

# Receding Horizon Chaos Synchronization Method

Choon Ki Ahn<sup>1</sup>, Chul Dong Lee<sup>2</sup>, and Moon Kyou Song<sup>3</sup>

<sup>1</sup> Department of Mechanical & Automotive Engineering, Seoul National University of Science & Technology, 172 Gongneung 2-dong, Nowon-gu, Seoul 139-743, Korea

<sup>2</sup> Jeonbuk Embedded System Research Center, Korea Electronics Technology Institute, 864-9 Dunsan-ri, Bongdong-eup, Wanju-gun, Jeonbuk 565-902, Korea

<sup>3</sup> Research Institute of Engineering Technology Development, Division of Electronics & Control Engineering, Wonkwang University, 344-2 Shinyong-dong, Iksan 570-749, Korea

Received: Oct. 26, 2011; Revised Mar. 19, 2012

Published online: 1 Sep. 2012

**Abstract:** This article proposes a new synchronization method, called a receding horizon synchronization (RHS) method, for a general class of chaotic systems. A new linear matrix inequality (LMI) condition on the finite terminal weighting matrix is proposed for chaotic systems under which non-increasing monotonicity of the optimal cost is guaranteed. It is shown that the proposed terminal inequality condition guarantees the closed-loop stability of the RHS method for chaotic systems. As an application of the proposed method, the RHS problem for Chua's chaotic system is investigated.

**Keywords:** receding horizon control (RHC); chaos synchronization; cost monotonicity; linear matrix inequality (LMI)

## 1. Introduction

Synchronization for chaotic dynamic systems has received much interest among scientists since a scheme to synchronize two identical nonlinear chaotic systems was introduced in [1]. It has been widely investigated in several fields including chemical, physical, and ecological systems [2]. In the literature, several synchronization schemes, such as OGY method [3], variable structure control [4], parameters adaptive control [5,6], observer-based control [7], active control [8,9], time-delay feedback approach [10], backstepping design technique [11,12], complete synchronization [13], have been applied to the chaos synchronization successfully.

Receding horizon control (RHC) scheme has been widely studied as an excellent feedback strategy [14–19]. RHC has made an important impact on industrial controls and is being increasingly applied in process controls. Various advantages are known for RHC, including the ability to handle time-varying and nonlinear systems, input/output constraint, uncertainty, and so on. The first method to guarantee the stability of the RHC is to impose an infinite terminal weighting. It is known that this method is equivalent to setting a zero terminal weighting matrix for the

inverse Riccati equation [14,15]. We call this method the terminal equality condition. Since the requirement for infinite terminal weighting is too demanding, studies of finite terminal weighting matrices have been made [16–19]. Although there are many advantages of RHC, to the best of our knowledge, the RHC based synchronization method for chaotic systems has not been established in the literature so far. This situation motivates our investigation.

In this paper, a new synchronization method based on the receding horizon control is proposed for chaotic systems. This method is called a receding horizon synchronization (RHS) method. First, we propose a new linear matrix inequality (LMI) condition on the finite terminal weighting matrix of the receding horizon cost function. Under this condition, non-increasing monotonicity of the optimal cost is shown to be guaranteed. Based on this LMI condition, we propose the RHS method for chaotic systems which guarantees the closed-loop asymptotic stability of the synchronization error system. We present a numerical example to illustrate the effectiveness of the proposed synchronization method.

This paper is organized as follows. In Section 2, we formulate the problem. In Section 3, an LMI condition for non-increasing monotonicity of the optimal cost is pro-

\* Corresponding author: e-mail: mksong@wku.ac.kr

posed. In Section 4, a new RHS method for chaotic systems is proposed. In Section 5, a numerical example is given, and finally, conclusions are presented in Section 6.

## 2. Problem Formulation

Consider a class of chaotic systems described by the following nonlinear differential equation:

$$\dot{x}(t) = Ax(t) + Bf(x(t)) \quad (1)$$

where  $x(t) \in R^n$  is the state vector,  $f(x(t)) \in R^n$  is a nonlinear function vector satisfying the global Lipschitz condition with Lipschitz constant  $L_f > 0$ ,  $A \in R^{n \times n}$  and  $B \in R^{n \times n}$  are known constant matrices. The system (1) is considered as a drive system. The synchronization problem of system (1) is considered using the drive-response configuration. According to the drive-response concept, the controlled response system is given by

$$\dot{z}(t) = Az(t) + Bf(z(t)) + Cu(t) \quad (2)$$

where  $z(t) \in R^n$  and  $u(t) \in R^m$  are the state vector and the control input of the controlled response system, respectively, and  $C \in R^{n \times m}$  is a known constant matrix. In fact, the matrix  $C$  is chosen arbitrarily. In this paper, we design a feedback control input  $u(t)$  via the RHC scheme. In order to design the feedback control input  $u(t)$ , we need information on states of drive and response systems. Thus, the control input  $u(t)$  in (2) depends on states of drive and response systems. Define the synchronization error  $e(t) = z(t) - x(t)$ . Then we obtain the synchronization error system

$$\dot{e}(t) = Ae(t) + B(f(z(t)) - f(x(t))) + Cu(t). \quad (3)$$

For the design of the RHS controller, the following finite horizon cost is associated with the synchronization error system (3):

$$J(e(t_0), t_0, t_1) = \int_{t_0}^{t_1} [e^T(t)Qe(t) + u^T(t)Ru(t)]dt + e^T(t_1)Q_f e(t_1), \quad (4)$$

where  $t_0 > 0$  is the initial time,  $t_1$  is the final time,  $Q > 0$ ,  $R > 0$ , and  $Q_f = Q_f^T > 0$ . The optimal control minimizing the cost function (4) and the corresponding optimal cost will be denoted by  $u^*(t)$ , ( $t_0 \leq t \leq t_1$ ), and  $J^*(e(t_0), t_0, t_1)$ , respectively. The RHS controller is then obtained by minimizing the cost function (4) with the initial time  $t_0$  and the terminal time  $t_1$  replaced by the current time  $t$  and the future time  $t+T$ , respectively, where  $T > 0$  is a constant. The stability of the proposed RHS controller depends on the choice of the terminal weighting matrix  $Q_f$ . In this paper, we show that the RHS controller with the cost function (4) guarantees the asymptotic stability under an LMI condition on the finite terminal weighting matrix  $Q_f$ .

## 3. Monotonicity of the Optimal Cost

In this section, we obtain a new LMI condition for the finite terminal weighting matrix  $Q_f$  under which the non-increasing cost monotonicity is guaranteed.

**Theorem 1.** Assume that there exist  $X = X^T > 0$  and  $Y \in R^{m \times n}$  such that

$$\begin{bmatrix} [1, 1] & X & Y^T & X & B \\ X & -Q^{-1} & 0 & 0 & 0 \\ Y & 0 & -R^{-1} & 0 & 0 \\ X & 0 & 0 & -\frac{1}{L_f^2}I & 0 \\ B^T & 0 & 0 & 0 & -I \end{bmatrix} \leq 0, \quad (5)$$

where  $[1, 1] = (AX + CY) + (AX + CY)^T$ . Then, the optimal cost  $J^*(e(\tau), \tau, \sigma)$  satisfies the following relation:

$$\frac{\partial J^*(e(\tau), \tau, \sigma)}{\partial \sigma} \leq 0, \quad \tau \leq \sigma. \quad (6)$$

Furthermore,  $Q_f$  is given by  $Q_f = X^{-1}$ .

*Proof:*  $\frac{\partial J^*(e(\tau), \tau, \sigma)}{\partial \sigma}$  satisfies the following relation:

$$\begin{aligned} & \frac{\partial J^*(e(\tau), \tau, \sigma)}{\partial \sigma} \\ &= \lim_{\Delta \rightarrow 0} \frac{1}{\Delta} \{J^*(e(\tau), \tau, \sigma + \Delta) - J^*(e(\tau), \tau, \sigma)\} \\ &= \lim_{\Delta \rightarrow 0} \frac{1}{\Delta} \left\{ \int_{\tau}^{\sigma} [e_1^T(t)Qe_1(t) + u_1^T(t)Ru_1(t)]dt \right. \\ & \quad + J^*(e_1(\sigma), \sigma, \sigma + \Delta) - \int_{\tau}^{\sigma} [e_2^T(t)Qe_2(t) \\ & \quad \left. + u_2^T(t)Ru_2(t)]dt - e_2^T(\sigma)Q_f e_2(\sigma) \right\}, \quad (7) \end{aligned}$$

where  $u_1(t)$  and  $u_2(t)$  are the optimal controls to minimize  $J(e(\tau), \tau, \sigma + \Delta)$  and  $J(e(\tau), \tau, \sigma)$ , respectively. If  $u_1(\cdot)$  is replaced by  $u_2(\cdot)$  up to  $\sigma$  and  $u_1(t) = Ke_2(t)$  for  $t \geq \sigma$ , then

$$\begin{aligned} & \frac{\partial J^*(e(\tau), \tau, \sigma)}{\partial \sigma} \\ & \leq \lim_{\Delta \rightarrow 0} \frac{1}{\Delta} \left\{ \int_{\tau}^{\sigma} [e_2^T(t)Qe_2(t) + u_2^T(t)Ru_2(t)]dt \right. \\ & \quad + J(e_2(\sigma), \sigma, \sigma + \Delta) \\ & \quad - \int_{\tau}^{\sigma} [e_2^T(t)Qe_2(t) + u_2^T(t)Ru_2(t)]dt - e_2^T(\sigma)Q_f e_2(\sigma) \left. \right\} \\ & = \lim_{\Delta \rightarrow 0} \frac{1}{\Delta} \left\{ \int_{\sigma}^{\sigma + \Delta} [e_2^T(t)Qe_2(t) + e_2^T(t)K^T RK e_2(t)]dt \right. \\ & \quad + e_2^T(\sigma + \Delta)Q_f e_2(\sigma + \Delta) - e_2^T(\sigma)Q_f e_2(\sigma) \left. \right\} \\ & = e_2^T(\sigma)Qx_2(\sigma) + e_2^T(\sigma)K^T RK e_2(\sigma) \\ & \quad + \frac{d}{d\sigma} \{e_2^T(\sigma)Q_f e_2(\sigma)\} \\ & = e_2^T(\sigma)Qx_2(\sigma) + e_2^T(\sigma)K^T RK e_2(\sigma) \\ & \quad + \dot{e}_2^T(\sigma)Q_f e_2(\sigma) + e_2^T(\sigma)Q_f \dot{e}_2(\sigma). \quad (8) \end{aligned}$$

By using (3), it can be shown that

$$\begin{aligned} & \frac{\partial J^*(e(\tau), \tau, \sigma)}{\partial \sigma} \\ & \leq e_2^T(\sigma) Q e_2(\sigma) + e_2^T(\sigma) K^T R K e_2(\sigma) \\ & + e_2^T(\sigma) Q_f [A e_2(\sigma) + B(f(z_2(\sigma)) - f(x_2(\sigma))) \\ & + C K e_2(\sigma)] + [A e_2(\sigma) + B(f(z_2(\sigma)) - f(x_2(\sigma))) \\ & + C K e_2(\sigma)]^T Q_f e_2(\sigma) \\ & = e_2^T(\sigma) Q e_2(\sigma) + e_2^T(\sigma) K^T R K e_2(\sigma) + e_2^T(\sigma) [Q_f A \\ & + Q_f C K + K^T C^T Q_f + A^T Q_f] e_2(\sigma) \\ & + e_2^T(\sigma) Q_f B(f(z_2(\sigma)) - f(x_2(\sigma))) \\ & + (f(z_2(\sigma)) - f(x_2(\sigma)))^T B^T Q_f e_2(\sigma). \end{aligned} \tag{9}$$

If we use the inequality  $X^T Y + Y^T X \leq X^T \Lambda X + Y^T \Lambda^{-1} Y$ , which is valid for any matrices  $X \in R^{n \times m}$ ,  $Y \in R^{n \times m}$ ,  $\Lambda = \Lambda^T > 0$ ,  $\Lambda \in R^{n \times n}$ , we have

$$\begin{aligned} & e_2^T(\sigma) Q_f B(f(z_2(\sigma)) - f(x_2(\sigma))) \\ & + (f(z_2(\sigma)) - f(x_2(\sigma)))^T B^T Q_f e_2(\sigma) \\ & \leq e_2^T(\sigma) Q_f B B^T Q_f e_2(\sigma) \\ & + (f(z_2(\sigma)) - f(x_2(\sigma)))^T (f(z_2(\sigma)) - f(x_2(\sigma))) \\ & \leq e_2^T(\sigma) [Q_f B B^T Q_f + L_f^2 I] e_2(\sigma). \end{aligned} \tag{10}$$

Using (10), we have

$$\begin{aligned} & \frac{\partial J^*(e(\tau), \tau, \sigma)}{\partial \sigma} \\ & \leq e_2^T(\sigma) [Q + K^T R K + Q_f A + Q_f C K + K^T C^T Q_f \\ & + A^T Q_f + Q_f B B^T Q_f + L_f^2 I] e_2(\sigma). \end{aligned} \tag{11}$$

If the following matrix inequality is satisfied:

$$\begin{aligned} & Q + K^T R K + Q_f A + Q_f C K + K^T C^T Q_f \\ & + A^T Q_f + Q_f B B^T Q_f + L_f^2 I \leq 0, \end{aligned} \tag{12}$$

it is clear that  $\frac{\partial J^*(e(\tau), \tau, \sigma)}{\partial \sigma} \leq 0$ . From Schur complement, the negative semi-definite of (12) is equivalent to

$$\begin{bmatrix} (1, 1) & I & K^T & I & Q_f B \\ I & -Q^{-1} & 0 & 0 & 0 \\ K & 0 & -R^{-1} & 0 & 0 \\ I & 0 & 0 & -\frac{1}{L_f^2} I & 0 \\ B^T Q_f & 0 & 0 & 0 & -I \end{bmatrix} \leq 0, \tag{13}$$

where  $(1, 1) = Q_f(A + CK) + (A + CK)^T Q_f$ . Pre- and post-multiplying (13) by  $diag(Q_f^{-1}, I, I, I, I)$  and introducing change of variables such as  $X = Q_f^{-1}$  and  $Y = K Q_f^{-1}$ , (13) is equivalently changed into the LMI (5). This completes the proof. ■

In the following theorem, it will be shown that the monotonicity of the optimal cost holds for all subsequent times if it holds once.

**Theorem 2.** If  $\frac{\partial J^*(e(\tau'), \tau', \sigma)}{\partial \sigma} \leq 0$  for some  $\tau'$ , then  $\frac{\partial J^*(e(\tau''), \tau'', \sigma)}{\partial \sigma} \leq 0$  where  $\tau' \leq \tau'' \leq \sigma$ .

*Proof:*  $\frac{\partial J^*(e(\tau'), \tau', \sigma)}{\partial \sigma}$  satisfies the following relation:

$$\begin{aligned} & \frac{\partial J^*(e(\tau'), \tau', \sigma)}{\partial \sigma} \\ & = \lim_{\Delta \rightarrow 0} \frac{1}{\Delta} \{ J^*(e(\tau'), \tau', \sigma + \Delta) - J^*(e(\tau'), \tau', \sigma) \} \\ & = \lim_{\Delta \rightarrow 0} \frac{1}{\Delta} \left\{ \int_{\tau'}^{\tau''} [e_1^T(t) Q e_1(t) + u_1^T(t) R u_1(t)] dt \right. \\ & + J^*(e_1(\tau''), \tau'', \sigma + \Delta) \\ & - \int_{\tau'}^{\tau''} [e_2^T(t) Q e_2(t) + u_2^T(t) R u_2(t)] dt \\ & \left. - J^*(e_2(\tau''), \tau'', \sigma) \right\}, \end{aligned} \tag{14}$$

where  $u_1(t)$  and  $u_2(t)$  are the optimal controls to minimize  $J(e(\tau'), \tau', \sigma + \Delta)$  and  $J(e(\tau'), \tau', \sigma)$ , respectively. If  $u_2(\cdot)$  is replaced by  $u_1(\cdot)$  up to  $\tau''$ , then we have

$$\begin{aligned} & \frac{\partial J^*(e(\tau'), \tau', \sigma)}{\partial \sigma} \\ & \geq \lim_{\Delta \rightarrow 0} \frac{1}{\Delta} \left\{ J^*(e_1(\tau''), \tau'', \sigma + \Delta) - J^*(e_1(\tau''), \tau'', \sigma) \right\} \\ & = \frac{\partial J^*(e_1(\tau''), \tau'', \sigma)}{\partial \sigma}. \end{aligned} \tag{15}$$

$\frac{\partial J^*(e(\tau'), \tau', \sigma)}{\partial \sigma} \leq 0$  implies  $\frac{\partial J^*(e(\tau''), \tau'', \sigma)}{\partial \sigma} \leq 0$ . This completes the proof. ■

#### 4. Closed-loop Stability of Receding Horizon Synchronization Controller

The RHS controller is obtained by replacing  $t_0$  and  $t_1$  by  $t$  and  $t+T$ , respectively, where  $T$  denotes the horizon length satisfying  $0 < T < \infty$ . The stability of the RHS controller is given in the following theorem:

**Theorem 3.** If  $\frac{\partial J^*(e(t), t, \sigma)}{\partial \sigma} \Big|_{\sigma=t+T} \leq 0$ , the synchronization error system (3) with the RHS controller is asymptotically stable.

*Proof:*  $J^*(e(t), t, t+T)$  is given by

$$\begin{aligned} & J^*(e(t), t, t+T) \\ & = \int_t^{t+\mu} [e^{*T}(t) Q e^*(t) + u^{*T}(t) R u^*(t)] dt \\ & + J^*(e(t+\mu), t+\mu, t+T). \end{aligned} \tag{16}$$

According to Theorem 2,  $\frac{\partial J^*(e(t), t, \sigma)}{\partial \sigma} \Big|_{\sigma=t+T} \leq 0$  implies

$\frac{\partial J^*(e(t+\mu), t+\mu, \sigma)}{\partial \sigma} \Big|_{\sigma=t+T} \leq 0$  for any  $0 < \mu < T$ . Hence,

we have

$$\begin{aligned} & J^*(e(t), t, t+T) \\ & \geq \int_t^{t+\mu} [e^{*T}(t)Qe^*(t) + u^{*T}(t)Ru^*(t)]dt \\ & + J^*(e(t+\mu), t+\mu, t+T+\mu), \end{aligned} \quad (17)$$

which means that  $J^*(e(t), t, t+T)$  is strictly decreasing. Therefore,  $J^*(e(t), t, t+T) \rightarrow c > 0$  as  $t \rightarrow \infty$ . Furthermore, from (17), it is clear that  $\int_t^{t+\mu} [e^{*T}(t)Qe^*(t) + u^{*T}(t)Ru^*(t)]dt \rightarrow 0$  as  $t \rightarrow \infty$ . Finally,  $e(t) \rightarrow 0$  and  $u(t) \rightarrow 0$  as  $t \rightarrow \infty$ . This completes the proof. ■

This result states that the non-increasing monotonicity of the optimal cost is a sufficient condition for the stability of the RHS controller. Based on Theorem 1, we obtain the following result on the stability of the RHS controller with the finite terminal weighting matrix  $Q_f$ .

**Corollary 1.** Assume that the finite terminal weighting matrix  $Q_f$  in (4) satisfies the LMI condition (5). Then, the synchronization error system (3) with the RHS controller is asymptotically stable.

*Proof:* The existence of  $Q_f$  satisfying the LMI condition (5) guarantees  $\left. \frac{\partial J^*(e(t), t, \sigma)}{\partial \sigma} \right|_{\sigma=t+T} \leq 0$ . Thus, the closed-loop stability follows from Theorem 3. This completes the proof. ■

## 5. Numerical Example

In this section, to verify and demonstrate the effectiveness of the proposed method, we discuss the simulation result for synchronizing Chua's chaotic system. Consider the following Chua's chaotic system:

$$\begin{aligned} \dot{x}_1(t) &= -10x_1(t) + 10x_2(t) + \left[ -0.69x_1(t) \right. \\ & \quad \left. - \frac{0.59}{2} (|x_1(t) + 1| - |x_1(t) - 1|) \right], \\ \dot{x}_2(t) &= x_1(t) - x_2(t) + x_3(t), \\ \dot{x}_3(t) &= -15x_2(t) - 0.0385x_3(t). \end{aligned} \quad (18)$$

The Chua's chaotic system (18) is rewritten as

$$\dot{x}(t) = Ax(t) + Bf(x(t)), \quad (19)$$

where

$$\begin{aligned} A &= \begin{bmatrix} -10.69 & 10 & 0 \\ 1 & -1 & 1 \\ 0 & -15 & -0.0385 \end{bmatrix}, \\ B &= \begin{bmatrix} -0.59 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \\ f(x(t)) &= \begin{bmatrix} \frac{1}{2} (|x_1(t) + 1| - |x_1(t) - 1|) \\ 0 \\ 0 \end{bmatrix}. \end{aligned}$$

For the numerical simulation, we use the following parameters:

$$\begin{aligned} L_f &= 1, \quad C = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \quad Q = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \\ R &= 1, \quad T = 0.5, \end{aligned}$$

where  $L_f = 1$  is obtained from the relation  $\|f(x(t))\| \leq \|x(t)\|$ . In this simulation, we use  $T = 0.5$ . If  $T$  is a big positive constant, the computational burden to obtain the RHS controller may increase very much. In this case, we need to use the high performance hardware for the implementation of the proposed RHS controller. Applying Theorem 1 to the Chua's chaotic system (19) yields

$$\begin{aligned} X &= \begin{bmatrix} 0.0576 & 0.0021 & 0.0323 \\ 0.0021 & 0.0235 & 0.0332 \\ 0.0323 & 0.0332 & 0.2377 \end{bmatrix}, \\ Y &= [0.0470 \quad -0.3578 \quad 0.0543]. \end{aligned}$$

In this section, in order to solve the LMI feasibility problem in Theorem 1, we utilized MATLAB LMI Control Toolbox [20], which implements state-of-the-art interior-point algorithms. Figure 1 shows state trajectories when the initial states are given by

$$\begin{bmatrix} x_1(0) \\ x_2(0) \\ x_3(0) \end{bmatrix} = \begin{bmatrix} 3.8 \\ 3.1 \\ 2.5 \end{bmatrix}, \quad \begin{bmatrix} z_1(0) \\ z_2(0) \\ z_3(0) \end{bmatrix} = \begin{bmatrix} 1.9 \\ 1.2 \\ -1.1 \end{bmatrix}.$$

From this figure, it can be seen that drive and response systems are indeed achieving chaos synchronization. Figure 2 shows that the proposed RHS method guarantees the asymptotic stability of the synchronization error system.

## 6. Conclusion

In this paper, we have proposed the RHS controller, which is a new synchronization controller, for chaotic systems. A new LMI condition on the finite terminal weighting matrix was proposed, which guaranteed the monotonicity of the optimal cost. Under this condition, it was shown that the asymptotic stability of the RHS method is guaranteed. Furthermore, the synchronization for the Chua's chaotic system was given to demonstrate the effectiveness of the proposed synchronization method.

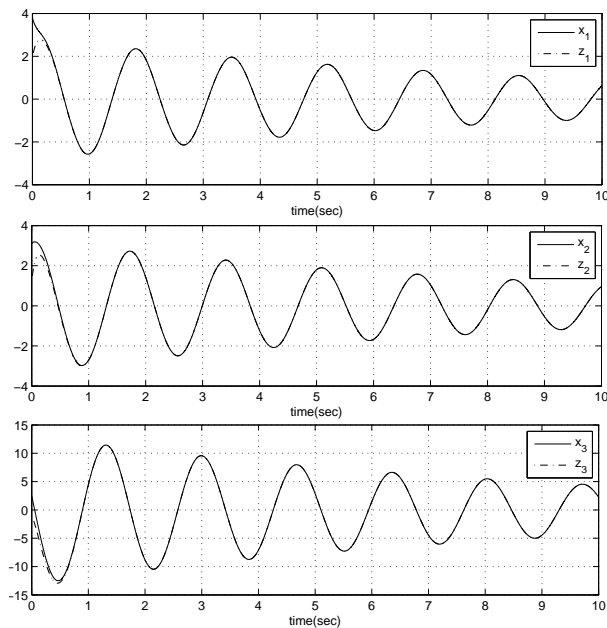


Figure 1 State trajectories

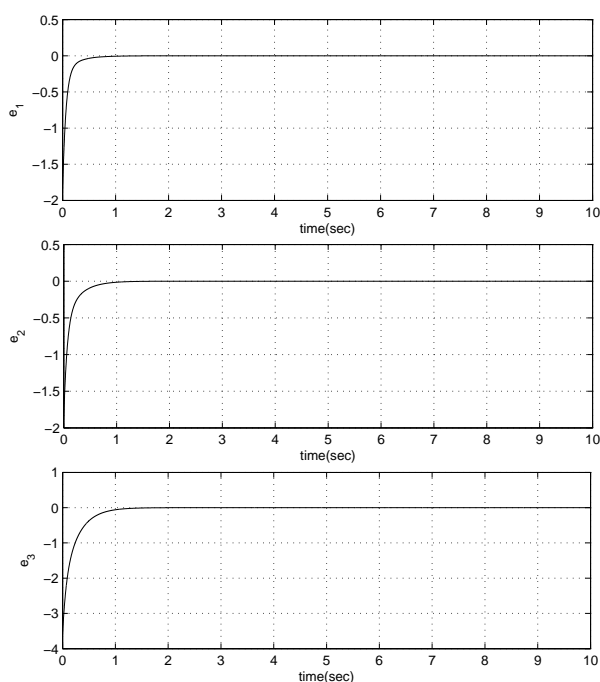


Figure 2 Synchronization errors

### Acknowledgments

This paper was supported by Wonkwang University in 2010.

### References

- [1] L. M. Pecora and T. L. Carroll. Synchronization in chaotic systems. *Phys. Rev. Lett.*, 64:821–824, 1990.
- [2] G. Chen and X. Dong. *From chaos to order*. Singapore: World Scientific, 1998.
- [3] E. Ott, C. Grebogi, and J.A. Yorke. Controlling chaos. *Phys. Rev. Lett.*, 64:1196–1199, 1990.
- [4] C.C Wang and J.P. Su. A new adaptive variable structure control for chaotic synchronization and secure communication. *Chaos, Solitons and Fractals*, 20:967–977, 2004.
- [5] J.H. Park. Adaptive synchronization of a unified chaotic systems with an uncertain parameter. *Int. J. Nonlinear Sci. Numer. Simul.*, 6:201–206, 2005.
- [6] Y. Wang, Z.H. Guan, and H.O Wang. Feedback an adaptive control for the synchronization of chen system via a single variable. *Phys. Lett. A*, 312:34–40, 2003.
- [7] X.S. Yang and G. Chen. Some observer-based criteria for discrete-time generalized chaos synchronization. *Chaos, Solitons and Fractals*, 13:1303–1308, 2002.
- [8] E. Bai and K. Lonngren. Synchronization of two Lorenz systems using active control. *Phys. Rev. E*, 8:51–58, 1997.
- [9] E.W. Bai and K.E. Lonngren. Sequential synchronization of two Lorenz systems using active control. *Chaos, Solitons and Fractals*, 11:1041–1044, 2000.
- [10] O.M. Kwon J.H. Park. LMI optimization approach to stabilization of time-delay chaotic systems. *Chaos, Solitons and Fractals*, 23:445–450, 2005.
- [11] X. Wu and J. Lu. Parameter identification and backstepping control of uncertain Lü system. *Chaos, Solitons and Fractals*, 18:721–729, 2003.
- [12] J. Hu, S. Chen, and L. Chen. Adaptive control for anti-synchronization of Chua’s chaotic system. *Phys. Lett. A*, 339:455–460, 2005.
- [13] M. Zhan, X. Wang, X. Gong, G. Wei, and C. Lai. Complete synchronization and generalized synchronization of one-way coupled time-delay systems. *Phys. Rev. E*, 68:6208–6213, 2003.
- [14] W. H. Kwon and A. E. Pearson. A modified quadratic cost problem and feedback stabilization of a linear system. *IEEE Trans. Automat. Contr.*, 22:838–842, 1977.
- [15] W. H. Kwon and A. E. Pearson. On feedback stabilization of time-varying discrete linear system. *IEEE Trans. Automat. Contr.*, 23:479–481, 1978.
- [16] W. H. Kwon, A. M. Bruckstein, and T. Kailath. Stabilizing state-feedback design via the moving horizon method. *International Journal of Control*, 37:631–643, 1983.
- [17] W. H. Kwon and D. G. Byun. Receding horizon tracking control as a predictive control and its stability properties. *International Journal of Control*, 50(5):1807–1824, 1989.
- [18] J. W. Lee, W. H. Kwon, and J. H. Choi. On stability of constrained receding horizon control with finite terminal weighting matrix. *Automatica*, 34(12):1607–1612, 1998.
- [19] W. H. Kwon and K. B. Kim. On Stabilizing Receding Horizon Controls for Linear Continuous Time-Invariant Systems. *IEEE Trans. Automat. Contr.*, 45(7):1329–1334, 2000.
- [20] P. Gahinet, A. Nemirovski, A. J. Laub, and M. Chilali. *LMI Control Toolbox*. The Mathworks Inc., Natick, 1995.



**Prof. Choon Ki Ahn** received the B.S. & M.S. degrees in School of Electrical Engineering from Korea University, Seoul, Korea. He received the Ph.D. degree in School of Electrical Engineering & Computer Science from Seoul National University, Seoul, Korea, in 2006. From 2006 to 2007, he has been a

Senior Research Engineer, Samsung Electronics, Korea. He is now on the faculty of the Department of Mechanical & Automotive Engineering, Seoul National University of Science & Technology, Seoul, Korea. He was the recipient of the Excellent Research Achievement Award of Wonkwang Univ. in 2010. He was awarded the Medal for 'Top 100 Engineers' 2010 by International Biographical Centre, Cambridge, England. He is a Senior Member of the IEEE. He published the vast majority of 88 journal papers as the sole author in reputed Science Citation Index (SCI) journals over recent 5 years. He is recognized as one of the world's most active researchers in fields of intelligent systems and nonlinear dynamics.



**Prof. Moon Kyou Song** received the B.S., M.S. and Ph.D. degrees in Electronics engineering from Korea University, Seoul, Korea, in 1998, 1990, and 1994, respectively. In 1994, he joined the faculty of Wonkwang University, Korea, where he is a Professor in the Division of Electronics and Control Engineer-

ing. He was an Invited Researcher with the Electronic Telecommunications Research Institute (ETRI), Daejeon, Korea, from 1997 to 1998 and 2000 to 2001. He was a Visiting professor with University of Victoria, BC, Canada, during 1999-2000 and Stanford University, CA, USA, during 2006-2007. His research interests include spread spectrum, error control coding and wireless communications.



**Dr. Chul-Dong Lee** received the M.S. degree in electrical and electronic engineering from Hanyang University, Seoul, Korea, and the Ph.D. degree from Chungbuk National University, Chungju, Korea. From 1977 to 1994, he was a Senior Researcher with Electronic Technology Research Institute. He is currently the di-

rector of Jeonbuk Embedded System Research Center, Korea Electronics Technology Institute, Jeonbuk, Korea.