

# Adaptive Actuator Failure Compensation Control of Second-order Nonlinear Systems with Unknown Time Delay

Jianping Cai<sup>†</sup>, Haoyi Que<sup>‡</sup>, Dehui Lin<sup>§</sup>, Qingping Zhou<sup>♯</sup>, Lujuan Shen<sup>†</sup>

**Abstract**—Unknown actuator failures and unknown time delays seem inevitable in practice systems. So it is meaningful to investigate the control problem for time delay systems with unknown actuator failures. In this paper, a robust adaptive control law is proposed based on backstepping technique for such a problem. Unlike existing results, the bounding function of time delay term  $h_i(x_1, x_2)$  in system (1) is allowed to depend on all system states which leads to triangular structure being destroyed. By maintaining a linear relation between the state vector  $(x_1, x_2)$  and its transformed vector  $(z_1, z_2)$  in controller design, such difficulty has been successfully solved. Namely, triangular structure requirement of backstepping technology is no longer needed in this paper. It is shown that the designed controller can ensure the stability of the closed-loop system.

**Index Terms**—Adaptive control, Time delay, Actuator failure, Backstepping.

## I. INTRODUCTION

**T**IME delay exists in wide range of practical systems, especially in mechanics, physics, biology and engineering systems and so on. Because the effect of time delays may lead to instability or even catastrophic accidents, so it cannot be ignored in the design and analysis of control systems. In the context of time delay systems, several schemes have been proposed in recent twenty years, see for example [1]-[7]. For neural network systems including coupled neural networks [1], uncertain neural systems [2], Markovian jump neural networks [3], discrete-time neural networks [4]-[5], several results have been achieved to compensate the effect of time delays. For uncertain nonlinear systems, especially parametric strict feedback systems, the existence of unknown time delay make the controller design and stability analysis become difficult. The triangular time delay term is considered in [6]-[7] and the control laws have been proposed by using backstepping technology.

On the other hand, some common phenomena such as actuator failures [8]-[10], fading [11], uncertain kinematics

and dynamics to robotic systems [12] and electronic circuits systems [13], dead-zone [14]-[17], input and sensor nonlinearities [18]-[20], backlash [21], state constraint [22] of practical systems have great effect on the performance of the controlled system. Actuator failures are inevitable in all practical systems and often uncertain in time, value and pattern. Therefore, the uncertainties caused by unknown failures are difficult to compensate. Especially, the existence of time delays renders such problem much more complex. In this paper, we address such a problem by considering controlling a second-order nonlinear systems with time delays. Different from [6]-[7], Time delays considered in this paper do not meet triangular structure condition. Therefore, standard backstepping technology [23] can not be used in the controller design. In order to solve it, the effects of such time delays are not considered in every step. Instead, their effects will be accumulated to the last step and compensated by choosing appropriate design parameters in control laws. At same time, the existence of unknown actuator failures make the controller design and stability analysis become more difficult. The compositive effects caused by time delays and unknown failures must be compensated by the same controller. To overcome it, the online estimation of unknown actuator failures for time delay systems has been designed. Throughout the whole process of controller design, it is important to maintain a linear relation between the state vector  $(x_1, x_2)$  and its transformed vector  $(z_1, z_2)$ .

The main contributions of this paper, compared with the existing results, are as follows: (1) The control problem is investigated for time delay systems with unknown actuator failures and unknown parameters. (2) In addition, the online estimation of unknown actuator failures for time delay systems has been designed to compensate the compositive effects caused by unknown time delays and unknown actuator failures. (3) Moreover, unlike existing results of time delay systems by using backstepping technology triangular structure requirement is no longer needed in this paper. Namely, the bounding function of time delay term  $h_i(x_1, x_2)$  in system (1) is allowed to depend on all system states and then triangular structure is destroyed.

The rest of the paper is organized as follows. In section 2, we formulate the second order time delay system by differential equations and give the model of unknown actuator failures. Adaptive control scheme and stability analyses of the closed-loop system are given in section 3. In section 4, some simulation results are shown in detail. Finally, the paper is concluded in section 5.

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<sup>†</sup>Jianping Cai is with Zhejiang University of Water Resources and Electric Power, Hangzhou, 310018, P.R.China. e-mail: caijianping2001@163.com; Lujuan Shen is with Zhejiang University of Water Resources and Electric Power, Hangzhou, 310018, P.R.China. e-mail: ljshen@zjweu.edu.cn.

<sup>‡</sup>Haoyi Que is with the State Key Laboratory of Industrial Control Technology Institute of Cyber-Systems and Control, Zhejiang University, Hangzhou, 310027, P.R.China. e-mail: 763898066@qq.com.

<sup>§</sup>Dehui Lin is with China Jiliang University, Hangzhou, 310018, P.R.China. e-mail: woihll@qq.com.

<sup>♯</sup>Qingping Zhou is with Tangshan Normal University, Tangshan, 063000, P.R.China. e-mail: zhouqingping1999@163.com.

## II. MODELS AND PROBLEM STATEMENT

For illustrating our design ideas, the following class of second-order nonlinear systems with uncertain parameters and unknown time delay is considered. The system model is given as

$$\begin{aligned}\dot{x}_1 &= x_2 + h_1(x_1(t-\tau), x_2(t-\tau)) \\ \dot{x}_2 &= \theta^T f(x) + \sum_{i=1}^m b_i u_i + h_2(x_1(t-\tau), x_2(t-\tau)) \\ y &= x_1\end{aligned}\quad (1)$$

where  $x = (x_1, x_2)^T$  is system state and  $u_i \in R$  ( $i = 1, 2, \dots, m$ ) is input,  $y = x_1$  is output.  $f(x) \in R^p$  is known function and  $b_i \in R$  ( $i = 1, 2, \dots, m$ ),  $\theta \in R^p$  are unknown parameters.  $h_1(\cdot), h_2(\cdot) \in R$  are known smooth functions and  $\tau \in R$  is unknown time delay of state  $x$ .

We now consider the  $i$ th actuator may fail during its operation. As in [8] and [9], the failure of the  $i$ th actuator at time instant  $t_{if}$  can be modelled as follows

$$\begin{aligned}u_i &= \rho_i u_{ci} + u_{ki}, \quad (\forall t \geq t_{if}) \\ \rho_i u_{ki} &= 0\end{aligned}\quad (2)$$

where  $u_{ci}$  is the input of the  $i$ th actuator and  $\rho_i \in [0, 1]$ ,  $u_{ki}$  and  $t_{if}$  are unknown constant. The following three cases are discussed in detail

- $\rho_i = 1$ ,  
It indicates  $u_i = u_{ci}$ . The  $i$ th actuator works normally.
- $0 < \rho_i < 1$ ,  
It indicates  $u_i = \rho_i u_{ci}$ . The  $i$ th actuator is called partial loss of effectiveness.
- $\rho_i = 0$ ,  
It indicates  $u_i = u_{ki}$ . The  $i$ th actuator is called total loss of effectiveness.

The control objective is to design adaptive control scheme for the system (1) with unknown actuator failures given in (2) to guarantee all signals bounded. To design adaptive control scheme, the following Assumptions for system functions, unknown time delay and actuator failures are made.

**Assumption 1:** Nonlinear functions  $h_i(x_1, x_2)$   $i = 1, 2$  such that

$$|h_i(x_1, x_2)| \leq \delta_i \|(x_1, x_2)\|_2 \quad (i = 1, 2) \quad (3)$$

where  $\delta_i > 0$  is a known constant.

**Remark 1:** In the context of adaptive control of strict-feedback systems with state delays by using backstepping techniques, the time delay term in the  $i$ th state differential equation is required to only depend on states  $x_1, \dots, x_i$ , for examples [6] and [7]. Such a requirement can keep the control system has triangular structure which is needed by backstepping. However according to Assumption 1, the bounding function of time delay term  $h_i(x_1, x_2)$  in system (1) can depend all system states. Thus it means that such triangular structure requirement is no longer needed in this paper.

**Remark 2:**  $\delta_i$  can be interpreted as the gain or strength of the time delay term. In this paper, we will discuss how to obtain the adaptive controller of system (1). The existence of time delay and unknown actuator failures given in (2)

render the controller design much more complex and difficult, therefore the result given in this paper will be aimed at enough small  $\delta_i$ . It makes our results limited.

**Assumption 2:** The number of total fail actuators is up to  $m - 1$  and the control objectives can be achieved by other normal actuators. Also any actuator can change only from normal to partial failure or total failure once.

**Remark 3:** Assumption 2 is a basic assumption required in adaptive failure compensation. It can be concluded that total number of actuator failures is finite and there exists a finite time instant  $T_f$  after which no new failure will occur.

**Assumption 3:** Unknown parameter  $b_i \neq 0$  and the sign of  $b_i$  is known.

With actuator failure model given in (2), system (1) can be rewritten as

$$\begin{aligned}\dot{x}_1 &= x_2 + h_1(x_1(t-\tau), x_2(t-\tau)) \\ \dot{x}_2 &= \theta^T f(x) + \sum_{i=1}^m b_i (\rho_i u_{ci} + u_{ki}) \\ &\quad + h_2(x_1(t-\tau), x_2(t-\tau)) \\ y &= x_1\end{aligned}\quad (4)$$

## III. DESIGN AND ANALYSIS OF ADAPTIVE CONTROLLERS

To obtain suitable control law and update laws based on the backstepping approach, the following change of coordinates are introduced.

$$\begin{aligned}z_1 &= y = x_1 \\ z_2 &= x_2 - \alpha_1\end{aligned}\quad (5)$$

where  $\alpha_1$  is the virtual control in step 1. Below we will give the design details following the recursive backstepping procedure.

Step 1: From (1) and (5) the derivative of  $z_1$  can be rewritten as

$$\begin{aligned}\dot{z}_1 &= \dot{x}_1 \\ &= x_2 + h_1(x_1(t-\tau), x_2(t-\tau)) \\ &= z_2 + \alpha_1 + h_1(x_1(t-\tau), x_2(t-\tau))\end{aligned}\quad (6)$$

We define a positive definite Lyapunov function as follows

$$\bar{V}_1 = \frac{1}{2} z_1^2 + \frac{e_1}{2} \int_{t-\tau}^t U_1(x_1(s), x_2(s)) ds \quad (7)$$

where  $e_1 > 0$  is a design parameter. Smooth function  $U_1(x_1(s), x_2(s))$  is chosen as

$$U_1(x_1(s), x_2(s)) = h_1^2(x_1(s), x_2(s)) \quad (8)$$

Then we can get the derivative of  $\bar{V}_1$  is

$$\begin{aligned}\dot{\bar{V}}_1 &= z_1 \dot{z}_1 + \frac{e_1}{2} U_1(x_1(s), x_2(s)) \Big|_{t-\tau}^t \\ &= z_1 (z_2 + \alpha_1 + h_1(x_1(t-\tau), x_2(t-\tau))) \\ &\quad + \frac{e_1}{2} (h_1^2(x_1(t), x_2(t)) - h_1^2(x_1(t-\tau), x_2(t-\tau)))\end{aligned}\quad (9)$$

Note that

$$z_1 h_1(x_1(t-\tau), x_2(t-\tau)) \leq \frac{z_1^2}{2e_1} + \frac{e_1}{2} h_1^2(x_1(t-\tau), x_2(t-\tau)) \quad (10)$$

So we have

$$\dot{V}_1 \leq z_1(z_2 + \alpha_1) + \frac{z_1^2}{2e_1} + \frac{e_1}{2}h_1^2(x_1(t), x_2(t)) \quad (11)$$

Virtual control  $a_1$  can be chosen as

$$\alpha_1 = -k_1 z_1 - \frac{1}{2e_1} z_1 \quad (12)$$

where  $k_1$  is a positive design parameter. Then from (11) and (12) we have

$$\begin{aligned} \dot{V}_1 &\leq z_1(z_2 - k_1 z_1 - \frac{1}{2e_1} z_1) + \frac{z_1^2}{2e_1} + \frac{e_1}{2}h_1^2(x_1(t), x_2(t)) \\ &= z_1 z_2 - k_1 z_1^2 + \frac{e_1}{2}h_1^2(x_1(t), x_2(t)) \end{aligned} \quad (13)$$

Step 2: From (1) and (5), we have

$$\begin{aligned} \dot{z}_2 &= \dot{x}_2 - \dot{\alpha}_1 \\ &= \theta^T f(x) + \sum_{i=1}^m b_i u_i + h_2(x_1(t-\tau), x_2(t-\tau)) - \dot{\alpha}_1 \end{aligned} \quad (14)$$

Note that

$$\begin{aligned} \dot{\alpha}_1 &= \frac{d\alpha_1}{dx_1} \dot{x}_1 \\ &= -(k_1 + \frac{1}{2e_1})(x_2 + h_1(x_1(t-\tau), x_2(t-\tau))) \end{aligned} \quad (15)$$

We have

$$\begin{aligned} \dot{z}_2 &= \theta^T f(x) + \sum_{i=1}^m b_i u_i + h_2(x_1(t-\tau), x_2(t-\tau)) \\ &\quad + (k_1 + \frac{1}{2e_1})(x_2 + h_1(x_1(t-\tau), x_2(t-\tau))) \end{aligned} \quad (16)$$

From (1) and (5), we have

$$\begin{aligned} z_2 \dot{z}_2 &= z_2(\theta^T f(x) + \sum_{i=1}^m b_i u_i) \\ &\quad + z_2 h_2(x_1(t-\tau), x_2(t-\tau)) \\ &\quad + (k_1 + \frac{1}{2e_1}) z_2 (x_2 + h_1(x_1(t-\tau), x_2(t-\tau))) \end{aligned} \quad (17)$$

Note that

$$\begin{aligned} &z_2 h_2(x_1(t-\tau), x_2(t-\tau)) \\ &\leq \frac{z_2^2}{2e_{22}} + \frac{e_{22}}{2} h_2^2(x_1(t-\tau), x_2(t-\tau)) \\ &z_2 h_1(x_1(t-\tau), x_2(t-\tau)) \\ &\leq \frac{z_2^2}{2e_{21}} + \frac{e_{21}}{2} h_1^2(x_1(t-\tau), x_2(t-\tau)) \end{aligned} \quad (18)$$

where  $e_{21}$  and  $e_{22}$  are positive design parameters. Then we have

$$\begin{aligned} z_2 \dot{z}_2 &\leq z_2(\theta^T f(x) + \sum_{i=1}^m b_i u_i) + \frac{z_2^2}{2e_{22}} + \frac{z_2^2}{2e_{21}} \\ &\quad + (k_1 + \frac{1}{2e_1}) z_2 x_2 + \frac{e_{22}}{2} h_2^2(x_1(t-\tau), x_2(t-\tau)) \\ &\quad + \frac{e_{21}}{2} h_1^2(x_1(t-\tau), x_2(t-\tau)) \end{aligned} \quad (19)$$

Different from the standard backstepping approach, a virtual control  $\alpha$  in this step is designed as follows

$$\begin{aligned} \alpha &= -z_1 - k_2 z_2 - \hat{\theta}^T f(x) - (k_1 + \frac{1}{2e_1})(\frac{1}{2e_{21}} z_2 + x_2) \\ &\quad - \frac{1}{2e_{22}} z_2 \end{aligned} \quad (20)$$

where  $k_2$  is a positive design parameter and  $\hat{\theta}$  is the estimation of  $\theta$ . Similar to [8] and [9], the control law and update laws are designed as follows

**Control Law:**

$$u_{ci} = \text{sign}(b_i) \hat{\varrho}^T \gamma \quad (21)$$

where  $\hat{\varrho}$  is the estimation of  $\varrho$ .  $\varrho$  is an unknown parameter and  $\gamma$  is a known function. Both are  $m+1$  dimensional vectors and can be denoted as

$$\varrho = (\varrho_1, \varrho_{21}, \dots, \varrho_{2m})^T, \gamma = (\gamma_1, \gamma_{21}, \dots, \gamma_{2m})^T \quad (22)$$

**Update Laws:**

$$\begin{aligned} \dot{\hat{\theta}} &= \Gamma_\theta f(x) z_2 \\ \dot{\hat{\varrho}} &= -\Gamma_\varrho \gamma z_2 \end{aligned} \quad (23)$$

where  $\Gamma_\theta, \Gamma_\varrho$  are positive definite matrices.

#### IV. STABILITY ANALYSIS

We now analyze the stability of closed loop system with control law and update laws in (21) and (23). We use  $T_j$  to represent the time instant at which the actuator failure occurred. Therefore, all actuators working condition do not change between two adjacent time point  $T_i$  and  $T_{i+1}$ . Namely, no new normal actuator fails in time interval  $(T_i, T_{i+1})$ . Let the set  $U_{iT}$  denotes the actuators of total failure in interval  $(T_i, T_{i+1})$ . Clearly, with Assumption 2 there exists a time instant  $T_f$  such that no new failure will occur after it. Let  $T_0 = 0, T_{f+1} = \infty$ , so we can get  $U_{0T} = \{1, 2, \dots, m\}$  and  $U_{fT}$  is not empty. Our main result is given as follows

**Theorem 1:** Consider the time delay system (1) with unknown actuator failures described by (2) with control law (21) and the update laws (23). Under Assumption 1 to Assumption 3, there exist a positive constant  $\delta^*$  such that for all  $\delta_i < \delta^*$  All the signals in the closed-loop system are globally bounded. Also output  $y$  satisfies

$$\lim_{t \rightarrow \infty} |y| = 0 \quad (24)$$

**Proof.** Firstly, we consider the following Lyapunov function in time interval  $[T_i, T_{i+1})$

$$\begin{aligned} V_i &= \bar{V}_1 + \frac{1}{2} z_2^2 + \frac{1}{2} \tilde{\theta}^T \Gamma_\theta^{-1} \tilde{\theta} + \sum_{i \in \bar{U}_{it}} \frac{\rho_i |b_i|}{2} \tilde{\varrho}^T \Gamma_\varrho^{-1} \tilde{\varrho} \\ &\quad + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1}) \int_{t-\tau}^t U_{21}(x_1(s), x_2(s)) ds \\ &\quad + \frac{e_{22}}{2} \int_{t-\tau}^t U_{22}(x_1(s), x_2(s)) ds \end{aligned} \quad (25)$$

where  $\tilde{\varrho} = \hat{\varrho} - \varrho$  and  $\tilde{\theta} = \hat{\theta} - \theta$  are estimation errors of  $\varrho$  and  $\theta$ . Let  $U_{21}(x_1(s), x_2(s)) = h_1^2(x_1(s), x_2(s))$  and

$U_{22}(x_1(s), x_2(s)) = h_2^2(x_1(s), x_2(s))$ . Then the derivative of  $V_i$  is

$$\begin{aligned}\dot{V}_i &= \dot{V}_1 + z_2 \dot{z}_2 + \tilde{\theta}^T \Gamma_{\theta}^{-1} \dot{\tilde{\theta}} + \sum_{i \in \bar{U}_{it}} \rho_i |b_i| \tilde{\varrho}^T \Gamma_{\varrho}^{-1} \dot{\tilde{\varrho}} \\ &\quad + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1}) (h_1^2(t) - h_1^2(t - \tau)) \\ &\quad + \frac{e_{22}}{2} (h_2^2(t) - h_2^2(t - \tau))\end{aligned}\quad (26)$$

where we will use  $h_i^2(t)$  and  $h_i^2(t - \tau)$  represent  $h_i^2(x_1(t), x_2(t))$  and  $h_i^2(x_1(t - \tau), x_2(t - \tau))$ , respectively. With (13) and (19) we have

$$\begin{aligned}\dot{V}_i &\leq z_1 z_2 - k_1 z_1^2 + \frac{e_1}{2} h_2^2(t) + z_2 \left( \sum_{i=1}^m (b_i \rho_i u_{ci} + u_{ki}) \right) \\ &\quad + \theta^T f(x) + (k_1 + \frac{1}{2e_1}) x_2 + \frac{z_2^2}{2e_{22}} + (k_1 + \frac{1}{2e_1}) \\ &\quad \frac{z_2^2}{2e_{21}} + \tilde{\theta}^T \Gamma_{\theta}^{-1} \dot{\tilde{\theta}} + \sum_{i \in \bar{U}_{it}} \rho_i |b_i| \tilde{\varrho}^T \Gamma_{\varrho}^{-1} \dot{\tilde{\varrho}} \\ &\quad + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1}) h_1^2(t) + \frac{e_{22}}{2} h_2^2(t)\end{aligned}\quad (27)$$

We take  $\varrho$  is

$$\begin{aligned}\varrho_1 &= \frac{1}{\sum_{i \in \bar{U}_{iT}} \rho_i |b_i|} \\ \varrho_{2i} &= 0, (i \in \bar{U}_{iT}) \\ \varrho_{2i} &= -\frac{b_i u_{ki}}{\sum_{i \in \bar{U}_{iT}} \rho_i |b_i|} (i \in U_{iT})\end{aligned}\quad (28)$$

and  $\gamma$  is

$$\gamma_1 = \alpha; \quad \gamma_{2i} = 1 \quad (29)$$

Clearly, we have

$$\begin{aligned}&\sum_{i=1}^m \rho_i b_i \text{sign}(b_i) \varrho^T \gamma \\ &= \sum_{i=1}^m \rho_i |b_i| \left( \frac{1}{\sum_{i \in \bar{U}_{iT}} \rho_i |b_i|} \alpha + \sum_{i \in U_{iT}} \left( -\frac{b_i u_{ki}}{\sum_{i \in \bar{U}_{iT}} \rho_i |b_i|} \right) \right) \\ &= \sum_{i \in \bar{U}_{iT}} \rho_i |b_i| \left( \frac{1}{\sum_{i \in \bar{U}_{iT}} \rho_i |b_i|} \alpha + \sum_{i \in U_{iT}} \left( -\frac{b_i u_{ki}}{\sum_{i \in \bar{U}_{iT}} \rho_i |b_i|} \right) \right) \\ &= \alpha - \sum_{i \in U_{iT}} b_i u_{ki}\end{aligned}\quad (30)$$

With control law in (21)

$$\begin{aligned}&\sum_{i=1}^m b_i (\rho_i u_{ci} + u_{ki}) = \sum_{i \in \bar{U}_{iT}} b_i \rho_i u_{ci} + \sum_{i \in U_{iT}} b_i u_{ki} \\ &= \sum_{i \in \bar{U}_{iT}} b_i \rho_i \text{sign}(b_i) \tilde{\varrho}^T \gamma + \sum_{i \in U_{iT}} b_i u_{ki} \\ &= \sum_{i \in \bar{U}_{iT}} b_i \rho_i \text{sign}(b_i) (\tilde{\varrho} + \varrho)^T \gamma + \sum_{i \in U_{iT}} b_i u_{ki} \\ &= \sum_{i \in \bar{U}_{iT}} b_i \rho_i \text{sign}(b_i) (\tilde{\varrho} + \varrho)^T \gamma + \sum_{i \in U_{iT}} b_i u_{ki}\end{aligned}\quad (31)$$

with (30), we can get

$$\sum_{i=1}^m b_i (\rho_i u_{ci} + u_{ki}) = \sum_{i \in \bar{U}_{iT}} |b_i| \rho_i \tilde{\varrho}^T \gamma + \alpha \quad (32)$$

Then the derivative of  $V_i$  can be rewritten as

$$\begin{aligned}\dot{V}_i &\leq z_1 z_2 - k_1 z_1^2 + \frac{e_1}{2} h_2^2(t) + z_2 \left( \sum_{i \in \bar{U}_{iT}} |b_i| \rho_i \tilde{\varrho}^T \gamma + \alpha \right) \\ &\quad + \theta^T f(x) + (k_1 + \frac{1}{2e_1}) x_2 + \frac{z_2^2}{2e_{22}} + (k_1 + \frac{1}{2e_1}) \\ &\quad \frac{z_2^2}{2e_{21}} + \tilde{\theta}^T \Gamma_{\theta}^{-1} \dot{\tilde{\theta}} + \sum_{i \in \bar{U}_{it}} \rho_i |b_i| \tilde{\varrho}^T \Gamma_{\varrho}^{-1} \dot{\tilde{\varrho}} \\ &\quad + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1}) h_1^2(t) + \frac{e_{22}}{2} h_2^2(t) \\ &= -k_1 z_1^2 - k_2 z_2^2 + \sum_{i \in \bar{U}_{it}} \rho_i |b_i| \tilde{\varrho}^T \Gamma_{\varrho}^{-1} (\dot{\tilde{\varrho}} + \Gamma_{\varrho} \gamma z_2) \\ &\quad + \tilde{\theta}^T \Gamma_{\theta}^{-1} (\dot{\tilde{\theta}} - \Gamma_{\theta} f(x) z_2) + \frac{e_{22}}{2} h_2^2(t) \\ &\quad + (\frac{e_1}{2} + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1})) h_1^2(t)\end{aligned}\quad (33)$$

With update laws given in (23), we can get

$$\begin{aligned}\dot{V}_i &\leq -k_1 z_1^2 - k_2 z_2^2 + \frac{e_{22}}{2} h_2^2(t) \\ &\quad + (\frac{e_1}{2} + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1})) h_1^2(t)\end{aligned}\quad (34)$$

From Assumption 1, we have

$$\begin{aligned}\dot{V}_i &\leq -\sum_{i=1,2} k_i z_i^2 + \frac{e_{22}}{2} \delta_2^2 \|x\|_2^2 \\ &\quad + (\frac{e_1}{2} + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1})) \delta_1^2 \|x\|_2^2\end{aligned}\quad (35)$$

From coordinate changes (5) and virtual controls (12), we can get

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ -(k_1 + \frac{1}{2e_1}) & 1 \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} \quad (36)$$

The following inequality can be obtained

$$\|x\|_2 \leq (2 + (k_1 + \frac{1}{2e_1})^2)^{\frac{1}{2}} \|z\|_2 \quad (37)$$

So inequality (35) can be rewritten as

$$\begin{aligned}\dot{V}_i &\leq -\sum_{i=1,2} k_i z_i^2 + (2 + (k_1 + \frac{1}{2e_1})^2) \\ &\quad [(\frac{e_1}{2} + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1})) \delta_1^2 + \frac{e_{22}}{2} \delta_2^2] \|z\|_2^2\end{aligned}\quad (38)$$

In order to stabilize the system, the following requirement on design parameters can be established

$$\begin{cases} k_1 \geq [(\frac{e_1}{2} + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1})) \delta_1^2 + \frac{e_{22}}{2} \delta_2^2] (2 + (k_1 + \frac{1}{2e_1})^2) \\ k_2 \geq [(\frac{e_1}{2} + \frac{e_{21}}{2} (k_1 + \frac{1}{2e_1})) \delta_1^2 + \frac{e_{22}}{2} \delta_2^2] (2 + (k_1 + \frac{1}{2e_1})^2) \end{cases} \quad (39)$$

Clearly, the second inequality is easy to be realized by choosing  $k_2$  big enough. So the key lies in how to select  $k_2$  such that the first inequality. We now discuss the existence of a positive constant  $\delta^*$  such that for all  $\delta_i < \delta^*$  design parameters  $k_1$  can be chosen to make the first inequality hold.

For a group of design parameters  $k_1^*, e_1^*, e_{21}^*, e_{22}^*$ , clearly there exists a  $\delta^*$  making

$$\Phi^* = \left[ \left( \frac{e_1^*}{2} + \frac{e_{21}^*}{2} \left( k_1^* + \frac{1}{2e_1^*} \right) \right) (\delta^*)^2 + \frac{e_{22}^*}{2} (\delta^*)^2 \right] \left( 2 + \left( k_1^* + \frac{1}{2e_1^*} \right)^2 \right) \quad (40)$$

small enough and then satisfying

$$k_1^* > \Phi^* \quad (41)$$

So for all  $\delta_i < \delta^*$ , we have

$$\begin{aligned} & \left[ \left( \frac{e_1^*}{2} + \frac{e_{21}^*}{2} \left( k_1^* + \frac{1}{2e_1^*} \right) \right) (\delta_1)^2 + \frac{e_{22}^*}{2} (\delta_2)^2 \right] \\ & \left( 2 + \left( k_1^* + \frac{1}{2e_1^*} \right)^2 \right) \leq \Phi^* < k_1^* \end{aligned} \quad (42)$$

Namely, for above  $\delta^*$  when  $\delta_i < \delta^*$  design parameters  $k_1^*, e_1^*, e_{21}^*, e_{22}^*$  can make the first inequality of (39) hold. Letting

$$k^* \triangleq k_2^* = k_1^* \quad (43)$$

we can get

$$\dot{V}_i \leq -\ell \sum_{i=1}^2 z_i^2 \quad (44)$$

where

$$\ell = k^* - \Phi^* > 0 \quad (45)$$

From (44), we know  $V_i$  is non increasing in time interval  $[T_i, T_{i+1})$ . So we can get  $V_j(T_{i+1}^-) \leq V_i(T_i^+)$ . When  $i = 0$ , it is  $V_0(T_1^-) \leq V_0(0)$ . It can be conclude that all signals  $z_i, \tilde{\theta}, \tilde{\varrho}$  are bounded in the time interval  $[0, T_1)$ . Noting that the difference between  $V_1(T_1^+)$  and  $V_0(T_1^-)$  is only the coefficients in front of the term  $\tilde{\varrho}^T \Gamma_{\varrho} \tilde{\varrho}$ . Since all the possible jumping on  $\varrho$  are bounded,  $V_1(T_1^+)$  is bounded, then  $V_1(T_2^-)$  is bounded. Similar to the above analysis, we can get  $V_i(T_{i+1}^-)$  is bounded from the bound of  $V_i(T_i^+)$  and  $V_{i-1}(T_i^-)$ . Also we can get  $V_f(T_f^+)$  is bounded by the bound of  $V_{f-1}(T_f^-)$  and  $V_{f-1}(T_{f-1}^+)$ . Therefore,  $V_f$  is non increasing in time interval  $(T_f, \infty)$ . Namely, it can be shown that  $V_f(t) \leq V_f(T_f^+)$ . Then we can get  $z, \tilde{\theta}, \tilde{\varrho}$  are all bounded in  $[0, \infty]$ . Easily, we can get  $\lim_{t \rightarrow \infty} |y| = 0$  from (44).  $\square$

#### Remark 4:

- In order to prove the derivative of  $V_i$  being non-positive in time interval  $[T_i, T_{i+1})$ , linear transformation between vector  $x$  and  $z$  is established in the controller design. Such analysis is quite different from standard backstepping procedures.
- The value of parameter  $\delta^*$  is difficult to obtain for a general uncertain nonlinear system, especially for high order nonlinear systems. If so, the significance of our results is to show the existence of such a parameter. Namely, when magnitude  $\delta_i$  of time delay term  $h_i(\cdot)$  satisfying the condition that  $\delta_i < \delta^*$ , the controller shown in this paper can stabilize system (1).

## V. SIMULATION STUDIES

In this section, the results of simulation are presented to verify the effectiveness of the proposed robust adaptive neural network tracking control law for uncertain missile systems with hysteresis input. In the simulation, following parameters of the missile system are chosen: the mass of missile  $m = 204kg$ , dynamic pressure  $Q = 29938kg/m^2$ , characteristic area  $S = 0.041m^2$ , characteristic length  $d = 0.23m$ , moment of inertia  $I = 247.4kg \cdot m^2$ , mach number  $M_{nb} = 2.5$  and velocity  $v = 340m/s$ . Unknown parameters in aerodynamic force model are given as  $a_n = 0.000103, b_n = 0.00945, c_n = 0.1619, d_n = 0.034, a_m = 0.000215, b_m = -0.0195, c_m = 0.051, d_m = 0.206$ .

## VI. CONCLUSION

A robust adaptive state feedback control scheme is proposed by using backstepping technique for second order time delay systems with unknown actuator failures. Unlike existing results of time delay systems by using backstepping technology, the bounding function of time delay term  $h_i(x_1, x_2)$  in system (1) is allowed to depend on all system states. The boundedness of all signals of the closed-loop system are ensured by the proposed control law and corresponding update laws. Simulation studies verify the effectiveness of the proposed control scheme.

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**Cai Jian-Ping** Received his Ph.D. degree from Zhejiang University in 2014. He is currently a associate professor in Zhejiang University of Water Resources and Electric Power. His main research interests include nonlinear systems and adaptive control. E-mail: caijianping2001@163.com.

**Shen Lu-Juan** Received her Ph.D. degree from Zhejiang University in 2013. She is currently a associate professor in Zhejiang University of Water Resources and Electric Power. Her main research interest is application of mathematics. E-mail: shenlj@zjweu.edu.cn.