# **Output-feedback** Control for a Class of Uncertain Nonlinear Systems with Linearly Unmeasured **States Dependent Growth**

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This paper is devoted to the problem of global stabilization by output-feedback for a class of nonlinear systems with Abstract uncertain control coefficients, stable zero-dynamics, and linearly unmeasured states dependent growth. First, by introducing two kinds of appropriate state transformations, the original system is converted into a new system with deterministic virtual control coefficients and the separated zero-dynamics. Then, a suitable observer based on high-gain K-filters is constructed for the new system, and the backstepping design approach is successfully proposed to the output-feedback controller. It is shown that the global asymptotic stability of the closed-loop system can be guaranteed by the appropriate choice of the design parameters. A simulation example is also provided to show the correctness of the theoretical results and the effectiveness of the proposed approach.

Key words Nonlinear systems, uncertain control coefficients, high-gain K-filters, output-feedback, unmeasured states dependent growth, global asymptotic stability

Because the system states are incompletely measurable, the problem of output-feedback control is more challenging and difficult than that of the state-feedback control [1-5]. Unlike linear systems, the separation principle is invalid for most nonlinear systems, hence the observer and controller should be designed and analyzed together. The objective of the control design based on observer is to guarantee not only the stability of the closed-loop system but also the convergence of the estimation error. Because the theory of observer develops  $slowly^{[6-9]}$ , the output-feedback control based on observer is far from maturity, and there are still many problems  $unsolved^{[2, 10]}$ . For example, a foundational question is what the sufficient and necessary conditions are to output-feedback stabilize nonlinear systems. It has been pointed out in the remarkable paper [11] that if the power of the nonlinearity growth with respect to unmeasured states is greater than 2, there are counterexamples for which no output-feedback controls exist.

Recently, the problem of output-feedback control design has received considerable attention and been intensively investigated for a class of nonlinear systems with unmeasured states dependent growth<sup>[3, 12-18]</sup>. In [14], adaptive outputfeedback tracking control was investigated in detail for a class of nonlinear systems linearly depending on unmeasured states and in generalized output-feedback canonical form. In [12], exponentially stable output-feedback control was considered for a family of nonlinear systems that are dominated by a triangular system satisfying linear growth condition, and in [15, 19], the more general case was considered where the linear growth rate was an unknown constant. In [18], the results of [12] were generalized to the systems with unknown control coefficients by introducing the novel observer based on high-gain K-filters. Furthermore, in [3, 13, 16], the output-feedback control was investigated for systems with output dependent growth rate, and by constructing norm estimators, an extension was obtained

to the systems with unmeasured states dependent growth rate in [17].

This paper continues the investigation proposed in [18] and considers the output-feedback stabilizing control design for a more general class of nonlinear systems with uncertain control coefficients, hidden zero-dynamics, and linearly unmeasured states dependent growth, and generalizes the relevant results distributed in [1, 12, 15, 18]. Mainly thanks to the presence of the unmeasured states dependent growth and the uncertain control coefficients, the output-feedback control problem of the systems under consideration is very hard and hence a very meaningful question. In addition, due to the existence of the hidden zero-dynamics, it is difficult to directly carry out the output-feedback control design. To achieve the control objective, two kinds of linear state transformations are firstly introduced. One is to lump the uncertain control coefficients together, and then a new system with deterministic virtual control coefficients is ob-The other is to separate the zero-dynamics from tained. the system. Then, enlightened by [18], we propose the appropriate high-gain K-filters based on which the state estimation is successfully constructed. Such kind of K-filters indeed play a central role in the output-feedback control design which will be realized by the backstepping approach. Moreover, by choosing the design parameters properly, the global asymptotic stability of the closed-loop system can be guaranteed.

The remainder of this paper is organized as follows. Section 1 describes the system to be considered and formulates the control problem to be solved. Section 2 provides two kinds of linear state transformations through which the original system can be converted into a new system that is convenient for observer design and observer-based outputfeedback control design. Section 3 gives the main results of this paper, that is, the high-gain K-filters based observer is constructed and the output-feedback stabilizing control design is given using backstepping method. Section 4 presents some concluding remarks.

Notations. Throughout this paper, I denotes identity matrix of appropriate dimension; for any  $\boldsymbol{x} \in \mathbf{R}^{n}$ ,  $x_{i}$  denotes its *i*-th element and  $\boldsymbol{x}_{[i]}$  denotes  $[x_{1}, \dots, x_{i}]^{\mathrm{T}}$ ; 
$$\begin{split} \mathbf{0}_{[i]} &\in \mathbf{R}^{i} \text{ denotes the zero vector; } \mathbf{e}_{i,j}(i \leq j) \text{ denotes} \\ [\mathbf{0}_{[i-1]}^{\mathrm{T}}, \mathbf{1}, \mathbf{0}_{[j-i]}^{\mathrm{T}}]^{\mathrm{T}} &\in \mathbf{R}^{j}; \mathbf{b}_{[i,0]}' \text{ denotes } [b'_{i}, b'_{i-1}, \cdots, b'_{0}]^{\mathrm{T}} \in \\ \mathbf{R}^{i+1}; \mathbf{c}_{i,[j]} \text{ denotes } [c_{i1}, \cdots, c_{ij}]^{\mathrm{T}} \in \mathbf{R}^{j}, \text{ and } \mathbf{\Phi}_{i,[j,k]} \text{ denotes} \\ [\Phi_{i,j}, \cdots, \Phi_{i,k}]^{\mathrm{T}} \in \mathbf{R}^{k-j+1}. \text{ We use } \|\cdot\|_{1}, \|\cdot\|(\text{i.e.}, \|\cdot\|_{2}), \end{split}$$

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and  $\|\cdot\|_F$  to denote 1-norm, Euclidean norm, and Frobenius norm for vectors, respectively, and the corresponding induced norm for matrices. Besides, for simplicity of expression, we sometimes drop the arguments of function when no confusion is caused.

# 1 System description and problem formulation

Consider the following class of single-input-single-output (SISO) nonlinear systems:

$$\begin{aligned} \zeta_i &= g_i \zeta_{i+1} + \phi_i(t, \boldsymbol{\zeta}, u), & i = 1, \cdots, \rho - 1\\ \dot{\zeta}_j &= g_j \zeta_{j+1} + b_{n-j} u + \phi_j(t, \boldsymbol{\zeta}, u), & j = \rho, \cdots, n - 1\\ \dot{\zeta}_n &= g_n b_0 u + \phi_n(t, \boldsymbol{\zeta}, u) \\ y &= \zeta_1 \end{aligned}$$
(1)

where  $\boldsymbol{\zeta} = [\zeta_1, \dots, \zeta_n]^{\mathrm{T}} \in \mathbf{R}^n$  is the system state with the initial condition  $\boldsymbol{\zeta}(0) = \boldsymbol{\zeta}_0$ ;  $u \in \mathbf{R}$  and  $y \in \mathbf{R}$  are the control input and output, respectively;  $\rho$  is the relative degree of the system satisfying  $\rho + m = n$ ; both  $g_i$  and  $b_i$  are unknown constants, called uncertain control coefficients, and specially,  $g_i \neq 0$ ,  $b_0 \neq 0$ , and  $b_m \neq 0$ ; functions  $\phi_i$ :  $\mathbf{R}^+ \times \mathbf{R}^n \times \mathbf{R} \to \mathbf{R}$ ,  $i = 1, \dots, n$  are piece-wise continuous in the first argument and locally Lipschitz in the rest arguments. In what follows, we suppose that only the system output is measurable and the relative degree  $\rho > 1$ .

Obviously, we can see that control u appears in the last (m+1) equations of system (1). This means that hidden zero-dynamics exist in the system. When control coefficients are known, if  $1 < \rho < n$ , then we can construct K-filters and the output-feedback controller to stabilize system (1) by consulting Chapter 8 of [1] and if  $\rho = n$ , system (1) degenerates to a simpler case which has been extensively investigated based on the Luenberger-like  $observer^{[12-15]}$ . Moreover, when control coefficients are unknown, various theoretical results have been obtained under somewhat strong assumption that the nonlinearities are dominated by some known functions of measurable output  $y^{[1]}$ . When the system nonlinearities inherently depend on unmeasured states, the output-feedback control design will become very hard. Recently, for the systems with unmeasured states dependent growth and without zero-dynamics, [12, 15-16, 18]have considered the problem of output-feedback stabilization for the cases of exactly known control coefficients and uncertain control coefficients, respectively.

This paper is to investigate the problem of global outputfeedback stabilization for system (1) with uncertain control coefficients, zero-dynamics, and unmeasured states dependent growth under the following assumptions.

Assumption 1. For  $i = 1, \dots, n$  and any  $t \in \mathbf{R}^+$ ,  $\boldsymbol{\zeta} \in \mathbf{R}^n$ , and  $u \in \mathbf{R}$ , there exists a known constant c > 0, such that

$$|\phi_i(t,\boldsymbol{\zeta},u)| \le c(|\zeta_1| + \dots + |\zeta_k|), \quad k = \min\{i,\rho\}$$
(2)

**Assumption 2.** The signs of  $g_i, i = 1, \dots, \rho - 1$  are known, and there exist known positive constants  $\underline{g}_i$  and  $\overline{g}_i$  satisfying

$$\underline{g}_i \le |g_i| \le \overline{g}_i, \quad i = 1, \cdots, n \tag{3}$$

Assumption 3. The sign of  $b_m$  is known, and there exist known constants  $b_N > 0$  and  $\overline{b}_i > 0$ ,  $i = 0, \dots, m$  such that

$$|b_i| \le \overline{b}_i, \quad i = 0, \cdots, m - 1, \quad b_N \le |b_m| \le \overline{b}_m \quad (4)$$

Assumption 4. For the polynomial  $p(s) = b'_m s^m + \cdots + b'_1 s + b'_0$ , where  $m \ge 1$ ,  $b'_i = b_i \prod_{k=1}^{n-i-1} g_k$ ,  $i = 1, \cdots, m$ , and  $b'_0 = b_0 \prod_{k=1}^n g_k$ , there exists a known positive constant d such that the real part of each pole of the polynomial is not larger than -(d/2).

For the aim of a better understanding, we would like to give further interpretation for the above four assumptions. Assumption 1 means that system (1) has linearly unmeasured states dependent growth. It seems to be stringent, but it is crucial to carry out the output-feedback control<sup>[3, 12-15, 18, 20]</sup>. On the other hand, maybe due to the existence of hidden zero-dynamics and uncertainties in control coefficients, it is hard to relax Assumption 1 to a weaker one.

Assumptions 2 and 3 give some available information about the uncertain coefficients  $g_i$  and  $b_i$ . The known signs of  $b_m$  and  $g_i, i = 1, \dots, \rho - 1$  will play an important role in control design. Otherwise, one cannot decide the direction along which the control operates, and the closed-loop system may be unstable. The boundary restrictions imposed on  $g_i$  and  $b_i$  look somewhat severe but cannot be removed as will be detailed later.

Assumption 4 means that the hidden zero-dynamics of the system possess the input-to-state stable (ISS) property. This assumption on the zero-dynamics is commonly implicitly assumed in most of the work in robust/adaptive output-feedback control of nonlinear systems<sup>[1, 10, 13]</sup>.

According to Assumption 2, there exist known positive constants  $g_N = \min\{\underline{g}_1, \underline{g}_1\underline{g}_2, \cdots, \Pi_{j=1}^n\underline{g}_j\}$  and  $g_M = \max\{\overline{g}_1, \overline{g}_1\overline{g}_2, \cdots, \Pi_{j=1}^n\overline{g}_j\}$  such that  $g_N \leq |\Pi_{j=1}^i g_j| \leq g_M$ ,  $i = 1, \cdots, n$ .

Similarly, by Assumption 3, we know that  $b_M = \max\{\overline{b}_0, \overline{b}_1, \dots, \overline{b}_m\}$  is a known positive constant such that  $|b_i| \leq b_M$ ,  $i = 0, \dots, m-1$  and  $b_N \leq |b_m| \leq b_M$ . Moreover, by Assumptions 2 and 3, the sign of  $b'_m = b_m \prod_{k=1}^{\rho-1} g_k \neq 0$  is known.

The objective of the paper is to design a dynamic outputfeedback controller for system (1), so that the closed-loop system is globally asymptotically stable. This will be carried out based on the methods of appropriate state transformation and observer-based backstepping, as will be detailed in the next section.

#### 2 Linear state transformations

In this section, it can be shown that through two kinds of linear state transformations, the system can be successfully transformed into a system with deterministic virtual control coefficients, and the hidden zero-dynamics can be separated out, and then the output-feedback control design becomes less difficult.

By carefully examining system (1), we find that the negative influence of uncertain control coefficient,  $g_i$ , disappears if the corresponding equation is multiplied by  $g_1g_2\cdots g_{i-1}$ with the scaled state being defined as the new state. This simple observation motivates us to introduce the following linear state transformation to convert the original system into a new system in which there are no uncertain virtual control coefficients, and the original uncertain control coefficients are lumped into the actual control coefficients:

$$x_1 = \zeta_1, \ x_2 = g_1 \zeta_2, \ \cdots, \ x_n = g_1 g_2 \cdots g_{n-1} \zeta_n$$
 (5)

The dynamics of  $\boldsymbol{x} = [x_1, \cdots, x_n]^{\mathrm{T}}$  are then given by

$$\begin{cases} \dot{\boldsymbol{x}} = A_n \boldsymbol{x} + \boldsymbol{b} u + \boldsymbol{f}(t, \, \boldsymbol{x}, \, u) \\ y = x_1 \end{cases}$$
(6)

where  $\boldsymbol{x} \in \mathbf{R}^n$  is the state of the new system with the initial condition depending on  $\boldsymbol{\zeta}_0$  and (5),  $\boldsymbol{f} = [f_1, \cdots, f_n]^{\mathrm{T}} = [\phi_1, g_1 \phi_2, \cdots, g_1 g_2 \cdots g_{n-1} \phi_n]^{\mathrm{T}}$ , and

$$A_n = \begin{bmatrix} 0 & & \\ \vdots & I \\ 0 & 0 & \cdots & 0 \end{bmatrix}, \quad \boldsymbol{b} = \begin{bmatrix} \mathbf{0}_{[\rho-1]} \\ \boldsymbol{b}'_{[m,0]} \end{bmatrix}$$

It is worth pointing out that when  $\rho = n$  (or m = 0), the above transformation is enough and output-feedback control design can be pursued since there are no zero-dynamics in this particular case. For the sake of integrality in the later development, system (6) with  $\rho = n$  is rewritten as

$$\begin{cases} \Sigma_{\boldsymbol{\chi}} : \dot{\boldsymbol{\chi}} = A_{\rho} \boldsymbol{\chi} + b'_{m} \boldsymbol{e}_{\rho,\rho} u + \boldsymbol{\Phi}_{\rho,[\rho]}(t, \boldsymbol{x}, u) \\ \chi_{1} = \boldsymbol{e}_{1,\rho}^{\mathrm{T}} \boldsymbol{\chi} = y \end{cases}$$
(7)

where  $\boldsymbol{\chi} = \boldsymbol{x}, \, \boldsymbol{\Phi}_{\rho, [\rho]} = \boldsymbol{f}.$ 

However, when  $\rho < n$ , the foregoing linear transformation is still necessary but somewhat insufficient for the control design purpose. This is chiefly because certain closedloop observer should be constructed to develop an outputfeedback stabilizing controller. But due to the presence of hidden zero-dynamics in (6), it is very difficult to determine what kind of observer is appropriate as well as to analyze the closed-loop observer error performance. Mainly enlightened by the existing relevant works on the stabilization of the similar systems with hidden zero-dynamics<sup>[21-22]</sup>, an effectual idea is to look for some kind of state transformation to separate the zero-dynamics from the system. To achieve this, let us introduce the second kind of linear state transformation for system (6) with  $\rho < n$ , which will be defined in a step-by-step manner, and simultaneously, the effect and meaning for each step of the state transformation will be shown accordingly.

The effect of the first transformation  $T_1$  (i.e., the linear transformation matrix) is to transform all the elements of **B** into zero except the  $\rho$ -th element. Let  $\boldsymbol{\varsigma}_1 = T_1 \boldsymbol{x}$ , where  $T_1$  and  $T_1^{-1}$  are the same as I except their  $\rho$ -th columns that are  $[\mathbf{0}_{[\rho-1]}^{\mathrm{T}}, 1, -b'_{m-1}/b'_m, \cdots, -b'_0/b'_m]^{\mathrm{T}}$  and  $[\mathbf{0}_{[\rho-1]}^{\mathrm{T}}, 1, b'_{m-1}/b'_m, \cdots, b'_0/b'_m]^{\mathrm{T}}$ , respectively. Then, we can get the dynamics of  $\boldsymbol{\varsigma}_1$  as

$$\dot{\boldsymbol{\varsigma}}_1 = C_1 \boldsymbol{\varsigma}_1 + \boldsymbol{B} u + \boldsymbol{\Phi}_1(t, \boldsymbol{x}, u) \tag{8}$$

where  $C_1$  is the same as  $A_n$  except its  $\rho$ -th and  $(\rho + 1)$ -th columns, which are  $[\mathbf{0}_{[\rho-2]}^{\mathrm{T}}, 1, \mathbf{c}_{1,[m+1]}^{\mathrm{T}}]^{\mathrm{T}}$  and  $[\mathbf{0}_{[\rho-1]}^{\mathrm{T}}, 1, -\frac{1}{b'_m} \mathbf{b}_{[m-1,0]}^{\prime \mathrm{T}}]^{\mathrm{T}}$ , respectively, and  $\mathbf{B} = T_1 \mathbf{b} = [\mathbf{0}_{[\rho-1]}^{\mathrm{T}}, \mathbf{b}_m', \mathbf{0}_{[m]}^{\mathrm{T}}]^{\mathrm{T}} = b'_m \mathbf{e}_{\rho,n}, \mathbf{\Phi}_1 = T_1 \mathbf{f}(t, \mathbf{x}, u).$ The rest  $\rho - 1$  transformations are designed mainly for

The rest  $\rho - 1$  transformations are designed mainly for matrix  $C_1$  such that we can eventually get  $C_{\rho}$  which is the same as matrix I except the first and  $(\rho + 1)$ -th columns. Specifically, the second transformation is to transform the  $\rho$ -th column of  $C_1$  into  $\boldsymbol{e}_{\rho-1,n}$ , i.e.,  $[\boldsymbol{0}_{[\rho-2]}^{\mathrm{T}}, 1, \boldsymbol{0}_{[m+1]}^{\mathrm{T}}]^{\mathrm{T}}$ . Let  $\boldsymbol{\varsigma}_2 = T_2 \boldsymbol{\varsigma}_1$ , where  $T_2$  and  $T_2^{-1}$  are the same as I except their  $(\rho - 1)$ -th columns that are  $[\boldsymbol{0}_{[\rho-2]}^{\mathrm{T}}, 1, -\boldsymbol{c}_{1,[m+1]}^{\mathrm{T}}]^{\mathrm{T}}$  and  $[\boldsymbol{0}_{[\rho-2]}^{\mathrm{T}}, 1, \boldsymbol{c}_{1,[m+1]}^{\mathrm{T}}]^{\mathrm{T}}$ , respectively. By noting that  $\boldsymbol{B} = T_2\boldsymbol{B}$ , the dynamics of  $\boldsymbol{\varsigma}_2$  is

$$\dot{\boldsymbol{\varsigma}}_2 = C_2 \boldsymbol{\varsigma}_2 + \boldsymbol{B} \boldsymbol{u} + \boldsymbol{\Phi}_2(t, \boldsymbol{x}, \boldsymbol{u}) \tag{9}$$

where  $C_2$  is the same as  $A_n$  except its  $(\rho - 1)$ -th and  $(\rho+1)$ -th columns which are  $[\mathbf{0}_{[\rho-3]}^{\mathrm{T}}, 1, \mathbf{c}_{2,[m+2]}^{\mathrm{T}}]^{\mathrm{T}}$  and  $[\mathbf{0}_{[\rho-1]}^{\mathrm{T}}, 1, -\frac{1}{b'_m}\mathbf{b}'_{[m-1,0]}]^{\mathrm{T}}$ , respectively, and  $\mathbf{\Phi}_2 = T_2\mathbf{\Phi}_1(t, \mathbf{x}, u)$ .

Similarly, the *i*-th  $(i = 3, \dots, \rho-1)$  linear transformation  $T_i$  can be constructed to transform the  $(\rho-i+2)$ -th column of  $C_{i-1}$  into  $\boldsymbol{e}_{\rho-i+1,n}$ . Let  $\boldsymbol{\varsigma}_i = T_i \boldsymbol{\varsigma}_{i-1}$ , where  $T_i$  and  $T_i^{-1}$  are the same as I except their  $(\rho-i+1)$ -th columns, which are  $[\mathbf{0}_{[\rho-i]}^{\mathrm{T}}, 1, -\boldsymbol{c}_{i-1,[m+i-1]}^{\mathrm{T}}]^{\mathrm{T}}$  and  $[\mathbf{0}_{[\rho-i]}^{\mathrm{T}}, 1, \boldsymbol{c}_{i-1,[m+i-1]}^{\mathrm{T}}]^{\mathrm{T}}$ , respectively. By noting that  $\boldsymbol{B} = T_i T_{i-1} \cdots T_2 \boldsymbol{B}$ , the dynamics of  $\boldsymbol{\varsigma}_i$  is

$$\dot{\boldsymbol{\varsigma}}_i = C_i \boldsymbol{\varsigma}_i + \boldsymbol{B} \boldsymbol{u} + \boldsymbol{\Phi}_i(t, \boldsymbol{x}, \boldsymbol{u}) \tag{10}$$

where  $C_i$  is the same as  $A_n$  except its  $(\rho - i + 1)$ -th and  $(\rho + 1)$ -th columns, which are  $[\mathbf{0}_{[\rho-i-1]}^{\mathrm{T}}, 1, \mathbf{c}_{i,[m+i]}^{\mathrm{T}}]^{\mathrm{T}}$  and  $[\mathbf{0}_{[\rho-1]}^{\mathrm{T}}, 1, -\frac{1}{b'_m} \mathbf{b}'_{[m-1,0]}]^{\mathrm{T}}$ , respectively, and  $\mathbf{\Phi}_i = T_i \mathbf{\Phi}_{i-1}(t, \mathbf{x}, u)$ .

Finally, the  $\rho$ -th linear transformation  $\boldsymbol{\varsigma}_{\rho} = T_{\rho} \boldsymbol{\varsigma}_{\rho-1}$  can be constructed to transform the second column of  $C_{\rho-1}$ into  $\boldsymbol{e}_{1,n}$ , where  $T_{\rho}$  and  $T_{\rho}^{-1}$  are the same as I except their first columns that are  $[1, -\boldsymbol{c}_{\rho-1,[n-1]}^{\mathrm{T}}]^{\mathrm{T}}$  and  $[1, \boldsymbol{c}_{\rho-1,[n-1]}^{\mathrm{T}}]^{\mathrm{T}}$ , respectively. By noting that  $\boldsymbol{B} = T_{\rho}T_{\rho-1}\cdots T_{2}\boldsymbol{B}$ , the dynamics of  $\boldsymbol{\varsigma}_{\rho}$  is

$$\dot{\boldsymbol{\varsigma}}_{\rho} = C_{\rho} \boldsymbol{\varsigma}_{\rho} + \boldsymbol{B} \boldsymbol{u} + \boldsymbol{\Phi}_{\rho}(t, \boldsymbol{x}, \boldsymbol{u}) \tag{11}$$

where  $C_{\rho}$  is the same as  $A_n$  except its first and  $(\rho + 1)$ -th columns, which are  $\boldsymbol{c}_{\rho,[n]}$  and  $[\boldsymbol{0}_{[\rho-1]}^{\mathrm{T}}, 1, -\frac{1}{b'_m}\boldsymbol{b}_{[m-1,0]}^{\mathrm{T}}]^{\mathrm{T}}$ , respectively, and  $\boldsymbol{\Phi}_{\rho} = T_{\rho}\boldsymbol{\Phi}_{\rho-1} = [\Phi_{\rho,1}, \cdots, \Phi_{\rho,n}]^{\mathrm{T}}$ .

Till now, the two kinds of linear state transformations are completely introduced. For the transformed system (11), it is natural to design observer and observer-based output-feedback control, as will be seen from the later development.

First, we would like to give the 2-norm estimation of the unknown vectors  $\boldsymbol{c}_{\rho,[\rho]}$  and  $\boldsymbol{c}_{\rho,[\rho+1,n]}$  which can be deduced from the above state transformations though a bit complicatedly. In fact, from its definition, one can see that each element of  $\boldsymbol{c}_{i,[m+i]}$  is in the form of a quotient whose denominator is a power function of  $b'_m$  and numerator a polynomial of the constants  $b'_0, b'_1, \dots, b'_m$ . From this observation and Assumptions 2 and 3, we can easily verify the following proposition.

**Proposition 1.** For  $c_{\rho,[n]}$  introduced above, there exist known positive constants  $d_0$  and  $d'_0$  such that

$$\|\boldsymbol{c}_{\rho,[\rho]}\| \le d_0, \quad \|\boldsymbol{c}_{\rho,[\rho+1,n]}\| \le d'_0$$
 (12)

Before designing high-gain K-filters based observer and output-feedback controller, we have to decompose system (11) appropriately. Let  $\boldsymbol{\chi} = [\chi_1, \cdots, \chi_p]^{\mathrm{T}} = \boldsymbol{\varsigma}_{\rho,[1,\rho]}$  and  $\boldsymbol{\eta} = [\eta_1, \cdots, \eta_m]^{\mathrm{T}} = \boldsymbol{\varsigma}_{\rho,[\rho+1,n]}$ . Then, system (11) can be divided into the following form:

$$\begin{cases} \Sigma_{\boldsymbol{\eta}} : \dot{\boldsymbol{\eta}} = D\boldsymbol{\eta} + \boldsymbol{c}_{\rho,[\rho+1,n]}\chi_1 + \boldsymbol{\Phi}_{\rho,[\rho+1,n]}(t, \boldsymbol{x}, u) \\ \Sigma_{\boldsymbol{\chi}} : \dot{\boldsymbol{\chi}} = A_{\rho}\boldsymbol{\chi} + b'_{m}\boldsymbol{e}_{\rho,\rho}u + \boldsymbol{e}_{\rho,\rho}\eta_1 + \boldsymbol{c}_{\rho,[\rho]}\chi_1 + \boldsymbol{\Phi}_{\rho,[\rho]}(t, \boldsymbol{x}, u) \\ \chi_1 = \boldsymbol{e}_{1,\rho}^{\mathrm{T}}\boldsymbol{\chi} = y \end{cases}$$
(13)

where 
$$D = \left[-\frac{\boldsymbol{b}'_{[m-1,0]}}{\boldsymbol{b}'_{m}}, [I, \boldsymbol{0}_{[m-1]}]^{\mathrm{T}}\right]$$
 is Hurwitz ensured by Assumption 4. So there is a positive definite matrix

$$Q = Q^{\mathrm{T}} = \int_0^\infty \exp(D^{\mathrm{T}}t) \exp(Dt) \mathrm{d}t$$
(14)

which satisfies the Lyapunov equation  $D^{\mathrm{T}}Q + QD = -I$ .

We have the following lemma for the 2-norm of Q, whose proof is omitted.

**Lemma 1.** For any Hurwitz matrix D with  $\operatorname{Re}(\lambda(D)) \leq -d/2 < 0$ , the matrix Q defined by (14) satisfies  $||Q|| \leq n/d$ .

The following lemma is natural and can be easily proved with Assumptions  $1\sim 3.$ 

**Lemma 2.** For  $\forall (t, \boldsymbol{x}, u) \in \mathbf{R}^+ \times \mathbf{R}^n \times \mathbf{R}$  and  $\boldsymbol{\chi} = [I \quad 0_{\rho \times m}]T_{\rho} \cdots T_1 \boldsymbol{x} \in \mathbf{R}^{\rho}$ , there is a known positive constant c' such that the unknown nonlinearities  $\Phi_{\rho,i} : \mathbf{R}^+ \times \mathbf{R}^n \times \mathbf{R} \to \mathbf{R}, \ i = 1, \cdots, n$  of system (7) or system (13) satisfy

$$|\Phi_{\rho,i}(t,\boldsymbol{x},u)| \le c'(|\chi_1| + \dots + |\chi_k|), \quad k = \min\{i,\rho\} \quad (15)$$

# 3 Output-feedback stabilizing control design

This section is the core of present paper. We will firstly construct the appropriate high-gain K-filters based observer, and then design the output-feedback controller by the backstepping method.

This section turns to output-feedback control design for system (7) ( $\rho = n$ ) or system (13) ( $\rho < n$ ). Since they are equivalent to system (1), the global asymptotic stability of the transformed and original systems can be achieved simultaneously. This section includes three parts: in Subsection 3.1, a novel approach for observer design will be proposed based on high-gain K-filters for system (7) or the subsystem  $\Sigma_{\chi}$  of system (13); Subsection 3.2 is restricted to the case of  $\rho < n$ , where a globally asymptotically stabilizing output-feedback controller will be constructed by the traditional backstepping method<sup>[1]</sup>, and the case of  $\rho = n$ is simple which will be studied in the next section by an example; in Subsection 3.3, the main result of the paper is summarized, which shows that the global asymptotic stability of the closed-loop system can be guaranteed by the appropriate choice of the design parameters.

#### 3.1 High-gain K-filters and state estimation

For system (7) or the subsystem  $\Sigma_{\boldsymbol{\chi}}$  of system (13)<sup>[18]</sup>, we introduce the following high-gain K-filters:

$$\begin{cases} \dot{\boldsymbol{\xi}} = A_{\boldsymbol{l}_{\varepsilon}}\boldsymbol{\xi} - \boldsymbol{l}_{\varepsilon}y\\ \dot{\boldsymbol{\lambda}} = A_{\boldsymbol{l}_{\varepsilon}}\boldsymbol{\lambda} + \boldsymbol{e}_{\rho,\rho}u \end{cases}$$
(16)

where  $\varepsilon \in (0,1)$  is a constant,  $\boldsymbol{l}_{\varepsilon} = \left[\frac{l_1}{\varepsilon}, \frac{l_2}{\varepsilon^2}, \cdots, \frac{l_{\rho}}{\varepsilon^{\rho}}\right]^{\mathrm{T}}$ , and  $\boldsymbol{\lambda}_{\varepsilon} = \boldsymbol{\lambda}_{\varepsilon} + \boldsymbol{l}_{\varepsilon} \boldsymbol{e}^{\mathrm{T}}$ 

 $A_{\boldsymbol{l}_{\varepsilon}} = A_{\rho} + \boldsymbol{l}_{\varepsilon} \boldsymbol{e}_{1,\rho}^{\mathrm{T}}.$ 

The constant vector  $\mathbf{l} = [l_1, \dots, l_{\rho}]^{\mathrm{T}}$  is chosen to ensure that matrix  $A_{\mathbf{l}} = A_{\rho} + \mathbf{l} \mathbf{e}_{1,\rho}^{\mathrm{T}}$  is Hurwitz. Then, there is a unique symmetric positive definite matrix  $P_{\mathbf{l}}$  satisfying  $A_{\mathbf{l}}^{\mathrm{T}} P_{\mathbf{l}} + P_{\mathbf{l}} A_{\mathbf{l}} = -I$ . Define  $I_{\varepsilon} = \text{diag}\{1, \varepsilon, \dots, \varepsilon^{\rho-1}\}$ . It can be shown that  $\varepsilon A_{\mathbf{l}_{\varepsilon}} = I_{\varepsilon}^{-1} A_{\mathbf{l}} I_{\varepsilon}$ . So matrix  $P_{\mathbf{l}_{\varepsilon}} = I_{\varepsilon} P_{\mathbf{l}} I_{\varepsilon}$ is symmetric positive definite, and satisfies

$$A_{\boldsymbol{l}_{\varepsilon}}^{\mathrm{T}} P_{\boldsymbol{l}_{\varepsilon}} + P_{\boldsymbol{l}_{\varepsilon}} A_{\boldsymbol{l}_{\varepsilon}} = -\varepsilon^{-1} I_{\varepsilon}^{2}$$
(17)

This means that  $A_{l_{\varepsilon}}$  is also a Hurwitz matrix.

Define the state estimate of state  $\boldsymbol{\chi}$  as  $\hat{\boldsymbol{\chi}} = \boldsymbol{\xi} + b'_m \boldsymbol{\lambda}$ , and the state estimation error as  $\tilde{\boldsymbol{\chi}} = \boldsymbol{\chi} - \hat{\boldsymbol{\chi}}$ .

In the following, we will consider the properties of the state estimation error for the cases of  $\rho < n$  and  $\rho = n$ , respectively.

**Case 1.** When  $\rho < n$ , the state estimation error  $\tilde{\chi}$  satisfies

$$\tilde{\boldsymbol{\chi}} = A_{\boldsymbol{l}_{\varepsilon}} \tilde{\boldsymbol{\chi}} + \boldsymbol{e}_{\rho,\rho} \eta_1 + \boldsymbol{c}_{\rho,[\rho]} \chi_1 + \boldsymbol{\Phi}_{\rho,[\rho]}$$
(18)

To prepare for the backstepping procedure in the next subsection, we have to rewrite the equation of  $\dot{y}$ . Since the state  $\chi_2$  is not available, we need to replace  $\chi_2$  by  $\tilde{\chi}_2 + \xi_2 + b'_m \lambda_2$ . Then, we have

$$\dot{y} = b'_m \lambda_2 + c_{\rho,1} y + \tilde{\chi}_2 + \xi_2 + \Phi_{\rho,1} \tag{19}$$

Till now, the whole system for control design is obtained, i.e.,

$$\begin{cases} \tilde{\boldsymbol{\chi}} = A_{\boldsymbol{l}_{\varepsilon}} \tilde{\boldsymbol{\chi}} + \boldsymbol{e}_{\rho,\rho} \eta_{1} + \boldsymbol{c}_{\rho,[\rho]} \chi_{1} + \boldsymbol{\Phi}_{\rho,[\rho]} \\ \dot{\boldsymbol{\xi}} = A_{\boldsymbol{l}_{\varepsilon}} \boldsymbol{\xi} - \boldsymbol{l}_{\varepsilon} y \\ \dot{\boldsymbol{y}} = b'_{m} \lambda_{2} + c_{\rho,1} y + \tilde{\chi}_{2} + \xi_{2} + \boldsymbol{\Phi}_{\rho,1} \\ \dot{\lambda}_{i} = \frac{l_{i}}{\varepsilon^{i}} \lambda_{1} + \lambda_{i+1}, \quad i = 2, \cdots, \rho - 1 \\ \dot{\lambda}_{\rho} = \frac{l_{\rho}}{\varepsilon^{\rho}} \lambda_{1} + u \end{cases}$$

$$(20)$$

Obviously, the study of the globally asymptotic stabilization of the above system is equivalent to that of system (13), as well as the original system (1). So, it is appropriate to consider system (20).

Let us next study the property of the subsystem  $\Sigma_{\eta}$ . The following proposition shows that this subsystem is ISS and can be regarded as the zero-dynamics of the whole system (13). The proof of the proposition is provided in Appendix A.

**Proposition 2.** For the subsystem  $\Sigma_{\eta}$  of system (13), let  $V_{\eta} = \eta^{\mathrm{T}} Q \eta$ . Then, there are known positive constants  $d_{0,1}, d_{0,2}$ , and  $d_{0,3}$ , independent of  $\varepsilon$ , such that

$$\dot{V}_{\boldsymbol{\eta}} \leq -\frac{3}{8} \|\boldsymbol{\eta}\|^{2} + d_{0,1} \left( d_{0}^{\prime 2} + m^{2} c^{\prime 2} \varepsilon^{-(2\rho-2)} \right) y^{2} + \varepsilon^{-(2\rho-2)} d_{0,2} \|I_{\varepsilon} \tilde{\boldsymbol{\chi}}\|^{2} + \varepsilon^{-(2\rho-2)} d_{0,2} \|I_{\varepsilon} \boldsymbol{\xi}\|^{2} + \varepsilon^{-(2\rho-2)} d_{0,3} \sum_{i=2}^{\rho} \varepsilon^{2i-2} \lambda_{i}^{2}$$

$$(21)$$

for any  $\tilde{\boldsymbol{\chi}}, \boldsymbol{\xi}, \boldsymbol{\eta}, y, \lambda_i, i = 2, \cdots, \rho$ , where  $d'_0$  has been specified in Proposition 1.

The following two propositions play an important role in control design in the sequent subsections. Specifically, Propositions 3 and 4 characterize the ISS-like properties of  $\tilde{\chi}$  and  $\boldsymbol{\xi}$  of system (20), respectively. Besides, for the sake of compactness, the proof of Proposition 3 is provided in Appendix B.

**Proposition 3.** For the subsystem  $\tilde{\boldsymbol{\chi}}$  of system (20), let  $V_{\tilde{\boldsymbol{\chi}}} = \tilde{\boldsymbol{\chi}}^{\mathrm{T}} \boldsymbol{P}_{l_{\varepsilon}} \tilde{\boldsymbol{\chi}}$ . Then, there are known positive constants  $d_{0,4} \sim d_{0,7}$  independent of  $\varepsilon$ , such that

$$\dot{V}_{\tilde{\boldsymbol{\chi}}} \leq -\left(\frac{1}{2\varepsilon} - d_{0,4}\right) \|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} + d_{0,5}\|I_{\varepsilon}\boldsymbol{\xi}\|^{2} + 2\varepsilon^{2\rho-1}\|P_{l}\|^{2}\|\boldsymbol{\eta}\|^{2} + d_{0,6}y^{2} + d_{0,7}\sum_{i=2}^{\rho}\varepsilon^{2i-2}\lambda_{i}^{2}$$
(22)

for any  $\tilde{\boldsymbol{\chi}}, \boldsymbol{\xi}, \boldsymbol{\eta}, y, \lambda_i, i = 2, \cdots, \rho$ .

**Proposition 4.** For the subsystem  $\boldsymbol{\xi}$  of system (20), let  $V_{\boldsymbol{\xi}} = \boldsymbol{\xi}^{\mathrm{T}} P_{l_{\varepsilon}} \boldsymbol{\xi}$ . Then, for any  $\boldsymbol{\xi}, y$ , the following inequality holds:

$$\dot{V}_{\boldsymbol{\xi}} \leq -\frac{1}{2\varepsilon} \|I_{\varepsilon}\boldsymbol{\xi}\|^2 + \frac{2\|P_{\boldsymbol{\ell}}\|^2 \|\boldsymbol{\ell}\|^2}{\varepsilon} y^2 \tag{23}$$

**Proof.** Noting that  $\boldsymbol{l}_{\varepsilon} = (1/\varepsilon)I_{\varepsilon}^{-1}\boldsymbol{l}$  and  $P_{\boldsymbol{l}_{\varepsilon}} = I_{\varepsilon}P_{\boldsymbol{l}}I_{\varepsilon}$ , we can show that the time derivative of  $V_{\boldsymbol{\xi}}$  satisfies

$$\dot{V}_{\boldsymbol{\xi}} = -\frac{\|I_{\varepsilon}\boldsymbol{\xi}\|^{2}}{\varepsilon} - 2\boldsymbol{\xi}^{\mathrm{T}}I_{\varepsilon}P_{\boldsymbol{t}}I_{\varepsilon}\boldsymbol{l}_{\varepsilon}y = -\frac{\|I_{\varepsilon}\boldsymbol{\xi}\|^{2}}{\varepsilon} - \frac{2}{\varepsilon}\boldsymbol{\xi}^{\mathrm{T}}I_{\varepsilon}P_{\boldsymbol{t}}\boldsymbol{l}y$$

**Case 2.** When  $\rho = n$ , the state estimation error  $\tilde{\chi}$  satisfies

$$\dot{\tilde{\boldsymbol{\chi}}} = A_{\boldsymbol{l}_{\varepsilon}} \tilde{\boldsymbol{\chi}} + \boldsymbol{\Phi}_{\rho,[\rho]} \tag{24}$$

Similarly, we rewrite the equation of  $\dot{y} = \chi_2 + \Phi_{\rho,1}$  as  $\dot{y} = b'_m \lambda_2 + \tilde{\chi}_2 + \xi_2 + \Phi_{\rho,1}$  and thus obtain the whole system for control design:

$$\begin{cases} \dot{\tilde{\boldsymbol{\chi}}} = A_{\boldsymbol{l}_{\varepsilon}} \tilde{\boldsymbol{\chi}} + \boldsymbol{\Phi}_{\rho,[\rho]} \\ \dot{\boldsymbol{\xi}} = A_{\boldsymbol{l}_{\varepsilon}} \boldsymbol{\xi} - \boldsymbol{l}_{\varepsilon} y \\ \dot{\boldsymbol{y}} = b'_{m} \lambda_{2} + \tilde{\chi}_{2} + \xi_{2} + \boldsymbol{\Phi}_{\rho,1} \\ \dot{\lambda}_{i} = \frac{l_{i}}{\varepsilon^{i}} \lambda_{1} + \lambda_{i+1}, \quad i = 2, \cdots, \rho - 1 \\ \dot{\lambda}_{\rho} = \frac{l_{\rho}}{\varepsilon^{\rho}} \lambda_{1} + u \end{cases}$$

$$(25)$$

## 3.2 Output-feedback control design

This subsection is devoted to the constructive design of output-feedback control for system (20) with  $\rho < n$  by the traditional backstepping method, which is presented in a step-by-step manner.

**Step 1.** Define  $V_0 = \varepsilon^{2\rho-2}V_{\eta} + V_{\tilde{\chi}} + V_{\xi}$ , where  $V_{\eta}, V_{\tilde{\chi}}$ , and  $V_{\xi}$  have been defined, respectively, in Propositions  $2 \sim 4$ . Let  $V_1 = V_0 + (1/2)y^2$  be the Lyapunov function candidate for this step.

From Proposition 1 and by the method of completing square, it is evident that

$$c_{\rho,1}y^{2} \leq d_{0}y^{2}, \quad y\tilde{\chi}_{2} \leq \frac{1}{4\varepsilon} \|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} + \frac{1}{\varepsilon}y^{2}$$
$$y\xi_{2} \leq \frac{1}{4\varepsilon} \|I_{\varepsilon}\boldsymbol{\xi}\|^{2} + \frac{1}{\varepsilon}y^{2}, \quad y\Phi_{\rho,1} \leq c'y^{2}$$
(26)

Then, from Propositions  $2 \sim 4$ , (19) and (26), and letting  $d'_{1,1} = d_{0,2} + d_{0,4}$ ,  $d'_{1,2} = d_{0,2} + d_{0,5}$ , and  $d_{\lambda} = d_{0,3} + d_{0,7}$ , we have

$$\dot{V}_{1} \leq -h_{\eta}(\varepsilon) \|\boldsymbol{\eta}\|^{2} - \left(\frac{1}{4\varepsilon} - d_{1,1}^{\prime}\right) \|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} - \left(\frac{1}{4\varepsilon} - d_{1,2}^{\prime}\right) \times \|I_{\varepsilon}\boldsymbol{\xi}\|^{2} + \left(d_{0}^{\prime 2}d_{0,1}\varepsilon^{2\rho-2} + m^{2}c^{\prime 2}d_{0,1} + d_{0,6} + c^{\prime} + d_{0} + \frac{2\|P_{I}\|^{2}\|\boldsymbol{\ell}\|^{2} + 2}{\varepsilon}\right)y^{2} + d_{\boldsymbol{\lambda}}\sum_{i=3}^{\rho}\varepsilon^{2i-2}\lambda_{i}^{2} + d_{\boldsymbol{\lambda}}\varepsilon^{2}\lambda_{2}^{2} + b_{m}^{\prime}y\lambda_{2}$$
(27)

where  $h_{\eta}(\varepsilon) = \varepsilon^{2\rho-1} \left( (3/8\varepsilon) - 2 \| P_{\mathbf{l}} \|^2 \right)$ . Choose the virtual controller as

$$\lambda_2^* = -\frac{1}{\varepsilon} \left( \frac{\operatorname{sgn}(b'_m)}{g_N b_N} L_1 + \bar{\alpha}_1 \right) y = -\frac{\alpha_1(L_1)}{\varepsilon} y \qquad (28)$$

where  $\bar{\alpha}_1 = \frac{\text{sgn}(b'_m)}{g_N b_N} (d_0'^2 d_{0,1} + m^2 c'^2 d_{0,1} + d_{0,6} + c' + d_0 + 2 \|P_l\|^2 \|l\|^2 + 2)$  is clearly a known constant, and  $L_1 \ge 1$  is a constant to be determined later.

Define  $z_1 = y$  and  $z_2 = \lambda_2 - \lambda_2^*$ . Observe that

$$d_{\boldsymbol{\lambda}}\varepsilon^{2}\lambda_{2}^{2} \leq 5d_{\boldsymbol{\lambda}}\varepsilon^{2}z_{2}^{2} + \mu_{1}^{z_{1}}(L_{1})y^{2} + \mu_{1}^{\tilde{\boldsymbol{\lambda}}} \|I_{\varepsilon}\tilde{\boldsymbol{\lambda}}\|^{2} + \mu_{1}^{\boldsymbol{\xi}} \|I_{\varepsilon}\boldsymbol{\xi}\|^{2}$$
(29)

where  $\mu_1^{z_1}(L_1) = 2d_{\lambda}\alpha_1^2(L_1)$  and  $\mu_1^{\tilde{\chi}} = \mu_1^{\xi} = 0$  for the initial assignment of the forthcoming inductive step. Then,

substituting (28) and (29) into (27) results in

$$\dot{V}_{1} \leq -h_{\eta}(\varepsilon) \|\boldsymbol{\eta}\|^{2} - h_{1,1}(\varepsilon) \|I_{\varepsilon} \tilde{\boldsymbol{\chi}}\|^{2} - h_{1,2}(\varepsilon) \|I_{\varepsilon} \boldsymbol{\xi}\|^{2} - h_{1,3}(\varepsilon, L_{1})y^{2} + d_{\boldsymbol{\lambda}} \sum_{i=3}^{\rho} \varepsilon^{2i-2} \lambda_{i}^{2} + 5d_{\boldsymbol{\lambda}} \varepsilon^{2} z_{2}^{2} + b'_{m} y z_{2}$$

$$(30)$$

where  $h_{1,1}(\varepsilon) = \frac{1}{4\varepsilon} - d_{1,1}, h_{1,2}(\varepsilon) = \frac{1}{4\varepsilon} - d_{1,2}, h_{1,3}(\varepsilon, L_1) = \frac{L_1}{\varepsilon} - d_{1,3}(L_1), d_{1,1} = d'_{1,1} + \mu_1^{\tilde{\mathbf{X}}}, d_{1,2} = d'_{1,2} + \mu_1^{\tilde{\mathbf{X}}}, and d_{1,3}(L_1) = \mu_1^{z_1}(L_1).$  In what follows,  $h_{i,1}(\varepsilon, \mathbf{L}_{[i-1]}) = \frac{1}{2^{i+1}\varepsilon} - d_{i,1}(\mathbf{L}_{[i-1]}), h_{i,2}(\varepsilon, \mathbf{L}_{[i-1]}) = \frac{1}{4\varepsilon} - d_{i,2}(\mathbf{L}_{[i-1]}), h_{i,3}(\varepsilon, \mathbf{L}_{[i]}) = \frac{L_1 - i}{\varepsilon} - d_{i,3}(\mathbf{L}_{[i]}), h_{i,4}(\varepsilon, \mathbf{L}_{[i]}) = \frac{L_1 - i + 1}{\varepsilon} - d_{i,4}(\mathbf{L}_{[i]}), h_{i,j+2}(\varepsilon, \mathbf{L}_{[i]}) = \frac{L_j - d_{i,j+2}(\mathbf{L}_{[i]}), j = 3, \cdots, i - 1, h_{i,i+2}(\varepsilon, \mathbf{L}_{[i]}) = \frac{L_i}{\varepsilon} - d_{i,i+2}(\mathbf{L}_{[i]}), all d_{i,j}$ 's will be clearly defined in later steps.

**Step 2.** Let  $V_2 = V_1 + (1/2)\varepsilon^2 z_2^2$ . Then, the time derivative of  $V_2$  satisfies

$$\dot{V}_2 = \dot{V}_1 + \varepsilon^2 z_2 \dot{z}_2 \tag{31}$$

By the definition of  $z_2$ , we have

$$\dot{z}_{2} = \lambda_{3} + \sum_{j=1}^{2} \gamma_{2}^{\lambda_{j}}(L_{1}) \varepsilon^{-(3-j)} \lambda_{j} + \sum_{j=1}^{2} \gamma_{2}^{\xi_{j}}(L_{1}) \varepsilon^{-(3-j)} \times \\ \xi_{j} + \gamma_{2}^{y}(L_{1}) \varepsilon^{-2} y + \gamma_{2}^{\Delta}(L_{1}) \varepsilon^{-1} (c_{\rho,1} y + b'_{m} \lambda_{2} + \tilde{\chi}_{2} + \Phi_{\rho,1})$$
(32)

with specified  $\gamma_2^{\lambda_2}(L_1) = 0, \gamma_2^{\xi_1}(L_1) = 0$ , and  $\gamma_2^y(L_1) = 0$  for the initial assignment of the forthcoming inductive step.

Before deriving the virtual controller  $\lambda_3^*$ , we should eliminate the "undesired" effect of  $z_2$  in (31). For this purpose, by the method of completing square, we have

$$b'_{m}yz_{2} \le b_{M}g_{M}|y||z_{2}| \le \frac{1}{\varepsilon}y^{2} + \frac{1}{4}\max\{1, b_{M}^{2}g_{M}^{2}\}\varepsilon z_{2}^{2}$$
 (33)

and the following inequalities

$$\begin{cases} \varepsilon \gamma_{2}^{\Delta}(L_{1})c_{\rho,1}yz_{2} \leq \frac{1}{4}d_{0}^{2}(\gamma_{2}^{\Delta}(L_{1}))^{2}\varepsilon^{2}z_{2}^{2} + y^{2} \\ b'_{m}\varepsilon \gamma_{2}^{\Delta}(L_{1})\lambda_{2}z_{2} \leq \frac{1}{\varepsilon}y^{2} + \left(b_{M}g_{M}|\gamma_{2}^{\Delta}(L_{1})| + \frac{1}{4}b_{M}^{2}g_{M}^{2}\alpha_{1}^{2}(L_{1})(\gamma_{2}^{\Delta}(L_{1}))^{2}\right)\varepsilon z_{2}^{2} \quad (34) \\ \varepsilon \gamma_{2}^{\Delta}(L_{1})\tilde{\chi}_{2}z_{2} \leq \frac{1}{8\varepsilon}\|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} + 2(\gamma_{2}^{\Delta}(L_{1}))^{2}\varepsilon z_{2}^{2} \\ \varepsilon \gamma_{2}^{\Delta}(L_{1})\Phi_{\rho,1}z_{2} \leq \frac{1}{4}c'^{2}(\gamma_{2}^{\Delta}(L_{1}))^{2}\varepsilon^{2}z_{2}^{2} + y^{2} \end{cases}$$

Substituting (32)  $\sim$  (34) into (31), we have

$$\dot{V}_{2} \leq -h_{\eta}(\varepsilon) \|\boldsymbol{\eta}\|^{2} - \left(h_{1,1}(\varepsilon) - \frac{1}{8\varepsilon}\right) \|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} - h_{1,2}(\varepsilon)\|I_{\varepsilon}\boldsymbol{\xi}\|^{2} - \left(h_{1,3}(\varepsilon, L_{1}) - \frac{2}{\varepsilon} - 2\right)y^{2} + d_{\boldsymbol{\lambda}}\sum_{i=4}^{\rho}\varepsilon^{2i-2}\lambda_{i}^{2} + \varepsilon^{2}z_{2}\left(\sum_{j=1}^{2}\gamma_{2}^{\lambda_{j}}(L_{1})\varepsilon^{-(3-j)}\lambda_{j} + \sum_{j=1}^{2}\gamma_{2}^{\xi_{j}}(L_{1})\varepsilon^{-(3-j)}\xi_{j} + \gamma_{2}^{y}(L_{1})\varepsilon^{-2}y + \varepsilon^{-1}\bar{\alpha}_{2}(L_{1})z_{2}\right) + d_{\boldsymbol{\lambda}}\varepsilon^{4}\lambda_{3}^{2} + \varepsilon^{2}z_{2}\lambda_{3}$$
(35)

where  $\bar{\alpha}_2(L_1) = 5d_{\mathbf{\lambda}} + 1/4(d_0^2 + c'^2)(\gamma_2^{\Delta}(L_1))^2 + b_M g_M |\gamma_2^{\Delta}(L_1)| + 2(\gamma_2^{\Delta}(L_1))^2 + 1/4(b_M^2 g_M^2 \alpha_1^2(L_1)(\gamma_2^{\Delta}(L_1))^2 + \max\{1, b_M^2 g_M^2\}).$ Thus, we can choose the virtual controller as

$$\lambda_{3}^{*} = -\varepsilon^{-1} \alpha_{2}(\boldsymbol{L}_{[2]}) z_{2} - \sum_{j=1}^{2} \gamma_{2}^{\lambda_{j}}(L_{1}) \varepsilon^{-(3-j)} \lambda_{j} - \sum_{j=1}^{2} \gamma_{2}^{\xi_{j}}(L_{1}) \varepsilon^{-(3-j)} \xi_{j} - \gamma_{2}^{y}(L_{1}) \varepsilon^{-2} y$$
(36)

where  $\alpha_2(\boldsymbol{L}_{[2]}) = L_2 + \bar{\alpha}_2(L_1)$  and  $L_2 \ge 1$  is a constant to be determined later.

Define  $z_3 = \lambda_3 - \lambda_3^*$ . Similar to Step 1, we should estimate  $d_{\boldsymbol{\lambda}}\varepsilon^4\lambda_3^2$  by an appropriate function of  $z_1, z_2, z_3, \boldsymbol{\xi}$ , and  $\tilde{\boldsymbol{\chi}}$ . Since  $\lambda_1 = 1/b'_m(y - \tilde{\chi}_1 - \xi_1)$ , we have  $\lambda_1^2 \leq 3/(b_N^2 g_N^2)(y^2 + \tilde{\chi}_1^2 + \xi_1^2)$ . By this and (29), we have

$$d_{\boldsymbol{\lambda}}\varepsilon^{4}\lambda_{3}^{2} \leq 7d_{\boldsymbol{\lambda}}\varepsilon^{4}z_{3}^{2} + \sum_{j=1}^{2}\mu_{2}^{z_{j}}(\boldsymbol{L}_{[2]})\varepsilon^{2j-2}z_{j}^{2} + \mu_{2}^{\tilde{\boldsymbol{\chi}}}(L_{1})\|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} + \mu_{2}^{\boldsymbol{\xi}}(L_{1})\|I_{\varepsilon}\boldsymbol{\xi}\|^{2}$$

$$z_{1}(\boldsymbol{I}_{\varepsilon}) = \frac{21d_{\boldsymbol{\lambda}}}{\varepsilon}(\lambda_{1})(\boldsymbol{I}_{\varepsilon})z_{\varepsilon}^{2} + \overline{z}(\lambda_{2})(\boldsymbol{I}_{\varepsilon})z_{\varepsilon}^{2} - \overline{z}(\lambda_{2})($$

where  $\mu_2^{z_1}(\mathbf{L}_{[2]}) = \frac{21a_{\mathbf{\lambda}}}{b_N^2 g_N^2} (\gamma_2^{\lambda_1}(L_1))^2 + 7(\gamma_2^{\lambda_2}(L_1))^2 \mu_1^{z_1}(L_1) + 7d_{\mathbf{\lambda}}(\gamma_2^y(L_1))^2, \ \mu_2^{z_2}(\mathbf{L}_{[2]}) = 7d_{\mathbf{\lambda}}\alpha_2^2(\mathbf{L}_{[2]}) + 35d_{\mathbf{\lambda}}(\gamma_2^{\lambda_2}(L_1))^2, \ \mu_2^{\mathbf{\tilde{\chi}}}(L_1) = \frac{21d_{\mathbf{\lambda}}}{b_N^2 g_N^2} (\gamma_2^{\lambda_1}(L_1))^2 + 7(\gamma_2^{\lambda_2}(L_1))^2 \mu_1^{\mathbf{\tilde{\chi}}}, \ \mu_2^{\mathbf{\xi}}(L_1) = \frac{21d_{\mathbf{\lambda}}}{b_N^2 g_N^2} (\gamma_2^{\lambda_1}(L_1))^2 + 7(\gamma_2^{\lambda_2}(L_1))^2 \mu_1^{\mathbf{\xi}} + \max\{7d_{\mathbf{\lambda}}(\gamma_2^{\xi_j}(L_1))^2, j = 1, 2\}.$ 

By substituting this and (36) into (35), we have

$$\begin{split} \dot{V}_{2} &\leq -h_{\eta}(\varepsilon) \|\boldsymbol{\eta}\|^{2} - h_{2,1}(\varepsilon, L_{1}) \|I_{\varepsilon} \tilde{\boldsymbol{\chi}}\|^{2} - h_{2,2}(\varepsilon, L_{1}) \times \\ \|I_{\varepsilon} \boldsymbol{\xi}\|^{2} - h_{2,3}(\varepsilon, \boldsymbol{L}_{[2]}) z_{1}^{2} - h_{2,4}(\varepsilon, \boldsymbol{L}_{[2]}) \varepsilon^{2} z_{2}^{2} + \\ d_{\boldsymbol{\lambda}} \sum_{i=4}^{\rho} \varepsilon^{2i-2} \lambda_{i}^{2} + 7 d_{\boldsymbol{\lambda}} \varepsilon^{4} z_{3}^{2} + \varepsilon^{2} z_{2} z_{3} \end{split}$$
(37)

with  $d_{2,1}(L_1) = d_{1,1} + \mu_2^{\tilde{\mathbf{X}}}(L_1), d_{2,2}(L_1) = d_{1,2} + \mu_2^{\boldsymbol{\xi}}(L_1), d_{2,3}(\boldsymbol{L}_{[2]}) = d_{1,3}(L_1) + 2 + \mu_2^{z_1}(\boldsymbol{L}_{[2]}), \text{ and } d_{2,4}(\boldsymbol{L}_{[2]}) = \mu_2^{z_2}(\boldsymbol{L}_{[2]}).$ 

Inductive Step. Suppose at step k - 1 ( $k = 3, \dots, \rho$ ), there exists a smooth, positive definite, and proper function  $V_{k-1}(\boldsymbol{\eta}, \boldsymbol{\tilde{\chi}}, \boldsymbol{\xi}, z_1, \dots, z_{k-1})$  whose time derivative satisfies

$$V_{k-1} \leq -h_{\eta}(\varepsilon) \|\boldsymbol{\eta}\|^{2} - h_{k-1,1}(\varepsilon, \boldsymbol{L}_{[k-2]}) \|I_{\varepsilon} \tilde{\boldsymbol{\chi}}\|^{2} - h_{k-1,2}(\varepsilon, \boldsymbol{L}_{[k-2]}) \|I_{\varepsilon} \boldsymbol{\xi}\|^{2} - \sum_{j=1}^{k-1} h_{k-1,j+2}(\varepsilon, \boldsymbol{L}_{[k-1]}) \times \varepsilon^{2j-2} z_{j}^{2} + d_{\boldsymbol{\lambda}} \sum_{i=k+1}^{\rho} \varepsilon^{2i-2} \lambda_{i}^{2} + (2k+1) d_{\boldsymbol{\lambda}} \varepsilon^{2k-2} z_{k}^{2} + \varepsilon^{2k-4} z_{k-1} z_{k}$$
(38)

where  $z_1 = y, z_i = \lambda_i - \lambda_i^*, i = 2, \dots, k$ , and  $\lambda_i^*$  are a set of virtual controllers in the following form:

$$\lambda_{i}^{*} = -\varepsilon^{-1} \alpha_{i-1} (\boldsymbol{L}_{[i-1]}) z_{i-1} - \sum_{j=1}^{i-1} \gamma_{i-1}^{\lambda_{j}} (\boldsymbol{L}_{[i-2]}) \varepsilon^{-(i-j)} \lambda_{j} - \sum_{j=1}^{i-1} \gamma_{i-1}^{\xi_{j}} (\boldsymbol{L}_{[i-2]}) \varepsilon^{-(i-j)} \xi_{j} - \gamma_{i-1}^{y} (L_{1}) \varepsilon^{-(i-1)} y \quad (39)$$

From (39), we can find nonnegative functions  $\mu_{i-1}^{z_j}(\cdot)$ ,  $\mu_{i-1}^{\tilde{\mathbf{x}}}(\cdot)$ ,  $\mu_{i-1}^{\boldsymbol{\xi}}(\cdot)$ ,  $i = 2, \cdots, k, j = 1, \cdots, i-1$ , such that

$$d_{\lambda}\varepsilon^{2i-2}\lambda_{i}^{2} \leq (2i+1)d_{\lambda}\varepsilon^{2i-2}z_{i}^{2} + \sum_{j=1}^{i-1}\mu_{i-1}^{z_{j}}(\boldsymbol{L}_{[i-1]})\varepsilon^{2j-2}z_{j}^{2} + \mu_{i-1}^{\tilde{\boldsymbol{\chi}}}(\boldsymbol{L}_{[i-2]})\|I_{\varepsilon}\boldsymbol{\tilde{\chi}}\|^{2} + \mu_{i-1}^{\boldsymbol{\xi}}(\boldsymbol{L}_{[i-2]})\|I_{\varepsilon}\boldsymbol{\xi}\|^{2} \quad (40)$$

The dynamics of variable  $z_i (i = 2, \dots, k-1)$  can be immediately computed from (39) as

$$\dot{z}_{i} = \lambda_{i+1} + \sum_{j=1}^{i} \gamma_{i}^{\lambda_{j}} (\boldsymbol{L}_{[i-1]}) \varepsilon^{-(i+1-j)} \lambda_{j} + \sum_{j=1}^{i} \gamma_{i}^{\xi_{j}} (\boldsymbol{L}_{[i-1]}) \varepsilon^{-(i+1-j)} \xi_{j} + \gamma_{i}^{y} (L_{1}) \varepsilon^{-i} y + \gamma_{i}^{\Delta} (\boldsymbol{L}_{[i-1]}) \varepsilon^{-(i-1)} (c_{\rho,1} y + b'_{m} \lambda_{2} + \tilde{\chi}_{2} + \Phi_{\rho,1})$$

In what follows, we will show that the above statements still hold at step k. For this aim, choose  $V_k = V_{k-1} + (1/2)\varepsilon^{2k-2}z_k^2$ , where  $z_k = \lambda_k - \lambda_k^*$  and the virtual controller  $\lambda_k^*$  are smooth functions. For notational convenience and consistency, let  $\lambda_{\rho+1} = u$ . Then, computing the time derivative of  $V_k$ , we have

$$\dot{V}_k = \dot{V}_{k-1} + \varepsilon^{2k-2} z_k \dot{z}_k \tag{41}$$

By the definition of  $z_k$ , we have

$$\dot{z}_{k} = \lambda_{k+1} + \sum_{j=1}^{k} \gamma_{k}^{\lambda_{j}} (\boldsymbol{L}_{[k-1]}) \varepsilon^{-(k+1-j)} \lambda_{j} + \sum_{j=1}^{k} \gamma_{k}^{\xi_{j}} (\boldsymbol{L}_{[k-1]}) \varepsilon^{-(k+1-j)} \xi_{j} + \gamma_{k}^{y} (L_{1}) \varepsilon^{-k} y + \gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}) \varepsilon^{-(k-1)} (c_{\rho,1}y + b'_{m} \lambda_{2} + \tilde{\chi}_{2} + \Phi_{\rho,1})$$
(42)

where  $\gamma_k^{\lambda_j}(\boldsymbol{L}_{[k-1]}), \ \gamma_k^{\xi_j}(\boldsymbol{L}_{[k-1]}), \ \gamma_k^y(L_1), \ \text{and} \ \gamma_k^{\Delta}(\boldsymbol{L}_{[k-1]}), \ j = 1, \cdots, k \text{ are defined as}$ 

$$\begin{split} \gamma_{k}^{\lambda_{1}}(\boldsymbol{L}_{[k-1]}) &= l_{k} + \sum_{j=1}^{k-1} l_{j} \gamma_{k-1}^{\lambda_{j}}(\boldsymbol{L}_{[k-2]}) + \alpha_{k-1}(\boldsymbol{L}_{[k-1]}) \gamma_{k-1}^{\lambda_{1}}(\boldsymbol{L}_{[k-2]}) \\ \gamma_{k}^{\lambda_{j}}(\boldsymbol{L}_{[k-1]}) &= \alpha_{k-1}(\boldsymbol{L}_{[k-1]}) \gamma_{k-1}^{\lambda_{j}}(\boldsymbol{L}_{[k-2]}) + \gamma_{k-1}^{\lambda_{j-1}}(\boldsymbol{L}_{[k-2]}), \\ j &= 2, \cdots, k-1 \\ \gamma_{k}^{\lambda_{k}}(\boldsymbol{L}_{[k-1]}) &= \gamma_{k-1}^{\lambda_{k-1}}(\boldsymbol{L}_{[k-2]}) + \alpha_{k-1}(\boldsymbol{L}_{[k-1]}) \\ \gamma_{k}^{\xi_{1}}(\boldsymbol{L}_{[k-1]}) &= \sum_{j=1}^{k-1} l_{j} \gamma_{k-1}^{\xi_{j}}(\boldsymbol{L}_{[k-2]}) + \alpha_{k-1}(\boldsymbol{L}_{[k-1]}) \gamma_{k-1}^{\xi_{1}}(\boldsymbol{L}_{[k-2]}) \\ \gamma_{k}^{\xi_{2}}(\boldsymbol{L}_{[k-1]}) &= \alpha_{k-1}(\boldsymbol{L}_{[k-1]}) \gamma_{k-1}^{\xi_{2}}(\boldsymbol{L}_{[k-2]}) + \gamma_{k-1}^{y}(\boldsymbol{L}_{1}) + \gamma_{k-1}^{\xi_{1}}(\boldsymbol{L}_{[k-2]}) \\ \gamma_{k}^{\xi_{j}}(\boldsymbol{L}_{[k-1]}) &= \alpha_{k-1}(\boldsymbol{L}_{[k-1]}) \gamma_{k-1}^{\xi_{j}}(\boldsymbol{L}_{[k-2]}) + \gamma_{k-1}^{\xi_{j-1}}(\boldsymbol{L}_{[k-2]}), \\ j &= 2, \cdots, k-1 \\ \gamma_{k}^{\xi_{k}}(\boldsymbol{L}_{[k-1]}) &= \gamma_{k-1}^{\xi_{k-1}}(\boldsymbol{L}_{[k-2]}) \\ \gamma_{k}^{y}(\boldsymbol{L}_{1}) &= \alpha_{k-1}(\boldsymbol{L}_{[k-1]}) \gamma_{k-1}^{y}(\boldsymbol{L}_{1}) - \sum_{j=1}^{k-1} l_{j} \gamma_{k-1}^{\xi_{j}}(\boldsymbol{L}_{[k-2]}) \end{split}$$

$$\gamma_{k}^{\Delta}(\boldsymbol{L}_{[k-1]}) = \alpha_{k-1}(\boldsymbol{L}_{[k-1]})\gamma_{k-1}^{\Delta}(\boldsymbol{L}_{[k-2]}) + \gamma_{k-1}^{y}(L_{1})$$

From the above equations, we can see that  $\gamma_k^{\lambda_j}(\cdot)$ ,  $\gamma_k^{\xi_j}(\cdot)$ ,  $\gamma_k^{\xi_j}(\cdot)$ ,  $\gamma_k^y(\cdot)$ , and  $\gamma_k^{\Delta}(\cdot)$ ,  $j = 1, \cdots, k$  are continuous functions

which can be derived recursively from their initial assignment in Step 2, and become constant once the value of  $L_{[k-1]}$  is specified.

Before deriving the virtual controller  $\lambda_{k+1}^*$ , we should eliminate the "undesired" effect of  $z_k$  in (41). For this purpose, by the method of completing square, we have

$$\varepsilon^{2k-4} z_{k-1} z_k \le \varepsilon^{2k-5} z_{k-1}^2 + \frac{1}{4} \varepsilon^{2k-3} z_k^2 \le \varepsilon^{2k-5} z_{k-1}^2 + \frac{1}{4} \max\{1, b_M^2 g_M^2\} \varepsilon^{2k-3} z_k^2$$
(43)

and the following inequalities

$$\begin{cases} \varepsilon^{k-1} \gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}) c_{\rho,1} y z_{k} \leq \frac{1}{4} d_{0}^{2} (\gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}))^{2} \varepsilon^{2k-2} z_{k}^{2} + y^{2} \\ b_{m}^{\prime} \varepsilon^{k-1} \gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}) \lambda_{2} z_{k} \leq \frac{1}{4} b_{M}^{2} g_{M}^{2} (\gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}))^{2} \varepsilon^{2k-3} z_{k}^{2} + \\ \varepsilon z_{2}^{2} + \frac{1}{\varepsilon} y^{2} + \frac{1}{4} b_{M}^{2} g_{M}^{2} \alpha_{1}^{2} (L_{1}) \times \\ (\gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}))^{2} \varepsilon^{2k-3} z_{k}^{2} \\ \varepsilon^{k-1} \gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}) \tilde{\chi}_{2} z_{k} \leq 2^{k-1} (\gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}))^{2} \varepsilon^{2k-3} z_{k}^{2} + \\ \frac{1}{2^{k+1} \varepsilon} \| I_{\varepsilon} \tilde{\boldsymbol{\chi}} \|^{2} \\ \varepsilon^{k-1} \gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}) \Phi_{\rho,1} z_{k} \leq \frac{1}{4} c^{\prime 2} (\gamma_{k}^{\Delta} (\boldsymbol{L}_{[k-1]}))^{2} \varepsilon^{2k-2} z_{k}^{2} + y^{2} \\ \end{cases}$$

$$(44)$$

Choose the virtual controller as

$$\lambda_{k+1}^{*} = -\varepsilon^{-1} \alpha_{k} (\boldsymbol{L}_{[k]}) z_{k} - \sum_{j=1}^{k} \gamma_{k}^{\lambda_{j}} (\boldsymbol{L}_{[k-1]}) \varepsilon^{-(k+1-j)} \lambda_{j} - \sum_{j=1}^{k} \gamma_{k}^{\xi_{j}} (\boldsymbol{L}_{[k-1]}) \varepsilon^{-(k+1-j)} \xi_{j} - \gamma_{k}^{y} (L_{1}) \varepsilon^{-k} y$$
(45)

where  $\alpha_k(\mathbf{L}_{[k]}) = L_k + \bar{\alpha}_k(\mathbf{L}_{[k-1]}), \quad \bar{\alpha}_k(\mathbf{L}_{[k-1]}) = (2k+1)d_{\mathbf{\lambda}} + (1/4)\max\{1, b_M^2 g_M^2\} + 2^{k-1}(\gamma_k^{\Delta}(\mathbf{L}_{[k-1]}))^2 + (1/4)(d_0^2 + c'^2)(\gamma_k^{\Delta}(\mathbf{L}_{[k-1]}))^2 + (1/4)b_M^2 g_M^2(\alpha_1^2(L_1) + 1)(\gamma_k^{\Delta}(\mathbf{L}_{[k-1]}))^2, \text{ and } L_k \geq 1 \text{ is a constant to be determined later. Note that once the value of } \mathbf{L}_{[k]} \text{ is specified,} \\ \alpha_k(\mathbf{L}_{[k]}) \text{ is a known positive constant.}$ 

Define  $z_{k+1} = \lambda_{k+1} - \lambda_{k+1}^*$  when  $k < \rho$  and  $z_{\rho+1} = 0$ . Then, similarly, we have

$$d_{\lambda}\varepsilon^{2k}\lambda_{k+1}^{2} \leq (2k+3)d_{\lambda}\varepsilon^{2k}z_{k+1}^{2} + (2k+3)d_{\lambda}(\gamma_{k}^{y}(L_{1}))^{2}y^{2} + (2k+3)d_{\lambda}\alpha_{k}^{2}(\boldsymbol{L}_{[k]})\varepsilon^{2k-2}z_{k}^{2} + (2k+3)d_{\lambda}\times \max\{(\gamma_{k}^{\xi_{j}}(\boldsymbol{L}_{[k-1]}))^{2}, j = 1, \cdots, k\}\|I_{\varepsilon}\boldsymbol{\xi}\|^{2} + (2k+3)d_{\lambda}\sum_{j=1}^{k}(\gamma_{k}^{\lambda_{j}}(\boldsymbol{L}_{[k-1]}))^{2}\varepsilon^{2j-2}\lambda_{j}^{2}$$
(46)

Clearly, the last term on the right-hand side of the above inequality is undesirable and has to be handled. By (40), we have the following inequality

$$d_{\lambda}(\gamma_{k}^{\lambda_{j}}(\boldsymbol{L}_{[k-1]}))^{2}\varepsilon^{2j-2}\lambda_{j}^{2} \leq (\gamma_{k}^{\lambda_{j}}(\boldsymbol{L}_{[k-1]}))^{2}(2j+1)d_{\lambda}\varepsilon^{2j-2}z_{j}^{2} + (\gamma_{k}^{\lambda_{j}}(\boldsymbol{L}_{[k-1]}))^{2}\sum_{i=1}^{j-1}\mu_{j-1}^{z_{i}}(\boldsymbol{L}_{[j-1]})\varepsilon^{2i-2}z_{i}^{2} + (\gamma_{k}^{\lambda_{j}}(\boldsymbol{L}_{[k-1]}))^{2}\mu_{j-1}^{\tilde{\boldsymbol{\chi}}}(\boldsymbol{L}_{[i-2]})\|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} + (\gamma_{k}^{\lambda_{j}}(\boldsymbol{L}_{[k-1]}))^{2}\mu_{j-1}^{\boldsymbol{\xi}}(\boldsymbol{L}_{[i-2]})\|I_{\varepsilon}\boldsymbol{\xi}\|^{2}$$
(47)

Substituting this into (46), we have for  $k < \rho$ 

$$d_{\boldsymbol{\lambda}}\varepsilon^{2k}\lambda_{k+1}^{2} \leq (2k+3)d_{\boldsymbol{\lambda}}\varepsilon^{2k}z_{k+1}^{2} + \sum_{j=1}^{k}\mu_{k}^{z_{j}}(\boldsymbol{L}_{[k]})\varepsilon^{2j-2}z_{j}^{2} + \mu_{k}^{\boldsymbol{\tilde{\chi}}}(\boldsymbol{L}_{[k-1]})\|I_{\varepsilon}\boldsymbol{\tilde{\chi}}\|^{2} + \mu_{k}^{\boldsymbol{\xi}}(\boldsymbol{L}_{[k-1]})\|I_{\varepsilon}\boldsymbol{\xi}\|^{2}$$
(48)

Specially, one can see that since  $d_{\lambda} \sum_{i=\rho+1}^{\rho} \varepsilon^{2i-2} \lambda_i^2 = 0$ , it is unnecessary for Step  $\rho$  to take the computation similar to (48). However, for the sake of the integrality of the inductive steps, we let  $\mu_{\rho}^{z_j}(\boldsymbol{L}_{[\rho]}) = 0, \ j = 1, \cdots, \rho,$  $\mu_{\rho}^{\tilde{\boldsymbol{\chi}}}(\boldsymbol{L}_{[\rho-1]}) = 0, \text{ and } \mu_{\rho}^{\boldsymbol{\xi}}(\boldsymbol{L}_{[\rho-1]}) = 0.$ 

Substituting  $(42) \sim (46)$  into (41), we obtain

$$\dot{\mathcal{V}}_{k} \leq -h_{\eta}(\varepsilon) \|\boldsymbol{\eta}\|^{2} - h_{k,1}(\varepsilon, \boldsymbol{L}_{[k-1]}) \| I_{\varepsilon} \tilde{\boldsymbol{\chi}} \|^{2} - h_{k,2}(\varepsilon, \boldsymbol{L}_{[k-1]}) \times \| I_{\varepsilon} \boldsymbol{\xi} \|^{2} - \sum_{i=1}^{k} h_{k,i+2}(\varepsilon, \boldsymbol{L}_{[k]}) \varepsilon^{2i-2} z_{i}^{2} + d_{\boldsymbol{\lambda}} \sum_{i=k+2}^{\rho} \varepsilon^{2i-2} \times \lambda_{i}^{2} + (2k+3) d_{\boldsymbol{\lambda}} \varepsilon^{2k} z_{k+1}^{2} + \varepsilon^{2k-2} z_{k} z_{k+1}$$

$$(49)$$

with

$$\begin{cases} d_{k,1}(\boldsymbol{L}_{[k-1]}) = d_{k-1,1}(\boldsymbol{L}_{[k-2]}) + \mu_{k}^{\tilde{\boldsymbol{\chi}}}(\boldsymbol{L}_{[k-1]}) \\ d_{k,2}(\boldsymbol{L}_{[k-1]}) = d_{k-1,2}(\boldsymbol{L}_{[k-2]}) + \mu_{k}^{\boldsymbol{\xi}}(\boldsymbol{L}_{[k-1]}) \\ d_{k,3}(\boldsymbol{L}_{[k]}) = d_{k-1,3}(\boldsymbol{L}_{[k-1]}) + 2 + \mu_{k}^{z_{1}}(\boldsymbol{L}_{[k]}) \\ d_{k,4}(\boldsymbol{L}_{[k]}) = d_{k-1,4}(\boldsymbol{L}_{[k-1]}) + \mu_{k}^{z_{2}}(\boldsymbol{L}_{[k]}) \\ d_{k,i+2}(\boldsymbol{L}_{[k]}) = d_{k-1,i+2}(\boldsymbol{L}_{[k-1]}) + \mu_{k}^{z_{i}}(\boldsymbol{L}_{[k]}), \\ i = 3, \cdots, k-1 \\ d_{k,k+2}(\boldsymbol{L}_{[k]}) = \mu_{k}^{z_{k}}(\boldsymbol{L}_{[k]}) \end{cases}$$

From the above equations, we can see that  $d_{k,i}(\cdot), k = 3, \cdots, \rho, i = 1, \cdots, k + 2$  are continuous functions which can be derived recursively from their initial assignment in Step 2, and are undoubtedly constant once the value of  $\boldsymbol{L}_{[k]}$  is specified. By more detailed analysis, we can see that they are all positive except  $d_{\rho,\rho+2}(\boldsymbol{L}_{[\rho]}) = 0$ .

At the last step, the design of u is quite similar to that of other steps, since all the terms containing the factors,  $\lambda_i^2, i = 1, \dots, \rho$  in (49), have already been canceled at Step  $\rho - 1$ . Using the inductive procedure and letting  $k = \rho$  in (45), we can design the controller as

$$u = \lambda_{\rho+1} = \lambda_{\rho+1}^* \tag{50}$$

Accordingly, letting  $k = \rho$  in (49) and noting that  $d_{\lambda} \sum_{i=\rho+1}^{\rho} \varepsilon^{2i-2} \lambda_i^2 = 0, \ z_{\rho+1} = 0$ , we have

$$\dot{V}_{\rho} \leq -h_{\eta}(\varepsilon) \|\boldsymbol{\eta}\|^{2} - h_{\rho,1}(\varepsilon, \boldsymbol{L}_{[\rho-1]}) \| I_{\varepsilon} \tilde{\boldsymbol{\chi}} \|^{2} - h_{\rho,2}(\varepsilon, \boldsymbol{L}_{[\rho-1]}) \| I_{\varepsilon} \boldsymbol{\xi} \|^{2} - \sum_{i=1}^{\rho} h_{\rho,i+2}(\varepsilon, \boldsymbol{L}_{[\rho]}) \varepsilon^{2i-2} z_{i}^{2}$$

$$\tag{51}$$

where  $V_{\rho}(\boldsymbol{\eta}, \tilde{\boldsymbol{\chi}}, \boldsymbol{\xi}, z_1, \cdots, z_{\rho})$  is a positive definite and proper function defined by

$$V_{\rho} = V_0 + \frac{1}{2} \sum_{i=1}^{\rho} \varepsilon^{2i-2} z_i^2$$
(52)

#### 3.3 Main results

From (51), we can see that to realize stabilization of the closed-loop system, the controller designed should ensure

the negative definiteness of  $\dot{V}_{\rho}$ . This can be guaranteed by choosing positive constants  $\varepsilon$  and  $L_i$  such that

$$\begin{cases} \frac{3}{8\varepsilon} - 2\|P_{\mathbf{l}}\|^{2} > 0\\ \frac{1}{2^{\rho+1}\varepsilon} - d_{\rho,1}(\boldsymbol{L}_{[\rho-1]}) > 0\\ \frac{1}{4\varepsilon} - d_{\rho,2}(\boldsymbol{L}_{[\rho-1]}) > 0\\ \frac{L_{i} - (\rho + 1 - i)}{\varepsilon} - d_{\rho,i+2}(\boldsymbol{L}_{[\rho]}) > 0\\ \frac{L_{j} - 1}{\varepsilon} - d_{\rho,j+2}(\boldsymbol{L}_{[\rho]}) > 0\\ \frac{L_{\rho}}{\varepsilon} - d_{\rho,\rho+2}(\boldsymbol{L}_{[\rho]}) > 0 \end{cases}$$
(53)

for i = 1, 2 and  $j = 3, \dots, \rho - 1$ .

The following lemma shows the existence of positive constants  $\varepsilon$  and  $L_i$  satisfying (53). Besides, the choice of such constants is discussed in the proof of the lemma.

**Lemma 3.** There are always positive constants  $L_i \geq 1, i = 1, \dots, \rho$  and  $0 < \varepsilon < 1$  satisfying inequalities (53).

**Proof.** We prove this lemma by construction. Firstly, for any specified constants  $L_i^* > 0, i = 1, \dots, \rho - 1$  and  $L_{\rho}^* \geq 1$ , choose

$$\begin{cases} L_i \ge L_i^* + \rho + 1 - i, & i = 1, 2\\ L_j \ge L_j^* + 1, & j = 3, \cdots, \rho - 1\\ L_\rho \ge L_\rho^* \end{cases}$$
(54)

Secondly, choose  $\varepsilon$  such that

$$0 < \varepsilon < \varepsilon^{*} = \min\left\{1, \frac{3}{16\|P_{i}\|^{2}}, \frac{1}{2^{\rho+1}d_{\rho,1}(\boldsymbol{L}_{[\rho-1]})}, \frac{1}{4d_{\rho,2}(\boldsymbol{L}_{[\rho-1]})}, \frac{L_{i}^{*}}{d_{\rho,i+2}(\boldsymbol{L}_{[\rho]})}, i = 1, \cdots, \rho\right\}$$
(55)

It is easy to verify that any  $L_i$  and  $\varepsilon$  determined by (54) and (55) satisfy inequalities (53).

The main result of the paper is summarized as follows. **Theorem 1.** Consider the output-feedback control problem of system (1) with  $\rho < n$ . Suppose the system satisfies Assumptions  $1 \sim 4$ . If  $\mathbf{l} = [l_1, \dots, l_\rho]^T$  is chosen such that matrix  $A_{\mathbf{l}} = A + \mathbf{l} \mathbf{e}_1^T$  is Hurwitz, then the closed-loop system is globally asymptotically stable under the dynamic output-feedback control (50) with positive constants  $\varepsilon$  and  $L_i$  satisfying (54) and (55).

**Proof.** First, by observing that  $\varepsilon$  and  $L_i$  satisfy (54) and (55), it is straightforward to deduce from (51) and (52) that there is a positive constant  $\beta$  such that  $\dot{V}_{\rho} \leq -\beta V_{\rho}$ , which implies that  $\eta$ ,  $\varepsilon^{i-1} \tilde{\chi}_i$ ,  $\varepsilon^{i-1} \xi_i$ , and  $\varepsilon^{2i-2} z_i$  are globally asymptotically stable for  $i = 1, \dots, \rho$ , and so are  $\eta$ ,  $\tilde{\chi}$ ,  $\xi$ , and  $z_i$  since  $\varepsilon$  is a positive constant. This together with the fact  $y = z_1 = \chi_1 = x_1$  concludes the global asymptotic stability of y,  $\chi_1$ , and  $x_1$ , and hence  $\lambda_1$  since  $\lambda_1 = 1/b'_m(y - \tilde{\chi}_1 - \xi_1)$ . Then, from (28) and  $\lambda_2 = z_2 + \lambda_2^*$ , it follows that  $\lambda_2^*$  and  $\lambda_2$  are globally asymptotically stable. Continuing in the same fashion, (39) and  $\lambda_i = z_i + \lambda_i^*$  for  $i = 3, \dots, \rho$  recursively establish that  $\lambda$  is globally asymptotically stable. By the global asymptotic stability of  $\xi$ ,  $\lambda$ , and  $\tilde{\chi}$ , and  $\hat{\chi} = \xi + b'_m \lambda$ , we know that  $\hat{\chi}$  and hence  $\chi (= \hat{\chi} + \tilde{\chi})$  are globally asymptotically stable as well.

Finally, from  $\boldsymbol{\varsigma}_{\rho,[1,\rho]} = \boldsymbol{\chi}, \boldsymbol{\varsigma}_{\rho,[\rho+1,n]} = \boldsymbol{\eta}$ , and the equivalent transformations defined above, we conclude that  $\boldsymbol{\varsigma}, \boldsymbol{x}$ , and  $\boldsymbol{\zeta}$  are all globally asymptotically stable.

**Remark 1.** It is easily seen that the design procedure given in the previous section cannot be unchangeably applied to the case  $\rho = n$ . In fact, there are minor technical differences between the control design procedures of the cases  $\rho = n$  and  $\rho < n$ , and hence the further investigation on the former case is omitted due to the limitation of space.

## 4 Concluding remarks

In this paper, the output-feedback stabilization has been investigated for a class of uncertain nonlinear systems. After introducing two kinds of linear state transformations, the control design becomes much natural since the converted system has known virtual control coefficients and separated stable zero-dynamics. For the new transformed system, an appropriate high-gain observer based on K-filters is introduced, and then by the backstepping approach, the output-feedback controller is successfully designed. It is shown that the global asymptotic stability of the closed-loop system can be guaranteed by the appropriate choice of the design parameters.

### Appendix

In this appendix, we give proofs of Propositions 2 and 3, respectively.

#### Appendix A Proof of Proposition 2

**Proof.** Along the trajectories of the subsystem  $\Sigma_{\eta}$ , the time derivative of  $V_{\eta}$  satisfies

$$\dot{V}_{\eta} \leq -\|\eta\|^{2} + 2\|Q\| \cdot \|\boldsymbol{c}_{\rho,[\rho+1,n]}\| \cdot \|\eta\| \cdot |\chi_{1}| + 2\|\eta\| \cdot \|Q\| \cdot \|\boldsymbol{\Phi}_{\rho,[\rho+1,n]}\|_{1}$$
(A1)

For the second term on the right-hand side of the above inequality, by Lemma 1 and Proposition 1, we have

$$2\|Q\| \cdot \|\boldsymbol{c}_{\rho,[\rho+1,n]}\| \cdot \|\boldsymbol{\eta}\| \cdot |\chi_1| \le \frac{1}{8}\|\boldsymbol{\eta}\|^2 + \frac{8n^2}{d^2}d_0'^2y^2 \qquad (A2)$$

By the fact  $0 < \varepsilon < 1$ , and Lemmas 1 and 2, we know that

$$2\|\boldsymbol{\eta}\| \cdot \|Q\| \cdot \|\boldsymbol{\Phi}_{\rho,[\rho+1,n]}\|_{1} \leq \frac{2mc'n}{d} \varepsilon^{-(\rho-1)} \|\boldsymbol{\eta}\| (|\chi_{1}| + |\varepsilon\chi_{2}| + \dots + |\varepsilon^{\rho-1}\chi_{\rho}|)$$

From this and  $\chi_1 = y$ ,  $\chi_i = \tilde{\chi}_i + \xi_i + b'_m \lambda_i$ ,  $i = 2, \dots, \rho$ , it follows that

$$2\|\boldsymbol{\eta}\| \cdot \|Q\| \cdot \|\boldsymbol{\Phi}_{\rho,[\rho+1,n]}\|_{1} \leq \frac{1}{2}\|\boldsymbol{\eta}\|^{2} + \frac{8m^{2}c'^{2}n^{2}}{d^{2}}\varepsilon^{-(2\rho-2)}y^{2} + \frac{8m^{2}c'^{2}n^{2}\rho}{d^{2}}\varepsilon^{-(2\rho-2)} \times \|I_{\varepsilon}\boldsymbol{\tilde{\chi}}\|^{2} + \frac{8m^{2}c'^{2}n^{2}\rho}{d^{2}}\varepsilon^{-(2\rho-2)}\|I_{\varepsilon}\boldsymbol{\xi}\|^{2} + \frac{8m^{2}c'^{2}n^{2}}{d^{2}}\rho b_{M}^{2}g_{M}^{2}\varepsilon^{-(2\rho-2)}\sum_{i=2}^{\rho}\varepsilon^{2i-2}\lambda_{i}^{2}$$
(A3)

Substituting (A2) and (A3) into (A1), we can get

$$\dot{V}_{\eta} \leq -\frac{3}{8} \|\eta\|^{2} + d_{0,1} (d_{0}^{\prime 2} + m^{2} c^{\prime 2} \varepsilon^{-(2\rho-2)}) y^{2} + \varepsilon^{-(2\rho-2)} d_{0,2} \|I_{\varepsilon} \tilde{\chi}\|^{2} + \varepsilon^{-(2\rho-2)} d_{0,2} \|I_{\varepsilon} \boldsymbol{\xi}\|^{2} + \varepsilon^{-(2\rho-2)} d_{0,3} \sum_{i=2}^{\rho} \varepsilon^{2i-2} \lambda_{i}^{2}$$
(A4)

where  $d_{0,1} = \frac{8n^2}{d^2}$ ,  $d_{0,2} = \frac{8m^2c'^2n^2\rho}{d^2}$ ,  $d_{0,3} = \frac{8m^2c'^2n^2}{d^2}\rho b_M^2 g_M^2$ . This leads to (21).

#### Appendix B Proof of Proposition 3

**Proof.** The time derivative of  $V_{\tilde{\mathbf{x}}}$  along (18) satisfies

$$\dot{V}_{\tilde{\mathbf{X}}} \leq -\frac{\|I_{\varepsilon}\tilde{\mathbf{X}}\|^{2}}{\varepsilon} + 2\varepsilon^{\rho-1}\|I_{\varepsilon}\tilde{\mathbf{X}}\| \cdot \|P_{l}\| \cdot \|\mathbf{\eta}\| + 2\|I_{\varepsilon}\tilde{\mathbf{X}}\| \times \|P_{l}\| \cdot \|\mathbf{c}_{\rho,[\rho]}\|\|y\| + 2\|I_{\varepsilon}\tilde{\mathbf{X}}\| \cdot \|P_{l}\| \cdot \|I_{\varepsilon}\mathbf{\Phi}_{\rho,[\rho]}\|_{1}$$
(B1)

We will deal with the last three terms on the right-hand side of the above inequality. Firstly, by Proposition 1, the second term and the third term satisfy, respectively,

$$2\varepsilon^{\rho-1} \|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\| \cdot \|P_{\mathbf{l}}\| \cdot \|\boldsymbol{\eta}\| \leq \frac{\|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2}}{2\varepsilon} + 2\varepsilon^{2\rho-1} \|P_{\mathbf{l}}\|^{2} \|\boldsymbol{\eta}\|^{2}$$
$$2\|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\| \cdot \|P_{\mathbf{l}}\| \cdot \|\boldsymbol{c}_{\rho,[\rho]}\| \cdot |y| \leq \|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} + d_{0}^{2}\|P_{\mathbf{l}}\|^{2}y^{2} \quad (B2)$$

For the last term, it is easy to show that  $||I_{\varepsilon} \mathbf{\Phi}_{\rho, [\rho]}||_1 \leq \rho c'(|\chi_1| + \varepsilon |\chi_2| + \cdots + \varepsilon^{\rho-1} |\chi_{\rho}|)$ . Then, noting that  $\chi_1 = y$  and  $\chi_i = \tilde{\chi}_i + \xi_i + b'_m \lambda_i, i = 2, \cdots, \rho$ , we have

$$2\|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\| \cdot \|P_{\mathbf{l}}\| \cdot \|I_{\varepsilon}\boldsymbol{\Phi}_{\rho,[\rho]}\|_{1} \leq (1+2\rho^{\frac{3}{2}}c'\|P_{\mathbf{l}}\|)\|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} + 3\rho^{3}c'^{2}\|P_{\mathbf{l}}\|^{2}\|I_{\varepsilon}\boldsymbol{\xi}\|^{2} + 3\rho^{2}c'^{2}\|P_{\mathbf{l}}\|^{2}y^{2} + 3\rho^{3}c'^{2}b_{M}^{2}g_{M}^{2}\|P_{\mathbf{l}}\|^{2}\sum_{i=2}^{\rho}\varepsilon^{2i-2}\lambda_{i}^{2} \quad (B3)$$

Substituting (B2) and (B3) into (B1), we have

$$\dot{V}_{\tilde{\boldsymbol{\chi}}} \leq -\left(\frac{1}{2\varepsilon} - d_{0,4}\right) \|I_{\varepsilon}\tilde{\boldsymbol{\chi}}\|^{2} + d_{0,5}\|I_{\varepsilon}\boldsymbol{\xi}\|^{2} + 2\varepsilon^{2\rho-1}\|P_{l}\|^{2}\|\boldsymbol{\eta}\|^{2} + d_{0,6}y^{2} + d_{0,7}\sum_{i=2}^{\rho}\varepsilon^{2i-2}\lambda_{i}^{2} \quad (B4)$$

where  $d_{0,4} = 2 + 2\rho^{\frac{3}{2}}c' ||P_{l}||, \ d_{0,5} = 3\rho^{3}c'^{2} ||P_{l}||^{2}, \ d_{0,6} = (d_{0}^{2} + 3\rho^{2}c'^{2}) ||P_{l}||^{2}, \ d_{0,7} = 3\rho^{3}c'^{2}b_{M}^{2}g_{M}^{2} ||P_{l}||^{2}$  are known positive constants independent of  $\varepsilon$ .

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