\*, and the "master" and "slave" system called partner A and partner B, respectively. When H=H(x,t) symplectic synchronization becomes traditional generalized synchronization. Therefore the latter is a special case of the former.

<sup>\*</sup> The term "symplectic" comes from the Greek for "intertwined". H. Weyl first introduced the term in 1939 in his book "The Classical Groups" (P. 165 in both the first edition, 1939, and second edition, 1946, Princeton University Press)

F(t) is a given chaotic vector of time from the states of another different order chaotic system. Based on the pragmatical asymptotical stability theorem, new dynamic surface control (NDSC) which makes the controllers more simple, and adaptive control, the synchronization is achieved. Numerical simulations are provided to verify the effectiveness of the proposed scheme.

#### 4.1 Pragmatical Chaotic Symplectic Synchronization Scheme

There are two identical nonlinear chaotic dynamical systems, and the "master" system controls the "slave" system. In symplectic synchronization, the "master" system is called partner A:

$$\dot{x} = Ax + f(x, B) \tag{4.2}$$

where  $x = [x_1, x_2, \dots x_n]^T \in \mathbb{R}^n$  denotes a state vector, A is an  $n \times n$  uncertain constant coefficient matrix, f is a nonlinear vector function, and B is a vector of uncertain constant coefficients in f. The "slave" system is called partner B:

$$\dot{y} = \hat{A}y + f(y, \hat{B}) \tag{4.3}$$

where  $y = [y_1, y_2, \dots y_n]^T \in \mathbb{R}^n$  denotes a state vector,  $\hat{A}$  is an  $n \times n$  estimated coefficient matrix,  $\hat{B}$  is a vector of estimated coefficients in f. With controllers, partner B becomes

$$\dot{y}_{u} = \hat{A}y + f(y, \hat{B}) + u(t)$$
 (4.4)

where  $u(t) = [u_1(t), u_2(t), \dots u_n(t)]^T \in \mathbb{R}^n$  is a control input vector. The chaotic system which affords chaotic F(t) vector, is called functional system. However, the PCSS also can be achieved even the order of functional system is different from that of partners A and B. Now we choose the order of the former is less than the latter. The augmented functional system can be easily obtained as shown in Section 4.3. The

augmented functional system becomes

$$\dot{F} = CF + g(F) \tag{4.5}$$

where  $F = [F_1, F_2, \dots F_n]^T \in \mathbb{R}^n$  denotes a state vector, C is an  $n \times n$  constant coefficient matrix, g is a nonlinear vector function. PCSS demands:

$$y = H(x, y, t) + F(t)$$

(4.6)

where H(x, y, t) consists of state vector x of partner A and state vector y of partner B. Our goal is to accomplish Eq. (4.6) via controller u(t) and parameter update dynamics. Define the error vector e:

$$e = H(x, y, t) - y_u + F(t)$$

$$(4.7)$$

The synchronization is achieved when ES

$$\lim_{t \to \infty} e_i = 0 \qquad (i = 1, 2, ..., n)$$
(4.8)

The error dynamics is

$$\dot{e} = \frac{\partial H}{\partial x}\dot{x} + \frac{\partial H}{\partial y}\dot{y} + \frac{\partial H}{\partial t} - \dot{y}_u + \dot{F}(t)$$
(4.9)

By Eq.  $(4.2) \sim \text{Eq. } (4.5)$ , Eq. (4.9) becomes

$$\dot{e} = \frac{\partial H}{\partial x} \left[ Ax + f(x, B) \right] + \frac{\partial H}{\partial y} \left[ \hat{A}y + f(y, \hat{B}) \right] + \frac{\partial H}{\partial t}$$

$$- \hat{A}y - f(y, \hat{B}) - u(t) + CF + g(F)$$
(4.10)

In order to reduce terms of the u(t), NDSC is used which makes u(t) more simple. This method extends the traditional dynamic surface control [44]. A virtual controller W is chosen as follows

$$m\dot{W} + W = H(x, y, t) + F(t),$$
  $\lim_{t \to \infty} W(t) = H(x, y, t) + F(t)$  (4.11)

The m is a time function to be determined. Define the boundary layer errors as

$$s = W - H(x, y, t) + F(t)$$
(4.12)

Its derivative is

$$\dot{s} = \frac{-s}{m} - \frac{d}{dt} \left( H(x, y, t) + F(t) \right) \tag{4.13}$$

Eq. (4.7) and Eq. (4.10) becomes

$$e = W - y_u \tag{4.14}$$

$$\dot{e} = \dot{W} - \hat{A}y - f\left(y, \hat{B}\right) - u\left(t\right) \tag{4.15}$$

A Lyapunov function  $V(e,s,\tilde{A}_c,\tilde{B}_c)$  is chosen as a positive definite function of  $e,\ s,\ \tilde{A}_c,\ \tilde{B}_c$ :

$$V(e, s, \tilde{A}_c, \tilde{B}_c) = \frac{1}{2}e^T e + \frac{1}{2}s^T s + \frac{1}{2}\tilde{A}_c^T \tilde{A}_c + \frac{1}{2}\tilde{B}_c^T \tilde{B}_c$$
 (4.16)

where  $\tilde{A} = A - \hat{A}$ ,  $\tilde{B} = B - \hat{B}$ ,  $\tilde{A}_c$  and  $\tilde{B}_c$  are two vectors whose elements are all the elements of matrix  $\tilde{A}$  and of matrix  $\tilde{B}$ , respectively. Its time derivative along any solution of Eq. (4.13) and Eq. (4.15) and parameter update differential equations for  $\tilde{A}_c$  and  $\tilde{B}_c$  is  $\dot{V}$ . Choose u(t), m(t),  $\dot{\tilde{A}}_c$  and  $\dot{\tilde{B}}_c$  so that

$$\dot{V} = e^T P e + s^T Q s \tag{4.17}$$

where P and Q are diagonal negative definite matrixes, and  $\dot{V}$  is a negative semi-definite function of  $e,s,\tilde{A}_c,\tilde{B}_c$ . In the current scheme of adaptive synchronization [24-28], the traditional Lyapunov stability theorem and Babalat lemma are used to prove that the error vector approaches zero, as time approaches infinity. But the question that why the estimated parameters also approach uncertain parameters remains unanswered. By the pragmatical asymptotical stability theorem, the question can be answered strictly. The equilibrium point  $e=s=\tilde{A}=\tilde{B}=0$  is pragmatically asymptotically stable (see Appendix). Under the assumption of equal

probability, it is actually asymptotically stable. Hence, the PCSS can be achieved.

### 4.2 Numerical Results for the PCSS by New Dynamic

#### **Surface Control**

Since the partner A, new Duffing-Van der Pol system, is described as

$$\begin{cases} \frac{dx_1}{dt} = x_2 \\ \frac{dx_2}{dt} = -x_1 - x_1^3 - ax_2 + dx_3 \\ \frac{dx_3}{dt} = x_4 \\ \frac{dx_4}{dt} = -bx_3 + c\left(1 - x_3^2\right)x_4 + fx_1 \end{cases}$$
(4.18)

where a, b, c, d, f are uncertain parameters. The partner B is described as

$$\begin{cases} \frac{dy_1}{dt} = y_2 \\ \frac{dy_2}{dt} = -y_1 - y_1^3 - \hat{a}y_2 + \hat{d}y_3 \\ \frac{dy_3}{dt} = y_4 \\ \frac{dy_4}{dt} = -\hat{b}y_3 + \hat{c}(1 - y_3^2)y_4 + \hat{f}y_1 \end{cases}$$
(4.19)

where  $\hat{a}$ ,  $\hat{b}$ ,  $\hat{c}$ ,  $\hat{d}$  and  $\hat{f}$  are estimated parameters.

For this scheme,  $u_1$ ,  $u_2$ ,  $u_3$  and  $u_4$  are added to the partner B then becomes controlled partner B:

$$\begin{cases} \frac{dy_1}{dt} = y_2 + u_1 \\ \frac{dy_2}{dt} = -y_1 - y_1^3 - \hat{a}y_2 + \hat{d}y_3 + u_2 \\ \frac{dy_3}{dt} = y_4 + u_3 \\ \frac{dy_4}{dt} = -\hat{b}y_3 + \hat{c}(1 - y_3^2)y_4 + \hat{f}y_1 + u_4 \end{cases}$$

$$(4.20)$$

The chaotic Lü system is chosen as functional system [45] and the augmented state

variable is  $z_4 = z_1^2$ :

$$\begin{cases} \frac{dz_{1}}{dt} = g(z_{2} - z_{1}) \\ \frac{dz_{2}}{dt} = -z_{1}z_{3} + hz_{2} \\ \frac{dz_{3}}{dt} = z_{1}z_{2} - kz_{3} \\ \frac{dz_{4}}{dt} = 2gz_{1}(z_{2} - z_{1}) \end{cases}$$

$$(4.21)$$

In the PCSS, we select the

$$H_{i}(x,y,t) = (-x_{i})^{j} y_{i} + z_{i}^{j}$$

$$\begin{cases} i = 1,2,\dots, n \\ j = \begin{cases} 2, & i = even \\ 3, & i = odd \end{cases}$$
(4.22)

Now n=4. By NDSC, the error dynamics Eq. (4.15) becomes:

$$\begin{cases} \dot{e}_{1} = \dot{W}_{1} - y_{2} - u_{1} \\ \dot{e}_{2} = \dot{W}_{2} + y_{1} + y_{1}^{3} + \hat{a}y_{2} - \hat{d}y_{3} - u_{2} \\ \dot{e}_{3} = \dot{W}_{3} - y_{4} - u_{3} \\ \dot{e}_{4} = \dot{W}_{4} + \hat{b}y_{3} - \hat{c}(1 - y_{3}^{2})y_{4} - \hat{f}y_{1} - u_{4} \end{cases}$$

$$(4.23)$$

and the boundary layer error dynamics Eq. (4.13) becomes:

$$\begin{vmatrix}
\dot{s}_{1} = \frac{-s_{1}}{m_{1}} - \left[ -3x_{1}^{2}x_{2}y_{1} - x_{1}^{3}y_{2} + 3gz_{1}^{2}(z_{2} - z_{1}) \right] \\
\dot{s}_{2} = \frac{-s_{2}}{m_{2}} - \left[ -2x_{2}y_{2}(-x_{1} - x_{1}^{3} - ax_{2} + dx_{3}) - x_{2}^{2}(-y_{1} - y_{1}^{3} - \hat{a}y_{2} + \hat{d}y_{3}) + 2z_{2}(-z_{1}z_{3} + hz_{2}) \right] \\
\dot{s}_{3} = \frac{-s_{3}}{m_{3}} - \left[ -3x_{3}^{2}x_{4}y_{3} - x_{3}^{3}y_{4} + 3z_{3}^{2}(z_{1}z_{2} - kz_{3}) \right] \\
\dot{s}_{4} = \frac{-s_{4}}{m_{4}} - \left[ -2x_{4}y_{4}(-bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1}) - x_{4}^{2}(-\hat{b}y_{3} + \hat{c}(1 - y_{3}^{2})y_{4} + \hat{f}y_{1}) + 4gz_{1}z_{4}(z_{2} - z_{1}) \right]$$
(4.24)

Choose a positive definite Lyapunov function for  $e_1$ ,  $e_2$ ,  $e_3$ ,  $e_4$ ,  $s_1$ ,  $s_2$ ,  $s_3$ ,  $s_4$ ,  $\tilde{a}$ ,  $\tilde{b}$ ,  $\tilde{c}$ ,  $\tilde{d}$ ,  $\tilde{f}$ :

$$V = \frac{1}{2} (e_1^2 + e_2^2 + e_3^2 + e_4^2 + s_1^2 + s_2^2 + s_3^2 + s_4^2 + \tilde{a}^2 + \tilde{b}^2 + \tilde{c}^2 + \tilde{d}^2 + \tilde{f}^2)$$
 (4.25)

where  $\tilde{a} = (a - \hat{a})$ ,  $\tilde{b} = (b - \hat{b})$ ,  $\tilde{c} = (c - \hat{c})$ ,  $\tilde{d} = (d - \hat{d})$ , and  $\tilde{f} = (f - \hat{f})$ . We

select controllers, estimated parameter dynamics, and the *m* as:

$$\begin{cases} u_{1} = \dot{W}_{1} - y_{2} + e_{1} \\ u_{2} = \dot{W}_{2} + y_{1} + y_{1}^{3} + \hat{a}y_{2} - \hat{d}y_{3} + e_{2} \\ u_{3} = \dot{W}_{3} - y_{4} + e_{3} \\ u_{4} = \dot{W}_{4} + \hat{b}y_{3} - \hat{c}(1 - y_{3}^{2})y_{4} - \hat{f}y_{1} + e_{4} \end{cases}$$

$$(4.26)$$

They are more simple.

$$\begin{cases}
\dot{\hat{a}} = -2x_2^2 y_2 s_2 \\
\dot{\hat{b}} = -2x_3 x_4 y_4 s_4 \\
\dot{\hat{c}} = 2x_4^2 y_4 \left(1 - x_3^2\right) s_4 \\
\dot{\hat{d}} = 2x_2 x_3 y_2 s_2 \\
\dot{\hat{f}} = 2x_1 x_4 y_4 s_4
\end{cases} \tag{4.27}$$

$$\begin{cases}
 m_{1} = \frac{-s_{1}}{-s_{1} - 3x_{1}^{2}x_{2}y_{1} - x_{1}^{3}y_{2} + 3gz_{1}^{2}(z_{2} - z_{1})} \\
 m_{2} = \frac{-s_{2}}{-s_{2} - 2x_{2}y_{2}(-x_{1} - x_{1}^{3} - \hat{a}x_{2} + \hat{d}x_{3}) - x_{2}^{2}(-y_{1} - y_{1}^{3} - \hat{a}y_{2} + \hat{d}y_{3}) + 2z_{2}(-z_{1}z_{3} + hz_{2})} \\
 m_{3} = \frac{-s_{3}}{-s_{3} - 3x_{3}^{2}x_{4}y_{3} - x_{3}^{3}y_{4} + 3z_{3}^{2}(z_{1}z_{2} - kz_{3})} \\
 m_{4} = \frac{-s_{4}}{-s_{4} - 2x_{4}y_{4}(-\hat{b}x_{3} + \hat{c}(1 - x_{3}^{2})x_{4} + \hat{f}x_{1}) - x_{4}^{2}(-\hat{b}y_{3} + \hat{c}(1 - y_{3}^{2})y_{4} + \hat{f}y_{1}) + 4gz_{1}z_{4}(z_{2} - z_{1})}
\end{cases}$$

$$(4.28)$$

The time derivative of V is

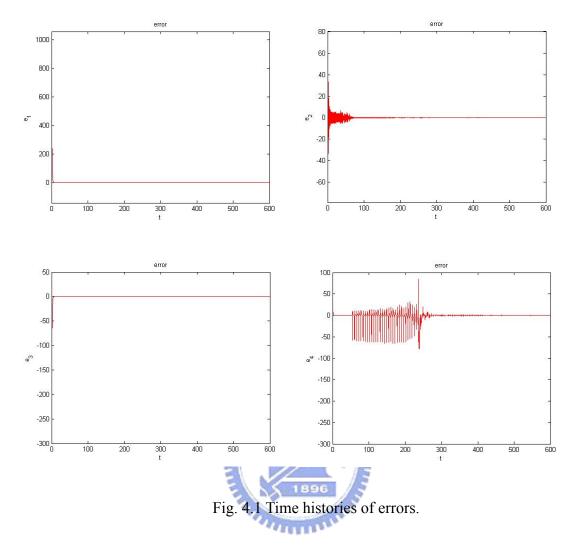
$$\dot{V} = -e_1^2 - e_2^2 - e_3^2 - e_4^2 - s_1^2 - s_2^2 - s_3^2 - s_4^2 \qquad \le 0$$
(4.29)

which is negative semi-definite function for  $e_1$ ,  $e_2$ ,  $e_3$ ,  $e_4$ ,  $s_1$ ,  $s_2$ ,  $s_3$ ,  $s_4$ ,  $\tilde{a}$ ,  $\tilde{b}$ ,  $\tilde{c}$ ,  $\tilde{d}$ ,  $\tilde{f}$ . The Lyapunov asymptotical stability theorem cannot be satisfied in this case. The common origin of error dynamics, parameter update dynamics, and boundary layer error dynamics cannot be concluded to be asymptotically stable. By pragmatical asymptotical theorem, D is a 13-manifold, n = 13 and the number of error state variables p = 8. When  $e_1 = e_2 = e_3 = e_4 = s_1 = s_2 = s_3 = s_4 = 0$  and  $\tilde{a}$ ,  $\tilde{b}$ ,  $\tilde{c}$ ,  $\tilde{d}$ ,  $\tilde{f}$  take arbitrary values,  $\dot{V} = 0$ , so X is a 5-manifold, m = n - p = 13 - 8 = 5. m + 1 < n are satisfied. By the pragmatical asymptotical stability theorem, the common origin of error dynamics (4.23), boundary layer error dynamics (4.24), and parameter dynamics (4.27) are asymptotically stable. The equilibrium point  $e_1 = e_2$ 

=  $e_3$  =  $e_4$  =  $s_1$  =  $s_2$  =  $s_3$  =  $s_4$  =  $\tilde{a}$  =  $\tilde{b}$  =  $\tilde{c}$  =  $\tilde{d}$  =  $\tilde{f}$  = 0 is pragmatically asymptotically stable. The PCSS is achieved under this scheme.

In this numerical simulation, we select the "unknown" parameter and initial states of the partner A and of functional system as a=0.01, b=1, c=5, d=0.67, f=0.05, g=36, h=20, k=3 to ensure the chaotic behavior. The initial states of those system are  $x_1(0)=2$ ,  $x_2(0)=2.4$ ,  $x_3(0)=5$ ,  $x_4(0)=6$ ,  $y_1(0)=5$ ,  $y_2(0)=5$ ,  $y_3(0)=10$ ,  $y_4(0)=10$ ,  $z_1(0)=z_2(0)=z_3(0)=z_4(0)=10$ . The estimated parameters have initial conditions  $\hat{a}(0)=\hat{b}(0)=\hat{c}(0)=\hat{d}(0)=\hat{f}(0)=0$ . The numerical results are shown in Fig. 4.1 ~ Fig. 4.4.





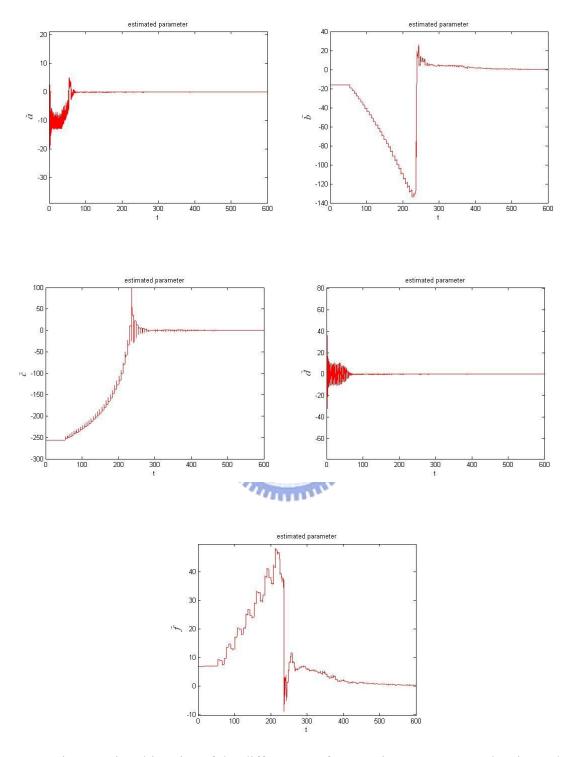


Fig. 4.2 Time histories of the differences of uncertain parameters and estimated parameters.

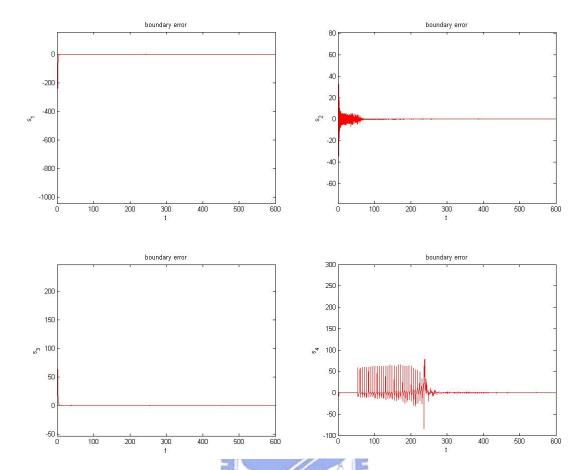


Fig. 4.3 Time histories of boundary layer errors.

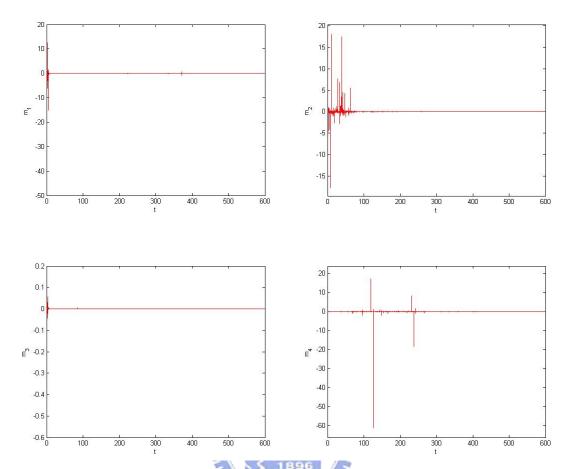


Fig. 4.4 Time histories of *m* which is a bounded function of time and approaches to zero.

## Chapter 5

# Chaos Generalized Synchronization of New Duffing-Van der Pol Systems by GYC Partial Region Stability Theory

A new chaos generalized synchronization strategy, using the GYC partial region stability theory, the controllers are of lower degree than that of controllers by using traditional Lyapunov asymptotical stability theorem. The simple linear homogeneous Lyapunov function of error states makes the controllers introducing less simulation error. A new Duffing-Van der Pol system and hyper-chaotic Lü system [46] are used as simulated examples.

# 5.1 Chaos Generalized Synchronization Strategy

Consider the following unidirectional coupled chaotic systems

$$\dot{\mathbf{x}} = \mathbf{f}(t, \mathbf{x})$$

$$\dot{\mathbf{y}} = \mathbf{h}(t, \mathbf{y}) + \mathbf{u}$$
(5.1)

where  $\mathbf{x} = \begin{bmatrix} x_1, x_2, \cdots, x_n \end{bmatrix}^T \in R^n$ ,  $\mathbf{y} = \begin{bmatrix} y_1, y_2, \cdots, y_n \end{bmatrix}^T \in R^n$  denote the master state vector and slave state vector respectively,  $\mathbf{f}$  and  $\mathbf{h}$  are nonlinear vector functions, and  $\mathbf{u} = \begin{bmatrix} u_1, u_2, \cdots, u_n \end{bmatrix}^T \in R^n$  is a control input vector.

The generalized synchronization can be accomplished when  $t \to \infty$ , the limit of the error vector  $\mathbf{e} = [e_1, e_2, \cdots, e_n]^T$  approaches zero:

$$\lim_{t \to \infty} \mathbf{e} = 0 \tag{5.2}$$

where

$$\mathbf{e} = \mathbf{G}(\mathbf{x}) - \mathbf{y} \tag{5.3}$$

G(x) is a given function of x.

By using the partial region stability theory (see Appendix), the linear homogeneous terms of the entries of **e** can be used to construct a positive definite Lyapunov function and the controllers can be designed in lower degree.

#### 5.2 Numerical Simulations

Two new Duffing-Van der Pol systems with unidirectional coupling are given:

$$\begin{cases} \dot{x}_{1} = x_{2} \\ \dot{x}_{2} = -x_{1} - x_{1}^{3} - ax_{2} + dx_{3} \\ \dot{x}_{3} = x_{4} \\ \dot{x}_{4} = -bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1} \end{cases}$$

$$\begin{cases} \dot{y}_{1} = y_{2} + u_{1} \\ \dot{y}_{2} = -y_{1} - y_{1}^{3} - ay_{2} + dy_{3} + u_{2} \\ \dot{y}_{3} = y_{4} + u_{3} \\ \dot{y}_{4} = -by_{3} + c(1 - y_{3}^{2})y_{4} + fy_{1} + u_{4} \end{cases}$$

$$(5.4)$$

CASE I. The generalized synchronization error function is

$$e_i = x_i - y_i + 30,$$
  $i = 1, 2, 3, 4$  (5.5)

The addition of the constant 30 makes the error dynamics always happens in the first quadrant. Our goal is  $y_i = x_i + 30$ , i.e.

$$\lim_{t \to \infty} e_i = \lim_{t \to \infty} (x_i - y_i + 30) = 0, \qquad i = 1, 2, 3, 4$$
(5.6)

The error dynamics becomes

$$\begin{cases} \dot{e}_{1} = \dot{x}_{1} - \dot{y}_{1} = x_{2} - y_{2} + x_{1}^{2} - x_{1}^{2} - u_{1} \\ \dot{e}_{2} = \dot{x}_{2} - \dot{y}_{2} = -x_{1} - x_{1}^{3} - ax_{2} + dx_{3} - (-y_{1} - y_{1}^{3} - ay_{2} + dy_{3}) + x_{1}^{2} - x_{1}^{2} - u_{2} \\ \dot{e}_{3} = \dot{x}_{3} - \dot{y}_{3} = x_{4} - y_{4} + x_{3}^{2} - x_{3}^{2} - u_{3} \\ \dot{e}_{4} = \dot{x}_{4} - \dot{y}_{4} = -bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1} - (-by_{3} + c(1 - y_{3}^{2})y_{4} + fy_{1}) + x_{3}^{2} - x_{3}^{2} - u_{4} \end{cases}$$

$$(5.7)$$

Let initial states be  $(x_1, x_2, x_3, x_4) = (2, 2.4, 5, 6), (y_1, y_2, y_3, y_4) = (5, 5, 1, 1)$  and system parameters a = 0.01, b = 1, c = 5, d = 0.67, f = 0.05, we find that the error

dynamic always exists in first quadrant as shown in Fig. 5.1. By GYC partial region asymptotical stability theory, one can choose a Lyapunov function in the form of a positive definite function in first quadrant:

$$V = e_1 + e_2 + e_3 + e_4 (5.8)$$

Its time derivative is

$$\dot{V} = \dot{e}_{1} + \dot{e}_{2} + \dot{e}_{3} + \dot{e}_{4} 
= \left(x_{2} - y_{2} + x_{1}^{2} - x_{1}^{2} - u_{1}\right) 
+ \left(-x_{1} - x_{1}^{3} - ax_{2} + dx_{3} + y_{1} + y_{1}^{3} + ay_{2} - dy_{3} + x_{1}^{2} - x_{1}^{2} - u_{2}\right) 
+ \left(x_{4} - y_{4} + x_{3}^{2} - x_{3}^{2} - u_{3}\right) 
+ \left(-bx_{3} + c\left(1 - x_{3}^{2}\right)x_{4} + fx_{1} + by_{3} - c\left(1 - y_{3}^{2}\right)y_{4} - fy_{1} + x_{3}^{2} - x_{3}^{2} - u_{4}\right)$$
(5.9)

Choose

$$\begin{cases} u_{1} = x_{2} - y_{2} + x_{1}^{2} + e_{1} \\ u_{2} = -x_{1} - x_{1}^{3} - ax_{2} + dx_{3} + y_{1} + y_{1}^{3} + ay_{2} - dy_{3} - x_{1}^{2} + e_{2} \\ u_{3} = x_{4} - y_{4} + x_{3}^{2} + e_{3} \\ u_{4} = -bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1} + by_{3} - c(1 - y_{3}^{2})y_{4} - fy_{1} - x_{3}^{2} + e_{4} \end{cases}$$
(5.10)

We obtain

$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0 \tag{5.11}$$

which is negative definite function in the first quadrant. Four state errors versus time and time histories of states are shown in Fig. 5.2 and Fig. 5.3.

CASE II. The generalized synchronization error function is

$$e_i = x_i - y_i + F \sin \omega t \cos \omega t + 50,$$
  $i = 1, 2, 3, 4$  (5.12)

Our goal is  $y_i = x_i + F \sin \omega t \cos \omega t + 50$ , i.e.

$$\lim_{t \to \infty} e_i = \lim_{t \to \infty} (x_i - y_i + F \sin \omega t \cos \omega t + 50) = 0, \qquad i = 1, 2, 3, 4$$
 (5.13)

The error dynamics becomes

$$\dot{e}_i = \dot{x}_i - \dot{y}_i + F\omega\cos^2\omega t - F\omega\sin^2\omega t$$

$$\begin{cases} \dot{e}_{1} = x_{2} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t - y_{2} + x_{2}^{2} - x_{2}^{2} - u_{1} \\ \dot{e}_{2} = -x_{1} - x_{1}^{3} - ax_{2} + dx_{3} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t \\ -(-y_{1} - y_{1}^{3} - ay_{2} + dy_{3}) + x_{2}^{2} - x_{2}^{2} - u_{2} \end{cases}$$

$$\dot{e}_{3} = x_{4} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t - y_{4} + x_{4}^{2} - x_{4}^{2} - u_{3}$$

$$\dot{e}_{4} = -bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t - (-by_{3} + c(1 - y_{3}^{2})y_{4} + fy_{1}) + x_{4}^{2} - x_{4}^{2} - u_{4}$$

$$(5.14)$$

Let initial states be  $(x_1, x_2, x_3, x_4) = (2, 2.4, 5, 6), (y_1, y_2, y_3, y_4) = (5, 5, 1, 1)$  and system parameters a = 0.01, b = 1, c = 5, d = 0.67, f = 0.05, F = 5 and  $\omega = 0.2$ , we find that the error dynamics always exists in first quadrant as shown in Fig. 5.4. By GYC partial region asymptotical stability theory, one can choose a Lyapunov function in the form of a positive definite function in first quadrant:

$$V = e_1 + e_2 + e_3 + e_4$$
 (5.15)  
me derivative is

Its time derivative is

$$\dot{V} = \left(x_{2} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t - y_{2} + x_{2}^{2} - x_{2}^{2} - u_{1}\right) + \\
+ \left(-x_{1} - x_{1}^{3} - ax_{2} + dx_{3} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t + y_{1} + y_{1}^{3} + ay_{2} - dy_{3} + x_{2}^{2} - x_{2}^{2} - u_{2}\right) \\
+ \left(x_{4} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t - y_{4} + x_{4}^{2} - x_{4}^{2} - u_{3}\right) \\
+ \left(-bx_{3} + c\left(1 - x_{3}^{2}\right)x_{4} + fx_{1} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t + by_{3}\right) \\
- c\left(1 - y_{3}^{2}\right)y_{4} - fy_{1} + x_{4}^{2} - x_{4}^{2} - u_{4}\right)$$
(5.16)

Choose

$$\begin{cases} u_{1} = x_{2} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t - y_{2} + x_{2}^{2} + e_{1} \\ u_{2} = -x_{1} - x_{1}^{3} - \alpha x_{2} + dx_{3} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t + y_{1} + y_{1}^{3} + \alpha y_{2} - dy_{3} - x_{2}^{2} + e_{2} \\ u_{3} = x_{4} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t - y_{4} + x_{4}^{2} + e_{3} \\ u_{4} = -bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1} + F\omega\cos^{2}\omega t - F\omega\sin^{2}\omega t + by_{3} \\ -c(1 - y_{3}^{2})y_{4} - fy_{1} - x_{4}^{2} + e_{4} \end{cases}$$

$$(5.17)$$

We obtain

$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0 \tag{5.18}$$

which is negative definite function in first quadrant. Four state errors versus time and time histories of states are shown in Fig. 5.5 and Fig. 5.6.

CASE III. The generalized synchronization error function is

$$e_i = \frac{1}{10}x_i^3 - y_i + 80,$$
  $i = 1, 2, 3, 4$  (5.19)

Our goal is  $y_i = \frac{1}{10}x_i^3 + 80$ , i.e.

$$\lim_{t \to \infty} e_i = \lim_{t \to \infty} (\frac{1}{10} x_i^3 - y_i + 80) = 0, \qquad i = 1, 2, 3, 4$$
 (5.20)

The error dynamics become

$$\dot{e}_{i} = \frac{3}{10}x_{i}^{2}\dot{x}_{i} - \dot{y}_{i}$$

$$\begin{vmatrix}
\dot{e}_{1} = \frac{3}{10}x_{1}^{2}x_{2} - y_{2} + x_{1}^{2} - x_{1}^{2} - u_{1} \\
\dot{e}_{2} = \frac{3}{10}x_{2}^{2}\left(-x_{1} - x_{1}^{3} - ax_{2} + dx_{3}\right) - \left(-y_{1} - y_{1}^{3} - ay_{2} + dy_{3}\right) + x_{1}^{2} - x_{1}^{2} - u_{2}$$

$$\dot{e}_{3} = \frac{3}{10}x_{3}^{2}x_{4} - y_{4} + x_{3}^{2} - x_{3}^{2} - u_{3}$$

$$\dot{e}_{4} = \frac{3}{10}x_{4}^{2}\left(-bx_{3} + c\left(1 - x_{3}^{2}\right)x_{4} + fx_{1}\right) - \left(-by_{3} + c\left(1 - y_{3}^{2}\right)y_{4} + fy_{1}\right) + x_{3}^{2} - x_{3}^{2} - u_{4}$$
(5.21)

Let initial states be  $(x_1, x_2, x_3, x_4) = (2, 2.4, 5, 6)$ ,  $(y_1, y_2, y_3, y_4) = (5, 5, 1, 1)$  and system parameters a = 0.01, b = 1, c = 5, d = 0.67, f = 0.05, we find the error dynamics always exists in first quadrant as shown in Fig. 5.7. By GYC partial region asymptotical stability theory, one can choose a Lyapunov function in the form of a positive definite function in first quadrant:

$$V = e_1 + e_2 + e_3 + e_4 (5.22)$$

Its time derivative is

$$\dot{V} = \left(\frac{3}{10}x_1^2x_2 - y_2 + x_1^2 - x_1^2 - u_1\right) 
+ \left(\frac{3}{10}x_2^2\left(-x_1 - x_1^3 - ax_2 + dx_3\right) + y_1 + y_1^3 + ay_2 - dy_3 + x_1^2 - x_1^2 - u_2\right) 
+ \left(\frac{3}{10}x_3^2x_4 - y_4 + x_3^2 - x_3^2 - u_3\right) 
+ \left(\frac{3}{10}x_4^2\left(-bx_3 + c\left(1 - x_3^2\right)x_4 + fx_1\right) + by_3 - c\left(1 - y_3^2\right)y_4 - fy_1 + x_3^2 - x_3^2 - u_4\right)$$
(5.23)

Choose

$$\begin{cases} u_{1} = \frac{3}{10} x_{1}^{2} x_{2} - y_{2} + x_{1}^{2} + e_{1} \\ u_{2} = \frac{3}{10} x_{2}^{2} \left( -x_{1} - x_{1}^{3} - ax_{2} + dx_{3} \right) + y_{1} + y_{1}^{3} + ay_{2} - dy_{3} - x_{1}^{2} + e_{2} \\ u_{3} = \frac{3}{10} x_{3}^{2} x_{4} - y_{4} + x_{3}^{2} + e_{3} \\ u_{4} = \frac{3}{10} x_{4}^{2} \left( -bx_{3} + c\left(1 - x_{3}^{2}\right)x_{4} + fx_{1} \right) + by_{3} - c\left(1 - y_{3}^{2}\right)y_{4} - fy_{1} - x_{3}^{2} + e_{4} \end{cases}$$

$$(5.24)$$

obtain 
$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0 \tag{5.25}$$

which is negative definite function in first quadrant. Four state errors versus time and time histories of states are shown in Fig. 5.8 and Fig. 5.9.

CASE IV. The generalized synchronization error function is

$$e_i = x_i - y_i + \frac{1}{2}z_i + 150,$$
  $i = 1, 2, 3, 4$  (5.26)

 $z = \begin{bmatrix} z_1 & z_2 & z_3 & z_4 \end{bmatrix}^T$  is the state vector of hyperchaotic Lü system.

The goal system for synchronization is hyperchaotic Lü system and initial states is (1, 1, 1, 1), system parameters  $a_1 = 36$ ,  $b_1 = 20$ ,  $c_1 = 3$ ,  $d_1 = 1.3$ .

$$\begin{cases} \dot{z}_{1} = a_{1}(z_{2} - z_{1}) + z_{4} \\ \dot{z}_{2} = b_{1}z_{2} - z_{1}z_{3} \\ \dot{z}_{3} = -c_{1}z_{3} + z_{1}z_{2} \\ \dot{z}_{4} = d_{1}z_{4} + z_{1}z_{3} \end{cases}$$
(5.27)

We have

$$\lim_{t \to \infty} e_i = \lim_{t \to \infty} (x_i - y_i + \frac{1}{2}z_i + 150) = 0 \qquad i = 1, 2, 3, 4$$
 (5.28)

The error dynamics becomes

$$\dot{e}_i = \dot{x}_i - \dot{y}_i + \frac{1}{2}\dot{z}_i$$

$$\begin{cases} \dot{e}_{1} = x_{2} + \frac{1}{2} \left[ a_{1} (z_{2} - z_{1}) + z_{4} \right] - y_{2} + x_{2}^{2} - x_{2}^{2} - u_{1} \\ \dot{e}_{2} = -x_{1} - x_{1}^{3} - ax_{2} + dx_{3} + \frac{1}{2} (b_{1}z_{2} - z_{1}z_{3}) - \left( -y_{1} - y_{1}^{3} - ay_{2} + dy_{3} \right) + x_{2}^{2} - x_{2}^{2} - u_{2} \\ \dot{e}_{3} = x_{4} + \frac{1}{2} \left( -c_{1}z_{3} + z_{1}z_{2} \right) - y_{4} + x_{4}^{2} - x_{4}^{2} - u_{3} \\ \dot{e}_{4} = -bx_{3} + c \left( 1 - x_{3}^{2} \right) x_{4} + fx_{1} + \frac{1}{2} \left( d_{1}z_{4} + z_{1}z_{3} \right) - \left[ -by_{3} + c \left( 1 - y_{3}^{2} \right) y_{4} + fy_{1} \right] + x_{4}^{2} - x_{4}^{2} - u_{4} \end{cases}$$

$$(5.29)$$

Let initial states be  $(x_1, x_2, x_3, x_4) = (2, 2.4, 5, 6)$ ,  $(y_1, y_2, y_3, y_4) = (1, 1, 1, 1)$  and system parameters a = 0.01, b = 1, c = 5, d = 0.67, f = 0.05, we find the error dynamics always exists in first quadrant as shown in Fig. 5.10. By GYC partial region asymptotical stability theorem, one can choose a Lyapunov function in the form of a positive definite function in first quadrant:

$$V = e_1 + e_2 + e_3 + e_4 (5.30)$$

Its time derivative is

$$\dot{V} = \left(x_{2} + \frac{1}{2}\left[a_{1}(z_{2} - z_{1}) + z_{4}\right] - y_{2} + x_{2}^{2} - x_{2}^{2} - u_{1}\right) \\
+ \left(-x_{1} - x_{1}^{3} - ax_{2} + dx_{3} + \frac{1}{2}(b_{1}z_{2} - z_{1}z_{3}) + y_{1} + y_{1}^{3} + ay_{2} - dy_{3} + x_{2}^{2} - x_{2}^{2} - u_{2}\right) \\
+ \left(x_{4} + \frac{1}{2}(-c_{1}z_{3} + z_{1}z_{2}) - y_{4} + x_{4}^{2} - x_{4}^{2} - u_{3}\right) \\
+ \left(-bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1} + \frac{1}{2}(d_{1}z_{4} + z_{1}z_{3}) + by_{3} - c(1 - y_{3}^{2})y_{4} - fy_{1} + x_{4}^{2} - x_{4}^{2} - u_{4}\right)$$
(5.31)

Choose

$$\begin{cases} u_{1} = x_{2} + \frac{1}{2} \left[ a_{1} (z_{2} - z_{1}) + z_{4} \right] - y + x_{2}^{2} + e_{1} \\ u_{2} = -x_{1} - x_{1}^{3} - ax_{2} + dx_{3} + \frac{1}{2} (b_{1}z_{2} - z_{1}z_{3}) + y_{1} + y_{1}^{3} + ay_{2} - dy_{3} - x_{2}^{2} + e_{2} \\ u_{3} = x_{4} + \frac{1}{2} (-c_{1}z_{3} + z_{1}z_{2}) - y_{4} - u_{3} + x_{4}^{2} + e_{3} \\ u_{4} = -bx_{3} + c(1 - x_{3}^{2})x_{4} + fx_{1} + \frac{1}{2} (d_{1}z_{4} + z_{1}z_{3}) + by_{3} - c(1 - y_{3}^{2})y_{4} - fy_{1} - x_{4}^{2} + e_{4} \end{cases}$$

$$(5.32)$$

We obtain

$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0 \tag{5.33}$$

which is negative definite function in first quadrant. Four state errors versus time and time histories of states are shown in Fig. 5.11 and Fig. 5.12.

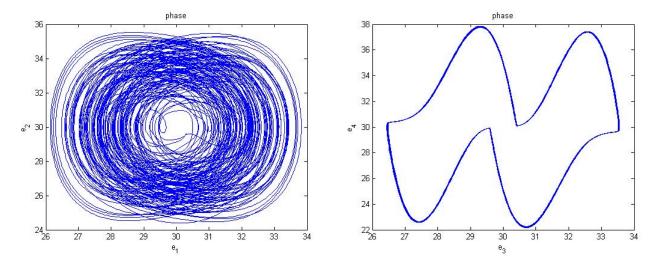


Fig. 5.1 Phase portraits of error dynamics for Case I.

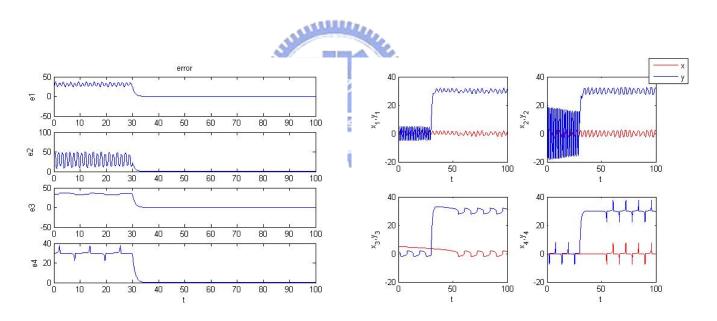


Fig. 5.2 Time histories of errors for Case I.

Fig. 5.3 Time histories of  $x_i$ ,  $y_i$  for Case I.

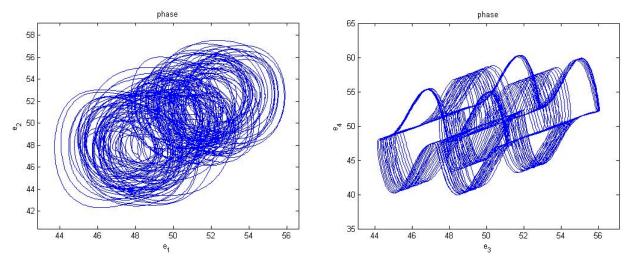


Fig. 5.4 Phase portrait of error dynamics for Case II.

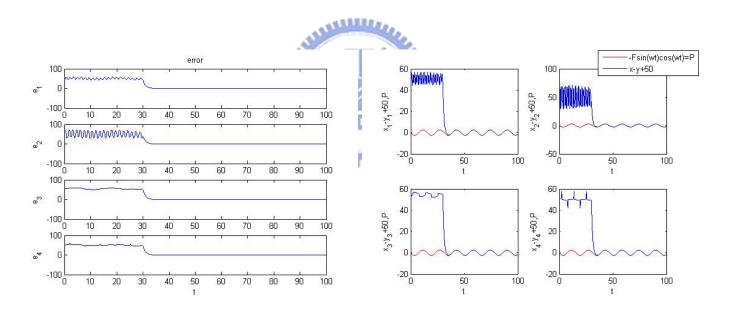


Fig. 5.5 Time histories of errors for Case II.

Fig. 5.6 Time histories of  $x_i - y_i + 50$ 

and  $-F \sin \omega t \cos \omega t$  for Case II.

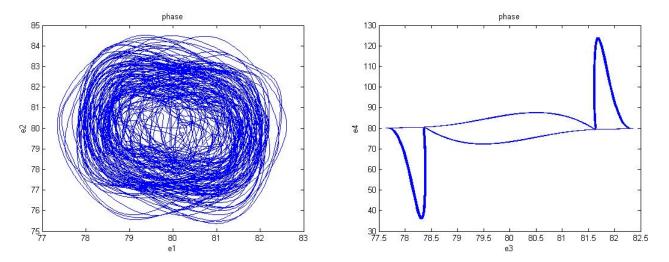


Fig. 5.7 Phase portraits of error dynamics for Case III.

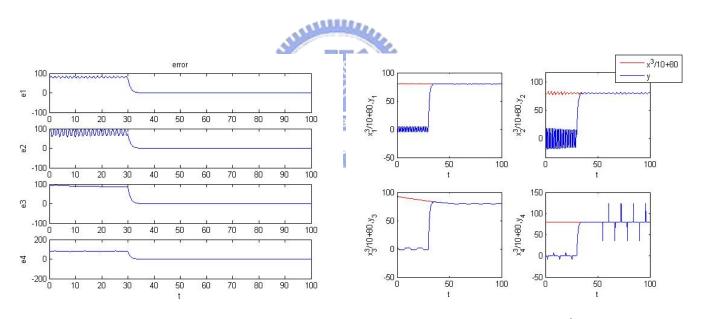


Fig. 5.8 Time histories of errors for Case III.

Fig. 5.9 Time histories of  $\frac{x_i^3}{10} + 80$  and  $y_i$  for Case III.

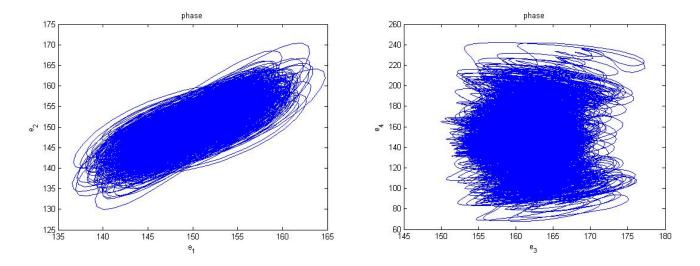


Fig. 5.10 Phase portrait of error dynamics for Case IV.

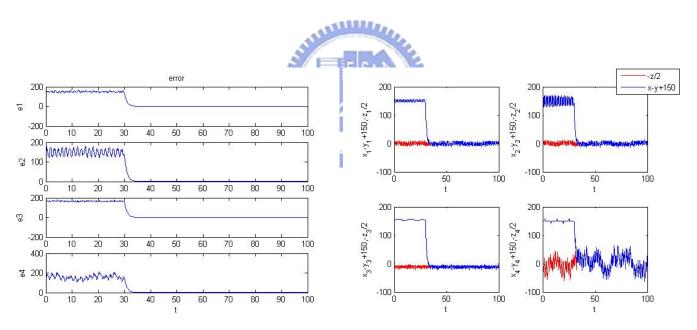


Fig. 5.11 Time histories of errors for

Fig. 5.12 Time histories of  $x_i - y_i + 150$ 

Case IV.

and 
$$-\frac{z_i}{2}$$
 for Case IV.

# Chapter 6

# Chaos Control and Anti-control of a New Duffing-Van der Pol System by GYC Partial Region Stability Theory

Using the GYC partial region stability theory, a new chaos control and anti-control strategy is proposed. The controllers are of lower degree than that of controllers by using traditional Lyapunov asymptotical stability theorem. The simple linear homogeneous Lyapunov function of error states makes the controllers introducing less simulation error. A new Duffing-Van der Pol system and hyper-chaotic Lü system are used as simulated examples.

### 6.1 Chaos Control Scheme

Consider the following chaotic systems

$$\dot{\mathbf{x}} = \mathbf{f}(t, \mathbf{x}) \tag{6.1}$$

where  $\mathbf{x} = [x_1, x_2, \dots, x_n]^T \in \mathbb{R}^n$  is a the state vector,  $\mathbf{f} : \mathbb{R}_+ \times \mathbb{R}^n \to \mathbb{R}^n$  is a vector function.

The goal system which can be either chaotic or regular, is

$$\dot{\mathbf{y}} = \mathbf{g}(t, \mathbf{y}) \tag{6.2}$$

where  $\mathbf{y} = [y_1, y_2, \dots, y_n]^T \in \mathbb{R}^n$  is a state vector,  $\mathbf{g} : \mathbb{R}_+ \times \mathbb{R}^n \to \mathbb{R}^n$  is a vector function.

In order to make the chaos state vector  $\mathbf{x}$  approaching the goal state vector  $\mathbf{y}$ , define  $\mathbf{e} = \mathbf{x} - \mathbf{y}$  as the state error. The chaos control is accomplished in the sense that [35-41]:

$$\lim_{t \to \infty} \mathbf{e} = \lim_{t \to \infty} (\mathbf{x} - \mathbf{y}) = 0 \tag{6.3}$$

In this Chapter, we will use examples in which the error dynamics is placed in the first quadrant of coordinate system and use the GYC partial region stability theory. The Lyapunov function is a simple linear homogeneous function of error states and the controllers are simpler because they are in lower degree than that of traditional controllers.

#### **6.2 Numerical Simulations for Chaos Control**

The following chaotic system is the new Duffing-Van der Pol system of which the old origin is translated to  $(x_1, x_2, x_3, x_4) = (50, 50, 50, 50, 50)$  and the chaotic motion always happens in the first quadrant of coordinate system  $(x_1, x_2, x_3, x_4)$ . This translated new Duffing-Van der Pol system is presented as simulated examples where the initial states of system are  $x_1(0) = 52, x_2(0) = 52.4, x_3(0) = 55, x_4(0) = 56$  and the parameters of system are a = 0.01, b = 1, c = 5, d = 0.67, f = 0.05. The chaotic motion is shown in Fig. 6.1.

parameters of system are 
$$a = 0.01$$
,  $b = 1$ ,  $c = 5$ ,  $d = 0.67$ ,  $f = 0.05$ . The chaotic motion is shown in Fig. 6.1.
$$\begin{cases}
\dot{x}_1 = (x_2 - 50) \\
\dot{x}_2 = -(x_1 - 50) - (x_1 - 50)^3 - a(x_2 - 50) + d(x_3 - 50) \\
\dot{x}_3 = (x_4 - 50) \\
\dot{x}_4 = -b(x_3 - 50) + c(1 - (x_3 - 50)^2)(x_4 - 50) + f(x_1 - 50)
\end{cases} (6.4)$$

In order to lead  $(x_1, x_2, x_3, x_4)$  to the goal, we add control terms  $u_1, u_2, u_3, u_4$  to each equation of Eq. (6.4), respectively.

$$\begin{cases} \dot{x}_1 = (x_2 - 50) + u_1 \\ \dot{x}_2 = -(x_1 - 50) - (x_1 - 50)^3 - a(x_2 - 50) + d(x_3 - 50) + u_2 \\ \dot{x}_3 = (x_4 - 50) + u_3 \\ \dot{x}_4 = -b(x_3 - 50) + c(1 - (x_3 - 50)^2)(x_4 - 50) + f(x_1 - 50) + u_4 \end{cases}$$

$$(6.5)$$

CASE I. Control the chaotic motion to zero.

In this case we will control the chaotic motion of the new Duffing-Van der Pol system (6.4) to zero. The goal is y = 0. The state error is e = x - y = x and error dynamics becomes

$$\begin{cases} \dot{e}_{1} = \dot{x}_{1} = (x_{2} - 50) + e_{1}^{2} - e_{1}^{2} + u_{1} \\ \dot{e}_{2} = \dot{x}_{2} = -(x_{1} - 50) - (x_{1} - 50)^{3} - a(x_{2} - 50) + d(x_{3} - 50) + e_{1}^{2} - e_{1}^{2} + u_{2} \\ \dot{e}_{3} = \dot{x}_{3} = (x_{4} - 50) + e_{2}^{2} - e_{2}^{2} + u_{3} \\ \dot{e}_{4} = \dot{x}_{4} = -b(x_{3} - 50) + c(1 - (x_{3} - 50)^{2})(x_{4} - 50) + f(x_{1} - 50) + e_{2}^{2} - e_{2}^{2} + u_{4} \end{cases}$$

$$(6.6)$$

In Fig. 6.2, we see that the error dynamics always exists in first quadrant.

By GYC partial region stability theory, one can easily choose a Lyapunov function in the form of a positive definite function in first quadrant as:

$$V = e_1 + e_2 + e_3 + e_4 \tag{6.7}$$

Its time derivative through error dynamics (6.6) is

$$\dot{V} = \dot{e}_1 + \dot{e}_2 + \dot{e}_3 + \dot{e}_4 
= (x_2 - 50) + e_1^2 - e_1^2 + u_1 - (x_1 - 50) - (x_1 - 50)^3 - a(x_2 - 50) + d(x_3 - 50) + e_1^2 - e_1^2 + u_2 
+ (x_4 - 50) + e_2^2 - e_2^2 + u_3 - b(x_3 - 50) + c(1 - (x_3 - 50)^2)(x_4 - 50) + f(x_1 - 50) + e_2^2 - e_2^2 + u_4$$
(6.8)

Choose

$$\begin{cases} u_{1} = -(x_{2} - 50) + e_{1}^{2} - e_{1} \\ u_{2} = -(-(x_{1} - 50) - (x_{1} - 50)^{3} - a(x_{2} - 50) + d(x_{3} - 50)) - e_{1}^{2} - e_{2} \\ u_{3} = -(x_{4} - 50) + e_{2}^{2} - e_{3} \\ u_{4} = -(-b(x_{3} - 50) + c(1 - (x_{3} - 50)^{2})(x_{4} - 50) + f(x_{1} - 50)) - e_{2}^{2} - e_{4} \end{cases}$$

$$(6.9)$$

We obtain

$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0$$

which is negative definite function in first quadrant. The numerical results are shown in Fig.6.3. After 100 sec, the motion trajectories approach the origin.

CASE II. Control the chaotic motion to a product of sine and cosine functions.

In this case we will control the chaotic motion of the new Duffing-Van der Pol system (6.4) to a product of sine and cosine functions of time. The goal is  $y = F \sin \omega t \cos \omega t$ . The error equation

$$\mathbf{e} = \mathbf{x} - \mathbf{y} = \mathbf{x} - F \sin \omega t \cos \omega t$$

$$\lim_{t \to \infty} e_i = \lim_{t \to \infty} (x_i - F \sin \omega_i t \cos \omega_i t) = 0, \qquad i = 1, 2, 3, 4$$

$$(6.10)$$

and 
$$\dot{e}_i = \dot{x}_i - \omega_i F \cos^2 \omega_i t + \omega_i F \sin^2 \omega_i t$$
  $(i = 1, 2, 3, 4)$  and  $F = 5$ ,  $\omega_1 = 0.2$ ,

 $\omega_2 = 0.4$ ,  $\omega_3 = 0.6$ ,  $\omega_4 = 0.8$ . The error dynamics is

$$\begin{cases} \dot{e}_{1} = \dot{x}_{1} - \omega_{1}F\cos^{2}\omega_{1}t + \omega_{1}F\sin^{2}\omega_{1}t + e_{3}^{2} - e_{3}^{2} \\ \dot{e}_{2} = \dot{x}_{2} - \omega_{2}F\cos^{2}\omega_{2}t + \omega_{2}F\sin^{2}\omega_{2}t + e_{3}^{2} - e_{3}^{2} \\ \dot{e}_{3} = \dot{x}_{3} - \omega_{3}F\cos^{2}\omega_{3}t + \omega_{3}F\sin^{2}\omega_{3}t + e_{4}^{2} - e_{4}^{2} \\ \dot{e}_{4} = \dot{x}_{4} - \omega_{4}F\cos^{2}\omega_{4}t + \omega_{4}F\sin^{2}\omega_{4}t + e_{4}^{2} - e_{4}^{2} \end{cases}$$

$$(6.11)$$

In Fig. 6.4, the error dynamics always exists in first quadrant.

By GYC partial region stability theory, one can easily choose a Lyapunov function in the form of a positive definite function in first quadrant as:

$$V = e_1 + e_2 + e_3 + e_4$$

Its time derivative is

$$\dot{V} = \dot{e}_{1} + \dot{e}_{2} + \dot{e}_{3} + \dot{e}_{4}$$

$$= (x_{2} - 50) - \omega_{1}F\cos^{2}\omega_{1}t + \omega_{1}F\sin^{2}\omega_{1}t + e_{3}^{2} - e_{3}^{2} + u_{1} - (x_{1} - 50) - (x_{1} - 50)^{3}$$

$$-a(x_{2} - 50) + d(x_{3} - 50) - \omega_{2}F\cos^{2}\omega_{2}t + \omega_{2}F\sin^{2}\omega_{2}t + e_{3}^{2} - e_{3}^{2} + u_{2} + (x_{4} - 50)$$

$$-\omega_{3}F\cos^{2}\omega_{3}t + \omega_{3}F\sin^{2}\omega_{3}t + e_{4}^{2} - e_{4}^{2} + u_{3} - b(x_{3} - 50) + c(1 - (x_{3} - 50)^{2})(x_{4} - 50)$$

$$+f(x_{1} - 50) - \omega_{4}F\cos^{2}\omega_{4}t + \omega_{4}F\sin^{2}\omega_{4}t + e_{4}^{2} - e_{4}^{2} + u_{4}$$
(6.12)

Choose

$$\begin{cases} u_{1} = -((x_{2} - 50) - \alpha_{1}F\cos^{2}\alpha_{1}t + \alpha_{2}F\sin^{2}\alpha_{1}t) + e_{3}^{2} - e_{1} \\ u_{2} = -(-(x_{1} - 50) - (x_{1} - 50)^{3} - a(x_{2} - 50) + d(x_{3} - 50) - \alpha_{2}F\cos^{2}\alpha_{2}t + \alpha_{2}F\sin^{2}\alpha_{2}t) - e_{3}^{2} - e_{2} \\ u_{3} = -((x_{4} - 50) - \alpha_{3}F\cos^{2}\alpha_{2}t + \alpha_{3}F\sin^{2}\alpha_{2}t) + e_{4}^{2} - e_{3} \\ u_{4} = -(-b(x_{3} - 50) + c(1 - (x_{3} - 50)^{2})(x_{4} - 50) + f(x_{1} - 50) - \alpha_{4}F\cos^{2}\alpha_{2}t + \alpha_{4}F\sin^{2}\alpha_{2}t) - e_{4}^{2} - e_{4} \end{cases}$$

$$(6.13)$$

We obtain

$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0$$

which is negative definite function in first quadrant. The numerical results are shown in Fig. 6.5 and Fig. 6.6. After 100 sec., the errors approach zero and the motion trajectories approach to sine and cosine functions.

CASE III. Control the chaotic motion to chaotic motion of hyper-chaotic Lü system.

In this case we will control chaotic motion of the new Duffing-Van der Pol system (6.4) to that of hyper-chaotic Lü system. The goal system is hyper-chaotic Lü system:

$$\begin{cases} \dot{z}_{1} = a_{1}(z_{2} - z_{1}) + z_{4} \\ \dot{z}_{2} = b_{1}z_{2} - z_{1}z_{3} \\ \dot{z}_{3} = -c_{1}z_{3} + z_{1}z_{2} \\ \dot{z}_{4} = d_{1}z_{4} + z_{1}z_{3} \end{cases}$$

$$(6.14)$$

The error equation is  $\mathbf{e} = \mathbf{x} - \frac{1}{5}\mathbf{z}$ . The goal is  $\lim_{t \to \infty} \mathbf{e} = 0$ . The error dynamics become

$$\begin{cases} \dot{e}_{1} = \dot{x}_{1} - \dot{z}_{1} = (x_{2} - 50) - \frac{1}{5} (a_{1}(z_{2} - z_{1}) + z_{4}) + z_{2}^{2} - z_{2}^{2} + u_{1} \\ \dot{e}_{2} = \dot{x}_{2} - \dot{z}_{2} = -(x_{1} - 50) - (x_{1} - 50)^{3} - a(x_{2} - 50) + d(x_{3} - 50) - \frac{1}{5} (b_{1}z_{2} - z_{1}z_{3}) + z_{2}^{2} - z_{2}^{2} + u_{2} \\ \dot{e}_{3} = \dot{x}_{3} - \dot{z}_{3} = (x_{4} - 50) - \frac{1}{5} (-c_{1}z_{3} + z_{1}z_{2}) + z_{4}^{2} - z_{4}^{2} + u_{3} \\ \dot{e}_{4} = \dot{x}_{4} - \dot{z}_{4} = -b(x_{3} - 50) + c(1 - (x_{3} - 50)^{2})(x_{4} - 50) + f(x_{1} - 50) - \frac{1}{5} (d_{1}z_{4} + z_{1}z_{3}) + z_{4}^{2} - z_{4}^{2} + u_{4} \end{cases}$$

$$(6.15)$$

By Fig. 6.7, we know that the error dynamics always exists in first quadrant.

By GYC partial region stability theory, one can easily choose a Lyapunov function in the form of a positive definite function in first quadrant as:

$$V = e_1 + e_2 + e_3 + e_4$$

Its time derivative is

$$\dot{V} = \dot{e}_{1} + \dot{e}_{2} + \dot{e}_{3} + \dot{e}_{4} 
= (x_{2} - 50) - \frac{1}{5} (a_{1}(z_{2} - z_{1}) + z_{4}) + z_{2}^{2} - z_{2}^{2} + u_{1} - (x_{1} - 50) - (x_{1} - 50)^{3} 
- a(x_{2} - 50) + d(x_{3} - 50) - \frac{1}{5} (b_{1}z_{2} - z_{1}z_{3}) + z_{2}^{2} - z_{2}^{2} + u_{2} + (x_{4} - 50) 
- \frac{1}{5} (-c_{1}z_{3} + z_{1}z_{2}) + z_{4}^{2} - z_{4}^{2} + u_{3} - b(x_{3} - 50) + c(1 - (x_{3} - 50)^{2})(x_{4} - 50) 
+ f(x_{1} - 50) - \frac{1}{5} (d_{1}z_{4} + z_{1}z_{3}) + z_{4}^{2} - z_{4}^{2} + u_{4}$$
(6.16)

Choose

$$\begin{cases}
 u_{1} = -\left((x_{2} - 50) - \frac{1}{5}(a_{1}(z_{2} - z_{1}) + z_{4})\right) + z_{2}^{2} - e_{1} \\
 u_{2} = -\left(-(x_{1} - 50) - (x_{1} - 50)^{3} - a(x_{2} - 50) + d(x_{3} - 50) - \frac{1}{5}(b_{1}z_{2} - z_{1}z_{3})\right) - z_{2}^{2} - e_{2} \\
 u_{3} = -\left((x_{4} - 50) - \frac{1}{5}(-c_{1}z_{3} + z_{1}z_{2})\right) + z_{4}^{2} - e_{3} \\
 u_{4} = -\left(-b(x_{3} - 50) + c\left(1 - (x_{3} - 50)^{2}\right)(x_{4} - 50) + f(x_{1} - 50) - \frac{1}{5}(d_{1}z_{4} + z_{1}z_{3})\right) - z_{4}^{2} - e_{4}
\end{cases}$$
(6.17)

We obtain

$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0$$

which is negative definite function in first quadrant. The numerical results are shown in Fig. 6.8 and Fig. 6.9 where  $a_1 = 36$ ,  $b_1 = 20$ ,  $c_1 = 3$ ,  $d_1 = 1.3$ . After 100 sec., the errors approach zero and the chaotic trajectories of the new Duffing-Van der Pol system approach to that of the hyper-chaotic Lü system.

#### 6.3 Numerical Simulations for Chaos Anti-control

In this section, we will control periodic motion of the new Duffing-Van der Pol system to that of hyper-chaotic Lü system. The new Duffing-Van der Pol system exhibits periodic motion when the parameters of system are a = 0.1, b = 1, c = 5, d = 0.67, f = 0.05 and the initial states of system are  $x_1(0) = 2, x_2(0) = 2.4, x_3(0) = 5, x_4(0) = 6$ . The following periodic system

$$\begin{cases} \dot{x}_1 = (x_2 - 50) \\ \dot{x}_2 = -(x_1 - 50) - (x_1 - 50)^3 - a(x_2 - 50) + d(x_3 - 50) \\ \dot{x}_3 = (x_4 - 50) \\ \dot{x}_4 = -b(x_3 - 50) + c(1 - (x_3 - 50)^2)(x_4 - 50) + f(x_1 - 50) \end{cases}$$
(6.18)

is the new Duffing-Van der Pol system of which the old origin is translated to  $(x_1, x_2, x_3, x_4) = (50, 50, 50, 50, 50)$  and the periodic motion always happens in the first quadrant of coordinate system  $(x_1, x_2, x_3, x_4)$ . This translated new Duffing-Van der Pol system is presented as simulated examples where the initial states of system are  $x_1(0) = 52, x_2(0) = 52.4, x_3(0) = 55, x_4(0) = 56$  and the parameters of system are a = 0.1, b = 1, c = 5, d = 0.67, f = 0.05. The periodic motion is shown in Fig. 6.10.

In order to lead  $(x_1, x_2, x_3, x_4)$  to the goal, we add control terms  $u_1, u_2, u_3, u_4$  to each equation of Eq. (6.18), respectively.

$$\begin{cases} \dot{x}_1 = (x_2 - 50) + u_1 \\ \dot{x}_2 = -(x_1 - 50) - (x_1 - 50)^3 - a(x_2 - 50) + d(x_3 - 50) + u_2 \\ \dot{x}_3 = (x_4 - 50) + u_3 \\ \dot{x}_4 = -b(x_3 - 50) + c(1 - (x_3 - 50)^2)(x_4 - 50) + f(x_1 - 50) + u_4 \end{cases}$$
(6.19)

The goal system is hyper-chaotic Lü system:

$$\begin{cases} \dot{z}_1 = a_1(z_2 - z_1) + z_4 \\ \dot{z}_2 = b_1 z_2 - z_1 z_3 \\ \dot{z}_3 = -c_1 z_3 + z_1 z_2 \\ \dot{z}_4 = d_1 z_4 + z_1 z_3 \end{cases}$$
(6.20)

The error equation is  $\mathbf{e} = \mathbf{x} - \frac{1}{5}\mathbf{z}$ . The goal is  $\lim_{t \to \infty} \mathbf{e} = 0$ . The error dynamics becomes

$$\begin{cases}
\dot{e}_{1} = \dot{x}_{1} - \dot{z}_{1} = (x_{2} - 50) - \frac{1}{5} (a_{1}(z_{2} - z_{1}) + z_{4}) + z_{1}^{2} - z_{1}^{2} + u_{1} \\
\dot{e}_{2} = \dot{x}_{2} - \dot{z}_{2} = -(x_{1} - 50) - (x_{1} - 50)^{3} - a(x_{2} - 50) + d(x_{3} - 50) - \frac{1}{5} (b_{1}z_{2} - z_{1}z_{3}) + z_{1}^{2} - z_{1}^{2} + u_{2} \\
\dot{e}_{3} = \dot{x}_{3} - \dot{z}_{3} = (x_{4} - 50) - \frac{1}{5} (-c_{1}z_{3} + z_{1}z_{2}) + z_{3}^{2} - z_{3}^{2} + u_{3} \\
\dot{e}_{4} = \dot{x}_{4} - \dot{z}_{4} = -b(x_{3} - 50) + c(1 - (x_{3} - 50)^{2})(x_{4} - 50) + f(x_{1} - 50) - \frac{1}{5} (d_{1}z_{4} + z_{1}z_{3}) + z_{3}^{2} - z_{3}^{2} + u_{4}
\end{cases}$$
(6.21)

By Fig. 6.11, we know the error dynamics always exists in first quadrant.

By GYC partial region stability theory, one can easily choose a Lyapunov function in the form of a positive definite function in first quadrant as:

$$V = e_1 + e_2 + e_3 + e_4$$

Its time derivative is

$$\dot{V} = \dot{e}_{1} + \dot{e}_{2} + \dot{e}_{3} + \dot{e}_{4}$$

$$= (x_{2} - 50) - \frac{1}{5} (a_{1}(z_{2} - z_{1}) + z_{4}) + z_{1}^{2} - z_{1}^{2} + u_{1} - (x_{1} - 50) - (x_{1} - 50)^{3}$$

$$- a(x_{2} - 50) + d(x_{3} - 50) - \frac{1}{5} (b_{1}z_{2} - z_{1}z_{3}) + z_{1}^{2} - z_{1}^{2} + u_{2} + (x_{4} - 50)$$

$$- \frac{1}{5} (-c_{1}z_{3} + z_{1}z_{2}) + z_{3}^{2} - z_{3}^{2} + u_{3} - b(x_{3} - 50) + c (1 - (x_{3} - 50)^{2})(x_{4} - 50)$$

$$+ f(x_{1} - 50) - \frac{1}{5} (d_{1}z_{4} + z_{1}z_{3}) + z_{3}^{2} - z_{3}^{2} + u_{4}$$
(6.22)

Choose

$$\begin{cases} u_{1} = -\left((x_{2} - 50) - \frac{1}{5}(a_{1}(z_{2} - z_{1}) + z_{4})\right) + z_{1}^{2} - e_{1} \\ u_{2} = -\left(-(x_{1} - 50) - (x_{1} - 50)^{3} - a(x_{2} - 50) + d(x_{3} - 50) - \frac{1}{5}(b_{1}z_{2} - z_{1}z_{3})\right) - z_{1}^{2} - e_{2} \\ u_{3} = -\left((x_{4} - 50) - \frac{1}{5}(-c_{1}z_{3} + z_{1}z_{2})\right) + z_{3}^{2} - e_{3} \\ u_{4} = -\left(-b(x_{3} - 50) + c\left(1 - (x_{3} - 50)^{2}\right)(x_{4} - 50) + f(x_{1} - 50) - \frac{1}{5}(d_{1}z_{4} + z_{1}z_{3})\right) - z_{3}^{2} - e_{4} \end{cases}$$

$$(6.23)$$

We obtain

$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0$$

which is negative definite function in first quadrant. The numerical results are shown in Fig. 6.12 and Fig. 6.13 where  $a_1 = 36$ ,  $b_1 = 20$ ,  $c_1 = 3$ ,  $d_1 = 1.3$ . After 200 sec., the errors approach zero and the periodic trajectories of the translated new Duffing-Van der Pol system approach to that of the hyper-chaotic Lü system.

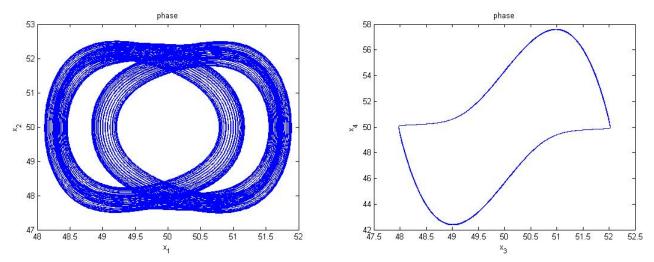


Fig. 6.1 Chaotic phase portraits for a new Duffing-Van der Pol system in the first quadrant.

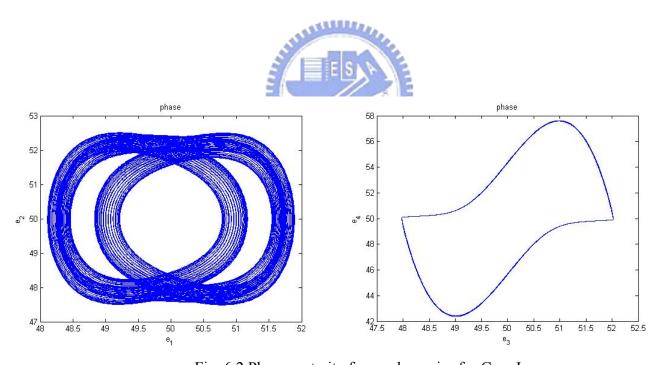


Fig. 6.2 Phase portrait of error dynamics for Case I.

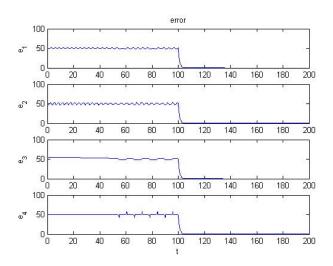


Fig.6.3 Time histories of errors for Case I.

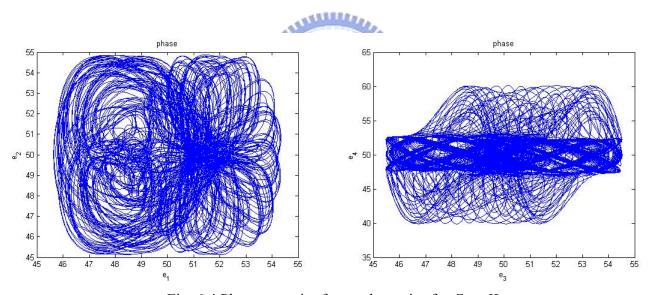


Fig. 6.4 Phase portrait of error dynamics for Case II.

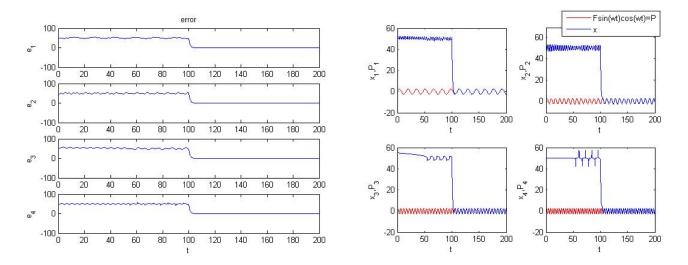


Fig. 6.5 Time histories of errors for

Fig. 6.6 Time histories of  $x_i$  for

Case II. Case II.

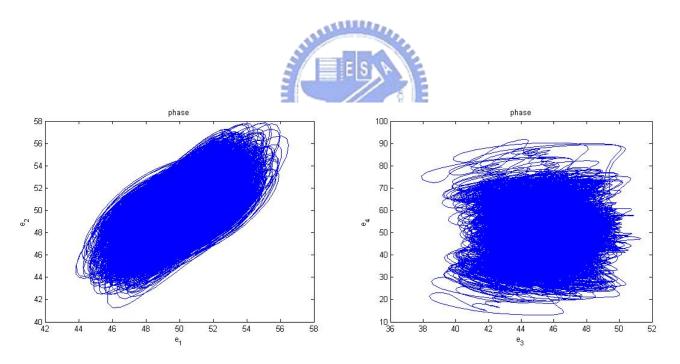


Fig. 6.7 Phase portrait of error dynamics for Case III.

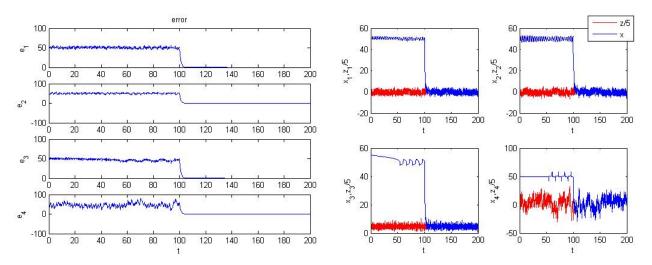


Fig. 6.8 Time histories of errors for

Fig. 6.9 Time histories of  $x_i$  for Case

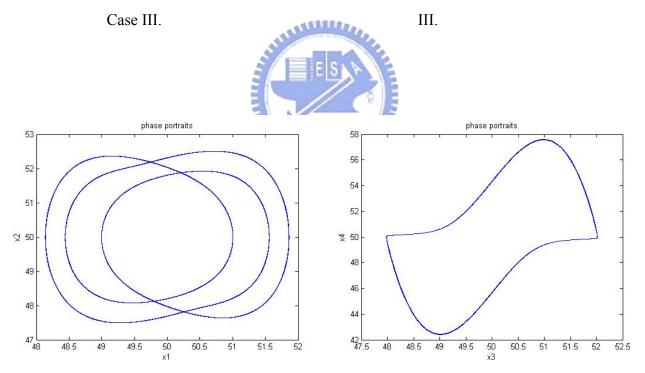


Fig. 6.10 Periodic phase portraits for a translated new Duffing-Van der Pol system in the first quadrant.

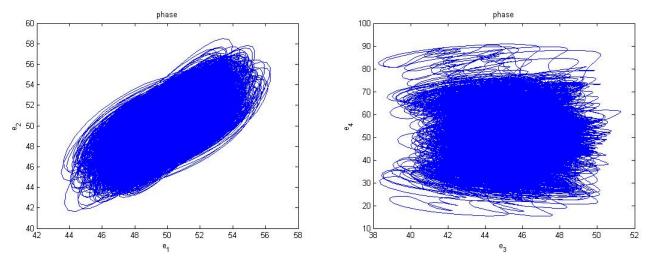


Fig. 6.11 Phase portraits of error dynamics.

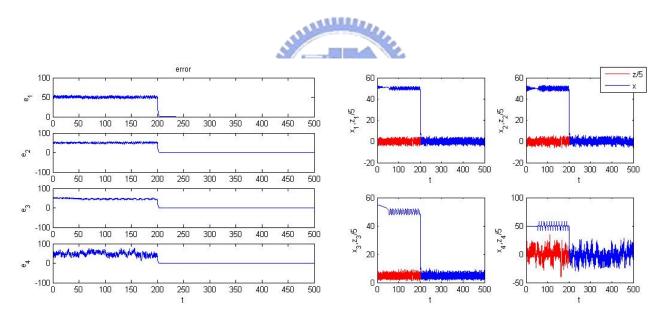


Fig. 6.12 Time histories of errors.

Fig. 6.13 Time histories of  $x_i$ .

# **Chapter 7**

# Hybrid Projective Symplectic Synchronization of a New Duffing-Van der Pol System with Legendre function Parameters by GYC Partial Region Stability Theory

A new type of chaotic synchronization, *hybrid projective symplectic synchronization* (HPSS), is obtained for a Duffing-Van der Pol system with constant parameter and a new Duffing-Van der Pol system with Legendre function parameters. The latter is used as "master" system and the former as "slave" system. Based on the GYC partial region stability theory, the scheme can be achieved not only for projective synchronization, but also for projective anti-synchronization. Numerical simulations are provided to verify the effectiveness of the proposed scheme.

#### 7.1 Hybrid Projective Symplectic Synchronization Scheme

There are two nonlinear chaotic dynamical systems. The "master" system controls the "slave" system partly. In symplectic synchronization, the "master" system is called partner A:

$$\dot{x} = f\left(x\right) \tag{7.1}$$

where  $x = [x_1, x_2, \dots x_n]^T \in \mathbb{R}^n$  is a state vector. The "slave" system is called partner B:

$$\dot{y} = f(y) \tag{7.2}$$

where  $y = [y_1, y_2, \dots y_n]^T \in \mathbb{R}^n$  is a state vector. With controllers, partner B becomes

$$\dot{y} = f(y) + u(t) \tag{7.3}$$

where  $u(t) = [u_1(t), u_2(t), \dots u_n(t)]^T \in \mathbb{R}^n$  is a control input vector. HPSS demands:

$$y = H(x, y, t) = Gxy \tag{7.4}$$

where H(x,y,t) consists of state vector x of partner A and state vector y of partner B,  $G = diag(g_1,g_2,...,g_n) \in R^{(n\times n)}$  is a constant scaling matrix with positive and negative entries. Our goal is to accomplish Eq. (7.4) via controller u(t). Define the error vector e:

$$e = H(x, y, t) - y \tag{7.5}$$

The synchronization is achieved when

$$\lim_{t \to \infty} e_i = 0 \qquad (i = 1, 2, ..., n)$$
(7.6)

The error dynamics is

$$\dot{e} = \frac{\partial H}{\partial x}\dot{x} + \frac{\partial H}{\partial y}\dot{y} + \frac{\partial H}{\partial t} - \dot{y}$$
(7.7)

By Eqs  $(7.2) \sim (7.4)$ , Eq. (7.7) becomes

$$\dot{e} = \frac{\partial H}{\partial x} \left[ f(x) \right] + \frac{\partial H}{\partial y} \left[ f(y) \right] + \frac{\partial H}{\partial t} - f(y) - u(t)$$
(7.8)

By using the GYC partial region stability theory (see Appendix), the linear homogeneous terms of the entries of **e** can be used to construct a positive definite Lyapunov function in first quadrant and the controllers can be designed in lower degree. Hence, the HPSS can be achieved.

# 7.2 Chaos of a New Duffing-Van der Pol System with Legendre Function Parameters

This section introduces a new Duffing-Van der Pol system with Legendre function parameters.

$$\begin{cases} \frac{dx_1}{dt} = x_2 \\ \frac{dx_2}{dt} = -x_1 - x_1^3 - ax_2 + dx_3 \\ \frac{dx_3}{dt} = x_4 \\ \frac{dx_4}{dt} = -bx_3 + c\left(1 - x_3^2\right)x_4 + fx_1 \end{cases}$$
(7.9)

where *a*, *b*, *c*, *d*, *f* are parameters. We select the Legendre functions [48] as parameters of the system. The Legendre functions are defined by

$$P_n^m(x) = (-1)^m (1 - x^2)^{m/2} \frac{d^m}{dx^m} P_n(x)$$
 (7.10)

where  $P_n(x)$  is the Legendre polynomial of degree n.

$$P_{n}(x) = \frac{1}{2^{n} n!} \left[ \frac{d^{n}}{dx^{n}} (x^{2} - 1)^{n} \right]$$
Choosing  $n=2$ , we obtain
$$\begin{cases} L_{1}(x) = P_{2}^{0}(x) = P_{2}(x) \\ L_{2}(x) = P_{2}^{1}(x) = (-1)(1 - x^{2})^{\frac{1}{2}} \frac{d}{dx} P_{2}(x) \\ L_{3}(x) = P_{2}^{2}(x) = (-1)^{2}(1 - x^{2}) \frac{d^{2}}{dx^{2}} P_{2}(x) \end{cases}$$

$$P_{2}(x) = \frac{1}{2^{2} \cdot 2!} \left[ \frac{d^{2}}{dx^{2}} (x^{2} - 1)^{2} \right]$$

$$(7.11)$$

Changing the variable

$$x = \cos(t), \qquad -1 \le x \le 1 \tag{7.13}$$

We have three periodic functions of time  $L_1(x)$ ,  $L_2(x)$ , and  $L_3(x)$  as shown in Fig. 7.1. When the parameters of system are  $a=L_2+k$ ,  $b=L_3$ ,  $c=L_1+0.5$ ,  $d=L_1+L_2$ ,  $f=L_1$  and the initial states of system are  $x_1(0)=2$ ,  $x_2(0)=2.4$ ,  $x_3(0)=5$ ,  $x_4(0)=6$ , the bifurcation diagram by changing constant parameter k is shown in Fig.

7.2. Its corresponding Lyapunov exponents are shown in Fig. 7.3. The phase portraits, time histories, and Poincaré maps of the systems are showed in Fig. 7.4 and Fig. 7.5. When k=0.35, period 1 phenomena are shown in Fig. 7.4. When k=0, the chaotic behaviors are given in Fig.7.5.

#### 7.3 Numerical Results

Since the partner A is described by Eq. (7.9). The partner B is described as

$$\begin{cases} \frac{dy_1}{dt} = y_2 \\ \frac{dy_2}{dt} = -y_1 - y_1^3 - a_1 y_2 + d_1 y_3 \\ \frac{dy_3}{dt} = y_4 \\ \frac{dy_4}{dt} = -b_1 y_3 + c_1 (1 - y_3^2) y_4 + f_1 y_1 \end{cases}$$
(7.14)

where  $a_1$ ,  $b_1$ ,  $c_1$ ,  $d_1$  and  $f_1$  are constant parameters. For this scheme,  $u_1$ ,  $u_2$ ,  $u_3$  and  $u_4$  are added to the partner B then becomes controlled partner B:

$$\begin{cases}
\frac{dy_1}{dt} = y_2 + u_1 \\
\frac{dy_2}{dt} = -y_1 - y_1^3 - a_1 y_2 + d_1 y_3 + u_2
\end{cases}$$

$$\frac{dy_3}{dt} = y_4 + u_3 \\
\frac{dy_4}{dt} = -b_1 y_3 + c_1 (1 - y_3^2) y_4 + f_1 y_1 + u_4$$
(7.15)

In the HPSS, the error function is

$$e = Gxy - y + 100 (7.16)$$

The addition of the constant 100 makes the error dynamics always happens in the first quadrant. Our goal is y = Gxy + 100, i.e.

$$\lim_{t \to \infty} e = \lim_{t \to \infty} (Gxy - y + 100) = 0 \tag{7.17}$$

The error dynamics Eq. (7.8) becomes:

$$\begin{cases}
\dot{e}_{1} = h_{1} \left( x_{2} y_{1} + x_{1} y_{2} \right) - y_{2} + y_{1}^{2} - y_{1}^{2} - u_{1} \\
\dot{e}_{2} = h_{2} \left\{ y_{2} \left[ -x_{1} - x_{1}^{3} - a x_{2} + d x_{3} \right] + x_{2} \left[ -y_{1} - y_{1}^{3} - a_{1} y_{2} + d_{1} y_{3} \right] \right\} \\
+ y_{1} + y_{1}^{3} + a_{1} y_{2} - d_{1} y_{3} + y_{1}^{2} - y_{1}^{2} - u_{2} \\
\dot{e}_{3} = h_{3} \left( x_{4} y_{3} + x_{3} y_{4} \right) - y_{4} + y_{4}^{2} - y_{4}^{2} - u_{3} \\
\dot{e}_{4} = h_{4} \left\{ y_{4} \left[ -b x_{3} + c \left( 1 - x_{3}^{2} \right) x_{4} + f x_{1} \right] + x_{4} \left[ -b_{1} y_{3} + c_{1} \left( 1 - y_{3}^{2} \right) y_{4} + f_{1} y_{1} \right] \right\} \\
+ b_{1} y_{3} - c_{1} \left( 1 - y_{3}^{2} \right) y_{4} - f_{1} y_{1} + y_{4}^{2} - y_{4}^{2} - u_{4}
\end{cases} \tag{7.18}$$

Let initial states be  $(x_1, x_2, x_3, x_4) = (y_1, y_2, y_3, y_4) = (2, 2.4, 5, 6)$ , system parameters  $a = L_2$ ,  $b = L_3$ ,  $c = L_1 + 0.5$ ,  $d = L_1 + L_2$ ,  $f = L_1$ ,  $a_1 = 0.01$ ,  $b_1 = 1$ ,  $c_1 = 5$ ,  $d_1 = 0.67$ ,  $f_1 = 0.05$ , and the scaling matrix is G = diag(2, -1, 2, -1) we find the error dynamic always exists in first quadrant as shown in Fig. 7.6. By GYC partial region asymptotical stability theory, one can choose a Lyapunov function in the form of a positive definite function in first quadrant:

$$V = e_1 + e_2 + e_3 + e_4 (7.19)$$

Its time derivative is

$$\dot{V} = \dot{e}_{1} + \dot{e}_{2} + \dot{e}_{3} + \dot{e}_{4} 
= h_{1} \left( x_{2} y_{1} + x_{1} y_{2} \right) - y_{2} + y_{1}^{2} - y_{1}^{2} - u_{1} 
+ h_{2} \left\{ y_{2} \left[ -x_{1} - x_{1}^{3} - ax_{2} + dx_{3} \right] + x_{2} \left[ -y_{1} - y_{1}^{3} - a_{1} y_{2} + d_{1} y_{3} \right] \right\} 
+ y_{1} + y_{1}^{3} + a_{1} y_{2} - d_{1} y_{3} + y_{1}^{2} - y_{1}^{2} - u_{2} + h_{3} \left( x_{4} y_{3} + x_{3} y_{4} \right) - y_{4} + y_{4}^{2} - y_{4}^{2} - u_{3} \right\} 
+ h_{4} \left\{ y_{4} \left[ -bx_{3} + c \left( 1 - x_{3}^{2} \right) x_{4} + fx_{1} \right] + x_{4} \left[ -b_{1} y_{3} + c_{1} \left( 1 - y_{3}^{2} \right) y_{4} + f_{1} y_{1} \right] \right\} 
+ b_{1} y_{3} - c_{1} \left( 1 - y_{3}^{2} \right) y_{4} - f_{1} y_{1} + y_{4}^{2} - y_{4}^{2} - u_{4}$$
(7.20)

Choose

$$\begin{cases} u_{1} = h_{1}(x_{2}y_{1} + x_{1}y_{2}) - y_{2} + y_{1}^{2} + e_{1} \\ u_{2} = h_{2}\left\{y_{2}\left[-x_{1} - x_{1}^{3} - ax_{2} + dx_{3}\right] + x_{2}\left[-y_{1} - y_{1}^{3} - a_{1}y_{2} + d_{1}y_{3}\right]\right\} \\ + y_{1} + y_{1}^{3} + a_{1}y_{2} - d_{1}y_{3} - y_{1}^{2} + e_{2} \\ u_{3} = h_{3}(x_{4}y_{3} + x_{3}y_{4}) - y_{4} + y_{4}^{2} + e_{3} \\ u_{4} = h_{4}\left\{y_{4}\left[-bx_{3} + c\left(1 - x_{3}^{2}\right)x_{4} + fx_{1}\right] + x_{4}\left[-b_{1}y_{3} + c_{1}\left(1 - y_{3}^{2}\right)y_{4} + f_{1}y_{1}\right]\right\} \\ + b_{1}y_{3} - c_{1}\left(1 - y_{3}^{2}\right)y_{4} - f_{1}y_{1} - y_{4}^{2} + e_{4} \end{cases}$$

$$(7.21)$$

We obtain

$$\dot{V} = -e_1 - e_2 - e_3 - e_4 < 0 \tag{7.22}$$

which is negative definite function in the first quadrant. The numerical result is shown in Fig. 7.7.



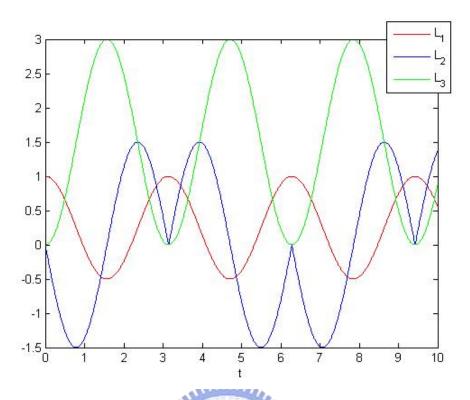


Fig. 7.1 Time histories of  $L_1$ ,  $L_2$ , and  $L_3$ .

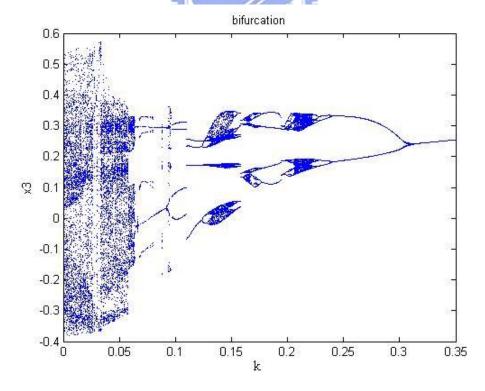


Fig. 7.2 The bifurcation diagram for new Duffing-Van der Pol system with Legendre function parameters.

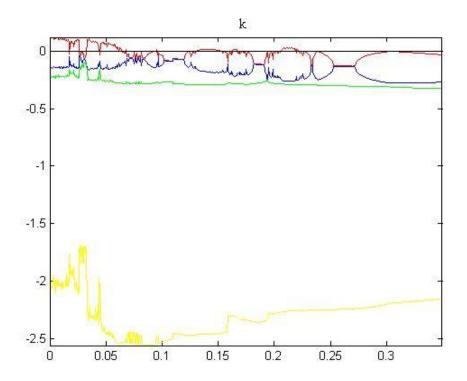


Fig. 7.3 The Lyapunov exponents for new Duffing-Van der Pol system with Legendre function parameters.

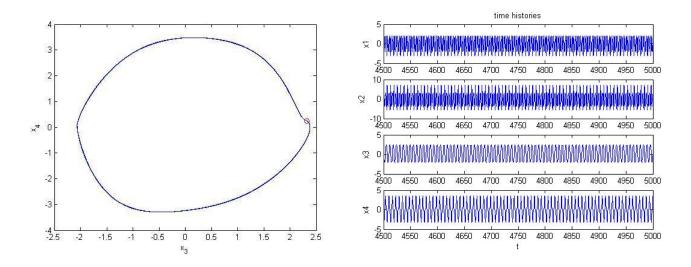


Fig. 7.4 Phase portrait, Poincaré maps, and time histories for new Duffing-Van der Pol system with Legendre function parameters when k=0.35 (period 1).

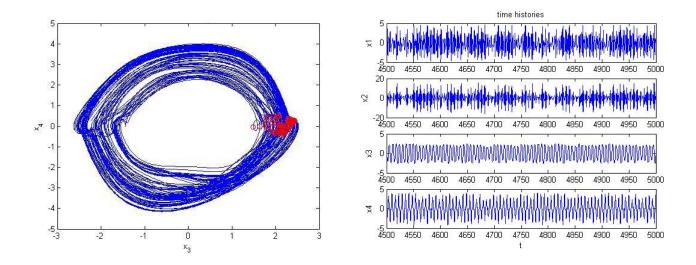


Fig. 7.5 Phase portrait, Poincaré maps, and time histories for new Duffing-Van der Pol system with Legendre function parameters when k=0 (chaos).

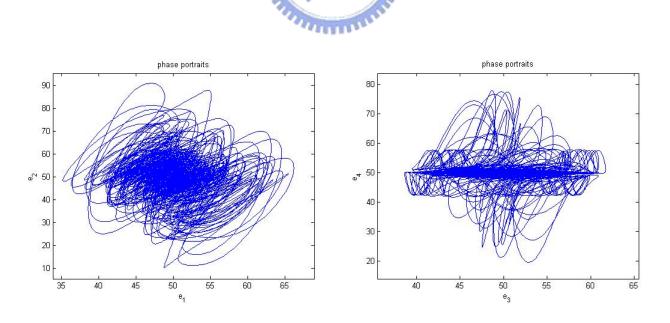


Fig. 7.6 Phase portraits of error dynamics.

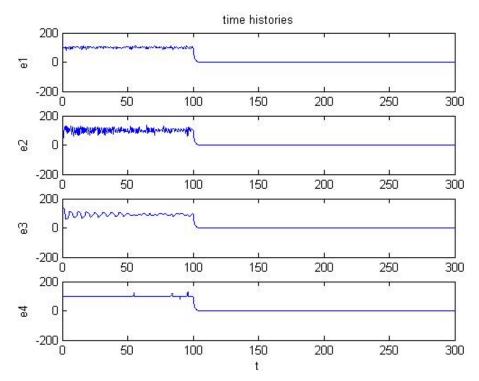


Fig. 7.7 Time histories of errors.

# Chapter 8

## **Conclusions**

In this thesis, the chaotic behavior in new Duffing-Van der Pol system is studied by phase portraits, time history, Poincaré maps, Lyapunov exponent, bifurcation diagrams, and parametric diagram.

Three kind of chaotic synchronization are presented. A new kind of generalized synchronization system in Chapter 3, *pragmatical hybrid projective chaotic generalized synchronization* (PHPCGS) of two chaotic systems with uncertain parameters, is obtained with the state variables of another hyperchaotic Mathieu-Duffing system as a constituent of the functional relation between master and slave. Based on the pragmatical asymptotical stability theorem, adaptive control law is used.

A new type for chaotic synchronization in Chapter 4, *pragmatical chaotic symplectic synchronization* (PCSS), is obtained with the state variables of another different order system as a constituent of the functional relation between "master" and "slave". Traditional generalized synchronizations are special cases of the symplectic synchronization. The pragmatical asymptotical stability theorem is used. New dynamic surface control is also used for making the controllers more simple.

A new chaos generalized synchronization method, using GYC partial region stability theory, is proposed in Chapter 5. Moreover, we also study the chaos control and anti-control by using the GYC partial region stability theory in Chapter 6. By using this theory, the controllers are of lower degree than that of controllers by using traditional Lyapunov asymptotical stability theorem. The simple linear homogeneous Lyapunov function of error states makes that the controllers are simpler and introduce

less simulation error. In addition, by replacing the parameters of the system with Legendre function, the chaos synchronization can be successfully obtained in Section 7.



## **Appendix I GYC Pragmatical Asymptotical Theorem [29]**

The stability for many problems in real dynamical systems is actual asymptotical stability, although may not be mathematical asymptotical stability. The mathematical asymptotical stability demands that trajectories from all initial states in the neighborhood of zero solution must approach the origin as  $t \to \infty$ . If there are only a small part or even one of initial states from which the trajectories or trajectory do not approach the origin as  $t \to \infty$ , the zero solution is not mathematically asymptotically stable. However, when the probability of occurrence of an event is zero, it means the event does not occur actually. If the probability of occurrence of the event that the trajectries from the initial states are that they do not approach zero when  $t \to \infty$ , is zero, the stability of zero solution is actual asymptotical stability though it is not mathematical asymptotical stability. In order to analyze the asymptotical stability of the equilibrium point of such systems, the pragmatical asymptotical stability theorem is used.

Let X and Y be two manifolds of dimensions m and n(m < n), respectively, and  $\varphi$  be a differentiable map from X to Y; then  $\varphi(X)$  is a subset of Lebesque measure 0 of Y [47]. For an autonomous system

$$\frac{dx}{dt} = f(x_1, x_2, \dots x_n) \tag{A.1}$$

where  $x = [x_1, x_2, \dots x_n]^T$ , the function  $f = [f_1, f_2, \dots f_n]^T$  is defined on  $D \subset \mathbb{R}^n$ . Let x = 0 be an equilibrium point for the system (A.1). Then

$$f(0) = 0 \tag{A.2}$$

For nonautonomous system,

$$\frac{dx}{dt} = f\left(x_1, x_2, \dots, x_{n+1}\right) \tag{A.3}$$

where  $t = x_{n+1} \subset R_+$ . The equilibrium point is

$$f(0, x_{n+1}) = 0 (A.4)$$

**Definition:** The equilibrium point for the dynamic system is pragmatically asymptotically stable provided that with initial points on C which is a subset of

Lebesque measure 0 of D, the behaviors of the corresponding trajectories cannot be determined, while with initial points on D-C, the corresponding trajectories behave as that agree with traditional asymptotical stability[29,30].

**Theorem:** Let  $V = [x_1, x_2, \dots x_n]^T : D \to R_+$  positive definite, analytic on D, where  $x_1, \dots, x_n$  are all space coordinates such that the derivative of V through differential equation,  $\dot{V}$ , is negative semi-definite.

Let *X* be the *m*-manifold consisting of point set for which  $\forall x \neq 0$ ,  $\dot{V}(x) = 0$  and *D* is an n-manifold. If m+1 < n, then the equilibrium point of the system is pragmatically asymptotically stable.

**Proof:** Since every point of X can be passed by a trajectory of Eq. (A.1), which is one dimensional, the collection of these trajectories, C, is a (m+1)-manifold [29,30]. If (m+1) < n, then the collection C is a subset of Lebesque measure 0 of D. By the above definition, the equilibrium point of the system is pragmatically asymptotically stable.  $\Box$ 

If an initial point is ergodicly chosen in D, the probability of that the initial point falls on the collection C is zero. Here, equal probability is assumed for every point chosen as an initial point in the neighborhood of the equilibrium point. Hence, the event that the initial point is chosen from collection C does not occur actually. Therefore, under the equal probability assumption, pragmatical asymptotical stability becomes actual asymptotical stability. When the initial point falls on D-C,  $\dot{V}(x) < 0$ , the corresponding trajectories behave as if they agree with traditional asymptotical stability because by the existence and uniqueness of the solution of initial-value problem, these trajectories never meet C.

For Eq. (9) Lyapunov function V is a positive definite function of n variables, i.e. p error state variables and n - p = m differences between unknown and estimated parameters, while  $\dot{V} = e^T C e$  is a negative semi-definite function of n variables. Since the number of error state variables is always more than one, p > 1, (m+1) < n is always satisfied. By pragmatical asymptotical stability theorem we have

$$\lim_{t \to \infty} e = 0 \tag{A.5}$$

and the estimated parameters approach the uncertain parameters. The parameteralized synchronizations is obtained. Therefore, the equilibrium point of the system is pragmatically asymptotically stable. Under the equal probability assumption, it is actually asymptotically stable for both error state variables and parameter variables.



### **Appendix II** GYC Partial Region Stability Theory [42]

Consider the differential equations of disturbed motion of a nonautonomous system in the normal form

$$\frac{dx_s}{dt} = X_s(t, x_1, \dots, x_n), \qquad (s = 1, \dots, n)$$
(A2.1)

where the function  $X_s$  is defined on the intersection of the partial region  $\Omega$  (shown in Fig. A2.1) and

$$\sum_{s} x_s^2 \le H \tag{A2.2}$$

and  $t > t_0$ , where  $t_0$  and H are certain positive constants.  $X_s$  which vanishes when the variables  $x_s$  are all zero, is a real valued function of t,  $x_1, \dots, x_n$ . It is assumed that  $X_s$  is smooth enough to ensure the existence, uniqueness of the solution of the initial value problem. When  $X_s$  does not contain t explicitly, the system is autonomous.

Obviously,  $x_s = 0$   $(s = 1, \dots n)$  is a solution of Eq.(A2.1). We are interested to the asymptotical stability of this zero solution on partial region  $\Omega$  (including the boundary) of the neighborhood of the origin which in general may consist of several subregions (Fig. A2.1).

#### **Definition 1:**

For any given number  $\varepsilon > 0$ , if there exists a  $\delta > 0$ , such that on the closed given partial region  $\Omega$  when

$$\sum_{s} x_{s0}^2 \le \delta, \quad (s = 1, \dots, n)$$
(A2.3)

for all  $t \ge t_0$ , the inequality

$$\sum_{s} x_s^2 < \varepsilon, \quad (s=1,\dots,n)$$
 (A2.4)

is satisfied for the solutions of Eq. (A2.1) on  $\Omega$ , then the disturbed motion  $x_s=0 \quad (s=1,\cdots n) \text{ is stable on the partial region } \Omega \, .$ 

#### **Definition 2:**

If the undisturbed motion is stable on the partial region  $\Omega$ , and there exists a  $\delta'>0$ , so that on the given partial region  $\Omega$  when

$$\sum_{s} x_{s0}^{2} \le \delta', \quad (s = 1, \dots, n)$$
 (A2.5)

The equality

$$\lim_{t \to \infty} \left( \sum_{s} x_s^2 \right) = 0 \tag{A2.6}$$

is satisfied for the solutions of Eq.(A2.1) on  $\Omega$ , then the undisturbed motion  $x_s = 0$   $(s = 1, \dots n)$  is asymptotically stable on the partial region  $\Omega$ .

The intersection of  $\Omega$  and region defined by Eq.(A2.5) is called the region of attraction.

### **Definition of Functions** $V(t, x_1, \dots, x_n)$ :

Let us consider the functions  $V(t,x_1,\cdots,x_n)$  given on the intersection  $\Omega_1$  of the partial region  $\Omega$  and the region

$$\sum_{s} x_s^2 \le h, \quad (s = 1, \dots, n)$$
 (A2.7)

for  $t \ge t_0 > 0$ , where  $t_0$  and h are positive constants. We suppose that the functions are single-valued and have continuous partial derivatives and become zero when  $x_1 = \cdots = x_n = 0$ .

#### **Definition 3:**

If there exists  $t_0 > 0$  and a sufficiently small h > 0, so that on partial region

 $\Omega_1$  and  $t \ge t_0$ ,  $V \ge 0$  (or  $\le 0$ ), then V is a positive (or negative) semidefinite, in general semidefinite, function on the  $\Omega_1$  and  $t \ge t_0$ .

#### **Definition 4:**

If there exists a positive (negative) definitive function  $W(x_1...x_n)$  on  $\Omega_1$ , so that on the partial region  $\Omega_1$  and  $t \ge t_0$ 

$$V - W \ge 0 \ (or - V - W \ge 0),$$
 (A2.8)

then  $V(t, x_1, ..., x_n)$  is a positive definite function on the partial region  $\Omega_1$  and  $t \ge t_0$ .

#### **Definition 5:**

If  $V(t,x_1,\ldots,x_n)$  is neither definite nor semidefinite on  $\Omega_1$  and  $t\geq t_0$ , then  $V(t,x_1,\ldots,x_n)$  is an indefinite function on partial region  $\Omega_1$  and  $t\geq t_0$ . That is, for any small h>0 and any large  $t_0>0$ ,  $V(t,x_1,\ldots,x_n)$  can take either positive or negative value on the partial region  $\Omega_1$  and  $t\geq t_0$ .

#### **Definition 6:** Bounded function *V*

If there exist  $t_0 > 0$ , h > 0, so that on the partial region  $\Omega_1$ , we have

$$|V(t,x_1,\ldots,x_n)| < L$$

where L is a positive constant, then V is said to be bounded on  $\Omega_1$ .

#### **Definition 7:** Function with infinitesimal upper bound

If V is bounded, and for any  $\lambda > 0$ , there exists  $\mu > 0$ , so that on  $\Omega_1$  when  $\sum_s x_s^2 \le \mu$ , and  $t \ge t_0$ , we have

$$|V(t, x_1, \dots, x_n)| \le \lambda$$

then V admits an infinitesimal upper bound on  $\Omega_1$ .

#### Theorem 1

If there can be found for the differential equations of the disturbed motion a definite function  $V(t, x_1, ..., x_n)$  on the partial region, and for which the derivative with respect to time based on these equations as given by the following:

$$\frac{dV}{dt} = \frac{\partial V}{\partial t} + \sum_{s=1}^{n} \frac{\partial V}{\partial x_s} X_s \tag{A2.9}$$

is a semidefinite function on the partial region whose sense is opposite to that of V, or if it becomes zero identically, then the undisturbed motion is stable on the partial region.

Proof:

Let us assume for the sake of definiteness that V is a positive definite function. Consequently, there exists a sufficiently large number  $t_0$  and a sufficiently small number h < H, such that on the intersection  $\Omega_1$  of partial region  $\Omega$  and

$$\sum_{s} x_s^2 \le h, \quad (s = 1, \dots, n)$$

and  $t \ge t_0$ , the following inequality is satisfied

$$V(t,x_1,\ldots,x_n)\geq W(x_1,\ldots,x_n),$$

where W is a certain positive definite function which does not depend on t. Besides that, Eq. (A2.9) may assume only negative or zero value in this region.

Let  $\varepsilon$  be an arbitrarily small positive number. We shall suppose that in any case  $\varepsilon < h$ . Let us consider the aggregation of all possible values of the quantities  $x_1, \ldots, x_n$ , which are on the intersection  $\omega_2$  of  $\Omega_1$  and

$$\sum_{s} x_s^2 = \varepsilon, \tag{A2.10}$$

and let us designate by l>0 the precise lower limit of the function W under this condition. by virtue of Eq. (A2.5), we shall have

$$V(t, x_1, ..., x_n) \ge l$$
 for  $(x_1, ..., x_n)$  on  $\omega_2$ . (A2.11)

We shall now consider the quantities  $x_s$  as functions of time which satisfy the differential equations of disturbed motion. We shall assume that the initial values  $x_{s0}$ of these functions for  $t = t_0$  lie on the intersection  $\Omega_2$  of  $\Omega_1$  and the region

$$\sum_{s} x_s^2 \le \delta,\tag{A2.12}$$

where  $\delta$  is so small that

$$V(t_0, x_{10}, ..., x_{n0}) < l$$
 (A2.13)

 $V(t_0,x_{10},...,x_{n0}) < l$  (A2.13) By virtue of the fact that  $V(t_0,0,...,0) = 0$ , such a selection of the number  $\delta$  is obviously possible. We shall suppose that in any case the number  $\delta$  is smaller than  $\varepsilon$ . Then the inequality

$$\sum_{s} x_s^2 < \varepsilon, \tag{A2.14}$$

being satisfied at the initial instant will be satisfied, in the very least, for a sufficiently small  $t-t_0$ , since the functions  $x_s(t)$  very continuously with time. We shall show that these inequalities will be satisfied for all values  $t > t_0$ . Indeed, if these inequalities were not satisfied at some time, there would have to exist such an instant t=T for which this inequality would become an equality. In other words, we would have

$$\sum_{s} x_s^2(T) = \varepsilon,$$

and consequently, on the basis of Eq. (A2.11)

$$V(T, x_1(T), ..., x_n(T)) \ge l$$
 (A2.15)

On the other hand, since  $\varepsilon < h$ , the inequality (Eq.(A2.4)) is satisfied in the entire interval of time [t<sub>0</sub>, T], and consequently, in this entire time interval  $\frac{dV}{dt} \le 0$ . This yields

$$V(T, x_1(T), ..., x_n(T)) \le V(t_0, x_{10}, ..., x_{n0}),$$

which contradicts Eq. (A2.14) on the basis of Eq. (A2.13). Thus, the inequality (Eq. (A2.4)) must be satisfied for all values of  $t > t_0$ , hence follows that the motion is stable.

Finally, we must point out that from the view-point of mathematics, the stability on partial region in general does not be related logically to the stability on whole region. If an undisturbed solution is stable on a partial region, it may be either stable or unstable on the whole region and vice versa. From the viewpoint of dynamics, we are not interesting to the solution starting from  $\Omega_2$  and going out of  $\Omega$ .

#### **Theorem 2**

If in satisfying the conditions of theorem 1, the derivative  $\frac{dV}{dt}$  is a definite function on the partial region with opposite sign to that of V and the function V itself permits an infinitesimal upper limit, then the undisturbed motion is asymptotically stable on the partial region.

#### Proof:

Let us suppose that V is a positive definite function on the partial region and that consequently,  $\frac{dV}{dt}$  is negative definite. Thus on the intersection  $\Omega_1$  of  $\Omega$  and the region defined by Eq. (A2.7) and  $t \ge t_0$  there will be satisfied not only the inequality (Eq. (A2.8)), but the following inequality as will:

$$\frac{dV}{dt} \le -W_1(x_1, \dots x_n),\tag{A2.16}$$

where  $W_1$  is a positive definite function on the partial region independent of t.

Let us consider the quantities  $x_s$  as functions of time which satisfy the differential equations of disturbed motion assuming that the initial values  $x_{s0} = x_s(t_0)$ of these quantities satisfy the inequalities (Eq. (A2.12)). Since the undisturbed motion is stable in any case, the magnitude  $\delta$  may be selected so small that for all values of  $t \ge t_0$  the quantities  $x_s$  remain within  $\Omega_1$ . Then, on the basis of Eq. (A2.16) the derivative of function  $V(t, x_1(t), ..., x_n(t))$  will be negative at all times and, consequently, this function will approach a certain limit, as t increases without limit, remaining larger than this limit at all times. We shall show that this limit is equal to some positive quantity different from zero. Then for all values of  $t \ge t_0$  the following be satisfied:  $V(t,x_1(t),...,x_n(t)) > \alpha$ inequality will be satisfied:

$$V(t, x_1(t), ..., x_n(t)) > \alpha$$
 (A2.17)

where  $\alpha > 0$ .

Since V permits an infinitesimal upper limit, it follows from this inequality that

$$\sum_{s} x_{s}^{2}(t) \ge \lambda, \quad (s = 1, ..., n), \tag{A2.18}$$

where  $\lambda$  is a certain sufficiently small positive number. Indeed, if such a number  $\lambda$ did not exist, that is , if the quantity  $\sum_{s} x_{s}(t)$  were smaller than any preassigned number no matter how small, then the magnitude  $V(t,x_1(t),...,x_n(t))$ , as follows from the definition of an infinitesimal upper limit, would also be arbitrarily small, which contradicts (A2.17).

If for all values of  $t \ge t_0$  the inequality (Eq. (A2.18)) is satisfied, then Eq.

(A2.16) shows that the following inequality will be satisfied at all times:

$$\frac{dV}{dt} \leq -l_1$$

where  $l_1$  is positive number different from zero which constitutes the precise lower limit of the function  $W_1(t,x_1(t),...,x_n(t))$  under condition (Eq. (A2.18)). Consequently, for all values of  $t \ge t_0$  we shall have:

$$V(t,x_1(t),...,x_n(t)) = V(t_0,x_{10},...,x_{n0}) + \int_{t_0}^{t} \frac{dV}{dt} dt \le V(t_0,x_{10},...,x_{n0}) - l_1(t-t_0),$$

which is, obviously, in contradiction with Eq.(A2.17). The contradiction thus obtained shows that the function  $V(t,x_1(t),...,x_n(t))$  approached zero as t increase without limit. Consequently, the same will be true for the function  $W(x_1(t),...,x_n(t))$  as well, from which it follows directly that

$$\lim_{t\to\infty} x_s(t) = 0, \quad (s =$$

which proves the theorem.

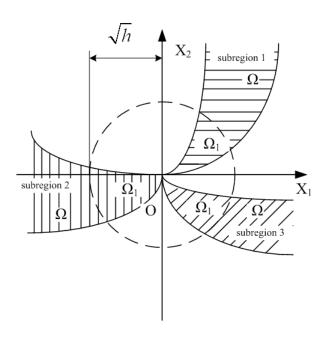


Fig. A2.1 Partial regions  $\Omega$  and  $\Omega_1$ .

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