Research Article

Synthesis of Adaptive Gain Robust Output Feedback Controllers for a Class of Lipschitz Nonlinear Systems with Unknown Upper Bound of Uncertainty

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We propose a new adaptive gain robust output feedback controller for a class of the Lipschitz nonlinear systems with unknown upper bound of uncertainty. The proposed adaptive gain robust output feedback controller is designed so as to reduce the effect of uncertainties and Lipschitz nonlinearities. In this paper, we show that sufficient conditions for the existence of the proposed adaptive gain robust output feedback controller are reduced to LMI conditions. Finally, the effectiveness of the proposed robust output feedback controller is demonstrated by numerical simulations.

1. Introduction

Robustness of control systems to uncertainties has always been the central issue in feedback control, and therefore the problems of stability analysis and stabilization for uncertain systems have received much attention for a long time (e.g., [1, 2] and references therein). In particular, there are lots of existing results for state feedback robust control such as quadratic stabilizing control and \( H_\infty \) control (see [3, 4] and references therein). Besides, some design methods of variable gain robust state feedback controllers for uncertain systems have been suggested (e.g., [5–8]). Yamamoto and Yamauchi [5] proposed a design method of a robust controller with the ability to adjust control performances adaptively. In [6], an adaptive robust controller with adaptation mechanism has been presented and the adaptive robust controller is tuned on-line based on the information about parameter uncertainties. Besides, we have proposed robust controllers with adaptive compensation inputs [7, 8]. These controllers consist of a fixed gain controller and a variable gain one, and the variable gain controller is tuned by updating laws.

However, not all the states are measurable in practical systems because of technical, physical, and/or economic reasons. Therefore, the control strategies via observer-based robust controllers (e.g., [9, 10]) or robust output feedback one (e.g., [11–13]), which is of interest in this paper, have also been well studied. For robust output feedback controllers, Moheimani and Petersen [11] have presented a set of cross-coupled algebraic Riccati equations and algebraic Lyapunov equations. Geromel et al. [12] and Iwasaki et al. [13] adopted linear matrix inequality (LMI) approaches to design static output feedback controllers based on a set of the Lyapunov inequalities coupled by the constraint that one Lyapunov matrix is the inverse of another. Additionally the work of Matsuoka and Hagino [14] has presented an observer-based variable gain controller for a class of linear systems with uncertainties of which upper bounds are unknown, and we have also proposed an adaptive robust output feedback controller for a class of linear systems with uncertainties [15].

By the way, in recent years, there have been increasing attention for the problem of global stabilization of nonlinear systems via output feedback control (e.g., [16–18]). In the work of Mazenc et al. [16], it was presented through counter examples that some extra growth conditions on the unmeasurable states of the controlled system are usually necessary for the global stabilization of nonlinear systems.
via output feedback. Additionally, finite-time method and homogeneous domination approach for the output feedback control problem have also been proposed (e.g., [19–21]). Polendo and Qian [20] have suggested output feedback controllers for a class of uncertain nonlinear systems via homogeneous domination approach, and Li and Qian [21] adopted the concept of finite-time stabilization so as to design a dynamic output feedback controller for a class of continuous but nonsmooth nonlinear systems. Besides, some results have focused on considering a selective class of nonlinear systems by placing some structural constraints on the nonlinearities in order to derive output feedback control. The nonlinear systems whose nonlinearity is in a triangular form are considered in [22]. In the work of Choi and Lim [23], a solution to the output feedback stabilization problem for a class of single-input single-output Lipschitz nonlinear systems and the nonlinearity characterization function (NCF) concept has been presented. However, in the existing results, the design parameters are determined by trial and error, and uncertainties in the system dynamics have not been considered.

From these viewpoints on the basis of our results [15, 24], we propose a new adaptive gain robust output feedback controller for a class of uncertain Lipschitz nonlinear systems. The uncertainties and the nonlinearities under consideration are denoted by $A$, that is, the operator “vec” vectorizes a matrix by stacking its columns. The notation $\text{diag}(A_1, \ldots, A_N)$ denotes a block diagonal matrix composed of matrices $A_i$ for $i = 1, \ldots, N$. For real symmetric matrices $A$ and $B$, $A > B$ (resp., $A \geq B$) means that $A - B$ is positive (resp., nonnegative) definite matrix. For a vector $\alpha \in \mathbb{R}^n$, $|\alpha|$ denotes standard Euclidian norm, and for a matrix $A$, $\|A\|$ represents its induced norm. The symbols “$\Leftrightarrow$” and “$*$” mean equality by definition and symmetric blocks in matrix inequalities, respectively. Besides, for a symmetric matrix $P$, $\lambda_{\min}(P)$ and $\lambda_{\max}(P)$ represents the minimal eigenvalue (resp. maximal eigenvalue). It is well-known that for any symmetric matrix $P \in \mathbb{R}^{n \times n}$, eigenvalues of $P \in \mathbb{R}^{n \times n}$ are real number [25].

Furthermore, the following well-known lemmas are used in this paper.

**Lemma 1.** For arbitrary vectors $\lambda$ and $\xi$ and the matrices $\hat{\mathcal{G}}$ and $\mathcal{H}$ which have appropriate dimensions, the following relation holds:

$$
\mathcal{H} \{ \lambda^T \hat{\mathcal{G}} \Delta(t) \mathcal{H} \xi \} \leq 2 \| \hat{\mathcal{G}}^T \lambda \| \| \Delta(t) \| \| \mathcal{H} \xi \|
$$

where $\Delta(t) \in \mathbb{R}^{p \times q}$ is a time-varying unknown matrix satisfying $\| \Delta(t) \| \leq \vartheta^*$. 

Proof. The above relation can be easily obtained by Schwartz’s inequality [25].

**Lemma 2** (Schur complement). For a given constant real symmetric matrix $\Xi$, the following items are equivalent:

(i) $\Xi = \left( \begin{array}{cc} \Xi_{11} & \Xi_{12} \\ \Xi_{12}^T & \Xi_{22} \end{array} \right) > 0$,

(ii) $\Xi_{11} > 0$ and $\Xi_{22} - \Xi_{12} \Xi_{11}^{-1} \Xi_{12} > 0$,

(iii) $\Xi_{22} > 0$ and $\Xi_{11} - \Xi_{12} \Xi_{22}^{-1} \Xi_{12} > 0$.

Proof. See Boyd et al. [26].

**Lemma 3** ($\delta$-procedure). Let $\mathcal{F}(x)$ and $\mathcal{G}(x)$ be two arbitrary quadratic forms over $\mathbb{R}^n$. Then $\mathcal{F}(x) < 0$ for all $x \in \mathbb{R}^n$ satisfying $\mathcal{G}(x) \leq 0$ if and only if there exists a nonnegative scalar $\tau$ such that

$$
\mathcal{F}(x) - \tau \mathcal{G}(x) \leq 0 \quad \text{for all } x \in \mathbb{R}^n.
$$

Proof. See Boyd et al. [26].

**Lemma 4** (Barbalat’s lemma). Let $\phi : \mathbb{R} \rightarrow \mathbb{R}$ be a uniformly continuous function on $[0, \infty)$. Suppose that $\lim_{t \rightarrow \infty} \int_0^t \phi(t) \, dt$ exists and is finite. Then

$$
\phi(t) \rightarrow 0 \quad \text{as } t \rightarrow \infty.
$$

Proof. See Khalil [27].

**3. Problem Formulation**

Consider the uncertain Lipschitz nonlinear system described by the following state equation:

$$
\frac{dx(t)}{dt} = (A + B\Delta(t)E)x(t) + Bu(t) + \delta(x, t)
$$

$$
y(t) = Cx(t),
$$
where \( x(t) \in \mathbb{R}^n \), \( u(t) \in \mathbb{R}^m \), and \( y(t) \in \mathbb{R}^l \) are the vectors of the state, the control input, and the measured output, respectively. In (4), the matrices \( A, B, \) and \( C \) are the nominal values of system parameters, and the matrix \( \Delta(t) \in \mathbb{R}^{p \times q} \) denotes unknown time-varying parameters which satisfy
\[
\| \Delta(t) \| \leq \theta^* ,
\]
where the upper bound \( \theta^* \) is bounded, but it is unknown. Additionally in this paper, we assume that the nonlinear term \( \delta(x, t) \in \mathbb{R}^n \) in (4) is given by
\[
\delta(x, t) = B \xi(x, t),
\]
and for the function \( \xi: \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}^m \), there exists a known positive constant scalar \( \chi^* \) such that for all \( x_1, x_2 \in \mathbb{R}^n \)
\[
\| \xi(x_1, t) - \xi(x_2, t) \| \leq \chi^* \| x_1 - x_2 \| .
\]
Note that since not all the states are measurable, the nonlinear term \( \delta(x, t) \) is unknown. Besides, we introduce the following assumption for the system parameters [15, 24]:
\[
B^T = T C,
\]
where \( T \in \mathbb{R}^{m \times l} \) is a known constant matrix.

The nominal system, ignoring unknown parameters and nonlinearities in (4), is given by
\[
\begin{align*}
\dot{x}(t) &= Ax(t) + Bu(t) + \delta(x, t), \\
y(t) &= Cx(t),
\end{align*}
\]
and it is supposed to be stabilizable via static output feedback control. Namely, there exists a static output feedback stabilizing control \( u(t) = Ky(t) \) (i.e., a fixed gain matrix \( K \in \mathbb{R}^{m \times l} \)). In other words since the nominal system of (9) is stabilizable via static output feedback control, the matrix \( A_K = A + BK \) is asymptotically stable. Note that the output feedback gain matrix \( K \in \mathbb{R}^{m \times l} \) is designed by using the existing results (e.g., [28, 29]). Besides, in this paper, we consider the following target model so as to generate the desirable trajectory:
\[
\begin{align*}
\dot{x}_t(t) &= Ax_t(t) + Bu_t(t) + \delta(x_t, t), \\
y_t(t) &= Cx_t(t).
\end{align*}
\]
In order to generate the desirable trajectory for the uncertain system of (4), we select the control input for the target model such as \( u_t(t) = K_{LGA}x_t(t) - \xi(x_t, t) \) where \( K_{LGA} \in \mathbb{R}^{m \times n} \) is determined by adopting the standard LQG problem. Namely, by using the solution of the algebraic Riccati equation
\[
H_s[A^T X_t] - X_t BR_t B^T X_t + Q_t = 0,
\]
the gain matrix \( K_{LGA} \) is determined as \( K_{LGA} = -R_t^{-1} B^T X_t \). Of course, some other design methods can also be utilized. Note that the target model with the control input \( u_t(t) = K_{LGA}x_t(t) - \xi(x_t, t) \) can be written as the following form:
\[
\begin{align*}
\dot{x}_t(t) &= (A + BK_{LGA})x_t(t), \\
y_t(t) &= Cx_t(t),
\end{align*}
\]
Now on the basis of the works of Oya and Hagino ([15, 24]), by introducing the error vectors \( e(t) \triangleq x(t) - x_t(t) \) and \( \dot{e}_t(t) \triangleq \dot{x}(t) - \dot{x}_t(t) \), we consider the following control input for the uncertain Lipschitz nonlinear system of (4):
\[
\begin{align*}
u(t) &= K_e(t) + K_{LGA}x_t(t) - \xi(x_t, t) + \psi(y_t, x_t, \hat{\theta}, t),
\end{align*}
\]
In (12), \( \psi(y_t, x_t, \hat{\theta}, t) \in \mathbb{R}^m \) is an adaptive compensation input where \( \hat{\theta}(t) \in \mathbb{R}^p \) is an adjustable parameter. Then one can see from (4), (6), and (10)–(12) that the following uncertain error system with nonlinear terms can be derived:
\[
\begin{align*}
\dot{e}(t) &= AKe(t) + B\Delta(t)Ex(t) \\
&\quad + B(\xi(x_t, t) - \xi(x, t)) + B\psi(y, x, \hat{\theta}, t),
\end{align*}
\]
From the above, our control objective is to design the adaptive gain robust output feedback controller which achieves not only robust stability for the uncertain Lipschitz nonlinear system of (4) but also satisfactory transient behavior as closely as possible to desired trajectory generated by the target model. That is to derive the adaptive compensation input \( \psi(y_t, x_t, \hat{\theta}, t) \in \mathbb{R}^m \) which stabilizes the uncertain nonlinear error system of (13).

### 4. Main Results

In this section, we show an LMI-based design method of the adaptive gain robust output feedback controller for the uncertain Lipschitz nonlinear system of (4). The following theorem gives an LMI-based design method of an adaptive gain robust output feedback controller.

**Theorem 5.** Consider the uncertain nonlinear error system of (13) with the adaptive compensation input \( \psi(y_t, x_t, \hat{\theta}, t) \in \mathbb{R}^m \). If there exist symmetric positive definite matrices \( \delta \in \mathbb{R}^{n \times n}, \Xi \in \mathbb{R}^{l \times l}, \) and \( \Psi \in \mathbb{R}^{l \times l} \) and the positive scalars \( \gamma_1, \gamma_2, \) and \( \epsilon \) satisfying the LMIs:
\[
\begin{align*}
H_s[A_K] + \gamma_1 E^T E + \epsilon (x^*)^2 I_n &\leq -\Xi, \\
C^T \Xi C - H_s \left\{ C^T T C \right\} \leq 0, \\
\left[ \begin{array}{c}
-C^T \Psi^T C & \delta C^T T & \delta C^T C \\
* & -\gamma_1 I_m & 0 \\
* & * & -\gamma_2 I_m
\end{array} \right] \leq 0,
\end{align*}
\]
then by using the solution of the LMIs of (14), we consider the adaptive compensation input
\[
\psi(y_t, x_t, \hat{\theta}, t) = -\frac{1}{\| \Xi^{1/2} C \epsilon \|} \omega(y_t, x_t, \hat{\theta}, t) T e_t(t),
\]
where \( \omega(\varepsilon, x, \hat{\theta}, t) \) is the positive scalar function given by
\[
\omega(\varepsilon, x, \hat{\theta}, t) \equiv - \frac{\left| \Psi^{1/2} \varepsilon_p(t) \right|^{4} |\hat{\theta}|^{2} (t) + \sigma(t)}{\left| \Psi^{1/2} \varepsilon_p(t) \right|^{2} |\hat{\theta}| + \sigma(t)} + \frac{\left| \Psi^{1/2} \varepsilon_p(t) \right|^{4}}{\left| \Psi^{1/2} \varepsilon_p(t) \right|^{2} + \sigma(t)} + \frac{\gamma_{2}^{2} \left| \dot{E} \right|^{4}}{\gamma_{2} \left| \dot{E} \right|^{2} + \sigma(t)}.
\]

Besides, we introduce the following updating law for the adjustable parameter \( \hat{\theta}(t) \in \mathbb{R}^{1} \):
\[
\frac{d}{dt} \hat{\theta}(t) \equiv - \frac{1}{2\vartheta} \left\{ \sigma(t) \left( \hat{\theta}(t) - \theta^{*} \right) \right. \\
\left. - \left| \Psi^{1/2} \varepsilon_p(t) \right|^{2} + \sigma(t) \theta^{*} \right\}.
\]

Hereby asymptotical stability of the uncertain nonlinear error system of (13) is guaranteed. In (14), \( Q \in \mathbb{R}^{n \times n} \) is a symmetric positive definite matrix selected by designers, and \( \sigma(t) \in \mathbb{R}^{1} \) in (16) is any positive uniform continuous and bounded function which satisfies
\[
\int_{t_{0}}^{t} \sigma(\tau) d\tau \leq \sigma^{*} < \infty,
\]
where \( t_{0} \) and \( \sigma^{*} \) are an initial time and any positive constant, respectively.

Proof of Theorem 5. Firstly, we introduce the quadratic function
\[
\mathcal{V}(\varepsilon, \hat{\theta}, t) \equiv \varepsilon^{T}(t) \delta e(t) + \theta \left( \hat{\theta}(t) - \theta^{*} \right)^{2}.
\]

The time derivative of the quadratic function \( \mathcal{V}(\varepsilon, \hat{\theta}, t) \) can be written as
\[
\frac{d}{dt} \mathcal{V}(\varepsilon, \hat{\theta}, t) = \varepsilon^{T}(t) [H_{c} \{ \delta A_{K} \}] e(t) + H_{c} \left\{ \varepsilon^{T}(t) \delta B \Delta(t) \dot{E} \right\} + \theta \left( \hat{\theta}(t) - \theta^{*} \right) \frac{d}{dt} \hat{\theta}(t).
\]

Now, using Lemma 1 and the assumptions of (6) and (7), we can obtain the following relation for the time derivative of the quadratic function \( \mathcal{V}(\varepsilon, \hat{\theta}, t) \):
\[
\frac{d}{dt} \mathcal{V}(\varepsilon, \hat{\theta}, t) \leq \varepsilon^{T}(t) \left[ H_{c} \{ \delta A_{K} \} + \gamma_{1} E^{T} E + \varepsilon \left( \chi^{*} \right)^{2} I_{n} \right] e(t) + H_{c} \left\{ \varepsilon^{T}(t) \delta B \Delta(t) \dot{E} \right\} + \theta \left( \hat{\theta}(t) - \theta^{*} \right) \frac{d}{dt} \hat{\theta}(t).
\]

Notice the fact that, for any positive constant \( \mu \) and any vectors \( a \) and \( b \) with appropriate dimensions,
\[
2a^{T}b \leq \mu a^{T}a + \frac{1}{\mu} b^{T}b.
\]

Then some algebraic manipulations yield
\[
\frac{d}{dt} \mathcal{V}(\varepsilon, \hat{\theta}, t) \leq \varepsilon^{T}(t) \left[ H_{c} \{ \delta A_{K} \} + \gamma_{1} E^{T} E + \varepsilon \left( \chi^{*} \right)^{2} I_{n} \right] e(t) + H_{c} \left\{ \varepsilon^{T}(t) \delta B \Delta(t) \dot{E} \right\} + \theta \left( \hat{\theta}(t) - \theta^{*} \right) \frac{d}{dt} \hat{\theta}(t).
\]
Besides, we obtain the following inequality for the time derivative of the quadratic function $V(e, \hat{\theta}, t)$,
\[
\frac{d}{dt} V(e, \hat{\theta}, t) \leq e^T(t) \left[ H_c \{ \delta A_k \} + y_1 E + e (\chi^*)^T I_n \right] e(t) + H_c \{ e^T(t) \delta C^T \mathcal{T} \psi(e_y, x, \hat{\theta}, t) \} + \gamma_s e^T(t) C^T \psi_C e(t) + e^T(t) C^T \psi_C e(t) + y_2 x_1^T(t) E_x(t) + 2\phi(e_y(t) - \theta^*) \frac{d}{dt} \hat{\theta}(t),
\]
(24)

because, by using Lemma 2 (Schur complement), one can see that the third LMI of (14) can be written as
\[
- C^T \psi_C + \frac{1}{\gamma_1} \delta C^T \mathcal{T} \mathcal{T} C \delta + \frac{1}{\gamma_2} \delta C^T \mathcal{T} \mathcal{T} C \delta \leq 0,
\]
(25)

and fourth LMI of (14) is equivalent to the following matrix inequality:
\[
- C^T \psi_C + \frac{1}{\epsilon} \delta C^T \mathcal{T} \mathcal{T} C \delta \leq 0.
\]
(26)

Using the first LMI and the second one of LMIs of (14) and introducing the adaptive compensation input of (15) and (16) and the updating law of (17), we have
\[
\frac{d}{dt} V(e, \hat{\theta}, t) \leq - e^T(t) Q e(t) + \gamma_s e^T(t) C^T \psi_C e(t) + e^T(t) C^T \psi_C e(t) + y_2 x_1^T(t) E_x(t) + H_c \left\{ e^T(t) \delta C^T \mathcal{T} \times \left( - \frac{1}{\| \Xi^{1/2} C e(t) \|} \right) \right\} \times \omega(e_y, x, \hat{\theta}, t) \mathcal{T} e_y(t) \right\} + 2\phi(e_y(t) - \theta^*) \frac{d}{dt} \hat{\theta}(t) = - e^T(t) Q e(t) + \| \psi_{1/2} e_y(t) \|^2 \hat{\theta}(t) - \sigma(t) \left( \hat{\theta}(t) - \theta^* \right)^2 - \sigma(t) \gamma_s \left( \hat{\theta}(t) - \theta^* \right)
\]
\[
- \left\| \psi_{1/2} e_y(t) \right\|^2 \hat{\theta}(t) + \sigma(t)
\]
\[
+ \frac{\left\| \psi_{1/2} e_y(t) \right\|^2}{\gamma_2 \left\| E_x(t) \right\|^2 + \sigma(t)} \sigma(t) + \frac{y_2 \left\| E_x(t) \right\|^2}{\gamma_2 \left\| E_x(t) \right\|^2 + \sigma(t)} \sigma(t).
\]
(27)

In addition, utilizing the well-known inequality for any positive constants $\alpha$ and $\beta$,
\[
0 \leq \frac{\alpha \beta}{\alpha + \beta} \leq \alpha \quad \forall \alpha, \beta > 0,
\]
(28)

and some trivial manipulations give the relation
\[
\frac{d}{dt} V(e, \hat{\theta}, t) \leq - e^T(t) Q e(t) + 3\sigma(t) - \sigma(t) (\hat{\theta}(t) - \theta^*)^2 - \sigma(t) \gamma_s \left( \hat{\theta}(t) - \theta^* \right).
\]
(29)

Namely, we have the following inequality:
\[
\frac{d}{dt} V(e, \hat{\theta}, t) \leq - e^T(t) Q e(t) + \phi \sigma(t).
\]
(30)

Here we have used the well-known inequality of (22), and $\phi$ in (30) is a positive constant given by $\phi \triangleq 3 + (\gamma_s)^2$. By letting $\zeta^* \triangleq \min \{ \lambda_{\text{min}}(Q) \}$, one can see that the inequality of (30) can be rewritten as
\[
\frac{d}{dt} V(e, \hat{\theta}, t) \leq - \zeta^* \| e(t) \|^2 + \phi \sigma(t).
\]
(31)

On the other hand, letting $e_{\theta}(t) = (e^T(t) \hat{\theta}(t))^T$, we see from the definition of the quadratic function $V(e, \hat{\theta}, t)$ that there always exist two positive constants $\delta_{\text{min}}$ and $\delta_{\text{max}}$ such that, for any $t \geq t_0$,
\[
\xi^- (\| e_{\theta}(t) \|) \leq V(e, \hat{\theta}, t) \leq \xi^+ (\| e_{\theta}(t) \|)
\]
(32)

where $\xi^- (\| e_{\theta}(t) \|)$ and $\xi^+ (\| e_{\theta}(t) \|)$ are given by
\[
\xi^- (\| e_{\theta}(t) \|) \triangleq \delta_{\text{min}} \| e_{\theta}(t) \|^2,
\]
\[
\xi^+ (\| e_{\theta}(t) \|) \triangleq \delta_{\text{max}} \| e_{\theta}(t) \|^2.
\]
(33)

It is obvious that any solution $e(t; t_0, e(t_0))$ of the uncertain nonlinear error system of (13) is continuous. In addition, it follows from (31) and (32) that, for any $t \geq t_0$, the relations of $0 \leq \xi^- (\| e_{\theta}(t) \|) \leq V(e, \hat{\theta}, t)$ and the following inequality holds:
\[
V(e, \hat{\theta}, t) = V(e, \hat{\theta}, t_0) + \int_{t_0}^{t} \frac{d}{dt} V(e, \hat{\theta}, \tau) d\tau
\]
\[
\leq \xi^+ (\| e_{\theta}(t_0) \|) - \int_{t_0}^{t} \xi^- (\| e(\tau) \|) d\tau + \phi \int_{t_0}^{t} \sigma(\tau) d\tau.
\]
(34)

In (34), $\xi^+ (\| e(t) \|)$ is defined as
\[
\xi^+ (\| e(t) \|) \triangleq \zeta^* \| e(t) \|^2.
\]
(35)

Therefore, from (34) we can obtain the following two results. Firstly, taking the limit as $t$ approaches infinity on both sides of the inequality of (34), we have the inequality
\[
0 \leq \xi^+ (\| e_{\theta}(t_0) \|) - \lim_{t \to \infty} \int_{t_0}^{t} \xi^- (\| e(\tau) \|) d\tau + \phi \lim_{t \to \infty} \int_{t_0}^{t} \sigma(\tau) d\tau.
\]
(36)
Thus one can see from (18) and (36) that
\[
\lim_{t \to \infty} \int_{t_0}^{t} \xi^*(\|e(\tau)\|) d\tau \leq \xi^*(\|e(t_0)\|) + \varphi \sigma^*.
\] (37)
On the other hand, from (34), we obtain
\[
0 \leq \xi^*(\|e(t)\|) \leq \xi^*(\|e(t_0)\|) + \varphi \int_{t_0}^{t} \sigma(\tau) d\tau.
\] (38)
It follows from (18) and (38) that
\[
0 \leq \xi^*(\|e(t)\|) \leq \xi^*(\|e(t_0)\|) + \varphi \sigma^*.
\] (39)

The relation of (39) implies that \( e(t) \) is uniformly bounded. Since \( e(t) \) has been shown to be continuous, it follows that \( e(t) \) is uniformly continuous. Therefore, one can see that \( \xi^*(\|e(t)\|) \) is also uniformly continuous. Thus applying Lemma 4 (Barbalat’s lemma) to (37) yields
\[
\lim_{t \to \infty} \xi^*(\|e(t)\|) = \lim_{t \to \infty} \xi^*(e(t)) = 0.
\] (40)
Namely, asymptotical stability of the uncertain nonlinear error system of (13) is ensured. Thus the uncertain Lipschitz nonlinear system of (4) is also stable.

It follows that the result of the theorem is true. Thus the proof of Theorem 5 is completed.

Theorem 5 provides a sufficient condition for the existence of an adaptive gain robust output feedback controller for uncertain Lipschitz nonlinear system of (4). Next, we consider a special case. In this case, we deal with the uncertain Lipschitz nonlinear system described by
\[
\frac{d}{dt} x(t) = (A + B\Delta(t)C)x(t) + Bu(t) + \delta_B(x,t),
\] (41)
\[
y(t) = Cx(t).
\]
In (41) the nonlinear term \( \delta_B(x,t) \in \mathbb{R}^n \) satisfies
\[
\delta_B(x,t) = \delta^{-1}C^T \xi(x,t),
\] (42)
where the matrix \( \delta \in \mathbb{R}^{n \times n} \) is symmetric positive definite and it is a solution of the following LMIs:
\[
H_x \{ \delta A_K \} + \epsilon \left( \chi^* \right)^2 I_n \leq -Q,
\]
\[
C^T \Xi C - H_x \{ \delta C^T \mathcal{T} \mathcal{T}^T C \} \leq 0,
\] (43)
\[
\begin{pmatrix}
-\mathcal{T}^T \Psi \mathcal{T} & \delta C^T \mathcal{T} \\
* & -\epsilon \mathcal{I}_m
\end{pmatrix} \leq 0.
\]

In (43), \( Q \in \mathbb{R}^{n \times n} \) is a symmetric positive definite matrix selected by designers. Thus one can see from (6), (10)–(12), (42), and (43) that we have
\[
\frac{d}{dt} e(t) = A_K e(t) + B\Delta(t)C x(t)
\]
\[
+ \delta^{-1}C^T(\xi(x,t) - \xi(x_0,t)) + B\psi \left( e_y, x_0, \hat{\theta}, t \right),
\]
\[
e_y(t) = Ce(t).
\] (44)

Next theorem gives an LMI-based design method of an adaptive gain robust output feedback controller for this case.

**Theorem 6.** Consider the uncertain nonlinear error system of (44) with the adaptive compensation input \( \psi(e_y, x_0, \hat{\theta}, t) \in \mathbb{R}^m \).

If there exist symmetric positive definite matrices \( \Sigma \in \mathbb{R}^{n \times n} \), \( \Psi \in \mathbb{R}^{l \times l} \), and \( \Xi \in \mathbb{R}^{l \times l} \) and the positive scalars \( y \) and \( \epsilon \) satisfying the LMIs of (43), then by using the solution of the LMIs of (43), one considers the adaptive compensation input \( \psi(e_y, x_0, \hat{\theta}, t) \in \mathbb{R}^m \) described as
\[
\psi(e_y, x_0, \hat{\theta}, t) \triangleq -\frac{1}{\|(\mathcal{T}^T \Xi C t)\|^2} \omega B \left( e_y, x_0, \hat{\theta}, t \right) \mathcal{T} e_y(t),
\] (45)
\[
\omega B \left( e_y, x_0, \hat{\theta}, t \right) \triangleq \frac{\|\Psi \mathcal{T}^T C t \|^4}{\|\mathcal{T}^T \Xi C t \|^2} \| \hat{\theta}(t) + \sigma(t) \|
\]
\[
+ \frac{\|\mathcal{T}^T \Xi C t \|^4}{\|\mathcal{T}^T \Xi C t \|^2} \| \hat{\theta}(t) + \sigma(t) \|
\]
\[
+ \frac{\| (e^{-1})e_y(t) \|^4}{\| (e^{-1})e_y(t) \|^2 + \sigma(t)}
\] (46)
and the updating law of (17) for the adjustable parameter \( \hat{\theta}(t) \in \mathbb{R}^l \). Hereby asymptotical stability of the uncertain error system with nonlinear terms of (4) is guaranteed. In (46), \( \sigma(t) \in \mathbb{R}^l \) is any positive uniform continuous and bounded function satisfying (18).

**Proof of Theorem 6.** By using the symmetric positive definite matrix \( \delta \in \mathbb{R}^{n \times n} \), we consider the quadratic function \( \mathcal{V}(e, \hat{\theta}, t) \) of (19). Then using the assumptions of (6) and (42) we have
\[
\frac{d}{dt} \mathcal{V}(e, \hat{\theta}, t) = e^T(t) [H_x \{ \delta A_K \}] e(t)
\]
\[
+ H_x \left\{ e^T(t) \delta C^T \mathcal{T} \mathcal{T}^T \Delta(t) C x(t) \right\}
\]
\[
+ e^T(t) \delta C^T \mathcal{T} \mathcal{T}^T \psi (e_y, x_0, \hat{\theta}, t) \right\} + 28 \left( \hat{\theta}(t) - \theta^* \right) \frac{d}{dt} \hat{\theta}(t)
\]
\[
\leq e^T(t) [H_x \{ \delta A_K \}] e(t)
\]
\[
+ H_x \left\{ e^T(t) \delta C^T \mathcal{T} \mathcal{T}^T \Delta(t) C x(t) \right\}
\]
\[
+ 2\chi^* \| C e(t) \| \| e(t) \|
\]
\[
+ H_x \left\{ e^T(t) \delta C^T \mathcal{T} \mathcal{T}^T \psi (e_y, x_0, \hat{\theta}, t) \right\}
\]
\[
+ 28 \left( \hat{\theta}(t) - \theta^* \right) \frac{d}{dt} \hat{\theta}(t).
\] (47)

In addition, applying the well-known inequality of (22) to the second term and the third one on the right-hand side of
(47), we obtain the following relation for the time derivative of the quadratic function $V(e, \hat{\theta}, t)$:

$$
\frac{d}{dt} V(e, \hat{\theta}, t) \leq e^T(t) \left[ H_e \{ A_k \} + \epsilon (\dot{\chi}^*)^2 I_n \right] e(t) + \frac{\dot{\chi}^*}{\gamma} e^T(t) C \dot{T} T^T C \dot{T} e(t) + \gamma y^T(t) y(t) + \frac{1}{\epsilon} e^T(t) e(t) + H_e \left\{ e^T(t) \delta C T^T T e(y, x_i, \hat{\theta}, t) \right\} + 2\dot{\theta}(\hat{\theta}(t) - \chi^*) \frac{d}{dt} \hat{\theta}(t).
$$

(48)

Now, one can see from (17), (43), (45), and (46) that if the following inequality holds, then the condition of (48) is also satisfied

$$
\frac{d}{dt} V(e, \hat{\theta}, t) \leq -e^T(t) Q e(t) + \frac{\chi^*}{\gamma} e^T(t) C^T \Psi \dot{T} C e(t) + \gamma y^T(t) y(t) + \frac{1}{\epsilon} e^T(t) e(t) + H_e \left\{ e^T(t) \delta C T^T T e(y, x_i, \hat{\theta}, t) \right\} + 2\dot{\theta}(\hat{\theta}(t) - \chi^*) \frac{d}{dt} \hat{\theta}(t).
$$

(49)

Besides, from (49) we obtain the inequality of (30). Therefore, one can see from Proof of Theorem 5 that the rest of proof of Theorem 6 is straightforward.

Remark 7. The proposed control scheme is adaptable when some assumptions are satisfied, and in cases where only the output signal of the system is available, the proposed scheme can be used widely. In addition, the proposed controller is more effective for systems with large uncertainties and Lipschitz constants. Note that the adjustable parameter $\hat{\theta}(t) \in \mathbb{R}^1$ is not an estimate of the unknown bound $\chi^*$.

5. Numerical Examples

In order to demonstrate the efficiency of the proposed control scheme, we have run a simple example. The control problem considered here is not necessary practical. However, the simulation results stated below illustrate the distinct feature of the proposed output feedback controller.

Consider the uncertain Lipschitz nonlinear system described by the following state equation

$$
\frac{d}{dt} x(t) = \begin{pmatrix} 2.0 & 0.0 & -6.0 \\ 0.0 & 1.0 & 1.0 \\ 3.0 & 0.0 & -7.0 \end{pmatrix} x(t) + \begin{pmatrix} 2.0 \\ 1.0 \\ 2.0 \end{pmatrix} \Delta(t) + \begin{pmatrix} 1.0 & 0.0 & 1.0 \\ 0.0 & 3.0 & 1.0 \end{pmatrix} x(t) + \begin{pmatrix} 2.0 \\ 1.0 \end{pmatrix} u(t) + \begin{pmatrix} 2.0 \\ 1.0 \end{pmatrix} \xi(x, t),
$$

(50)

$$
y(t) = \begin{pmatrix} 1.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 \end{pmatrix} x(t),
$$

that is, $\mathcal{T} = (2.0, 1.0)$. In this example we assume that the function $\xi(x, t)$ and the positive scalar $\chi^*(t) \in \mathbb{R}^1$ in (5) are given by $\chi(x, t) = \sqrt{5.0} \times \sin(x_3(t))$ and $\chi^*(t) = \sqrt{5.0}$, respectively.

Firstly by adopting the similar way to the standard linear quadratic control problem, we consider designing the fixed gain matrix $K_{\mathcal{LQ}} \in \mathbb{R}^{2 \times 3}$. Thus selecting the design parameters $Q_1 \in \mathbb{R}^{3 \times 3}$ and $R_1 \in \mathbb{R}^3$ such that $Q_1 = I_3$ and $R_1 = 1.0 \times 10^3$, respectively, and solving the algebraic Riccati equation $H_e \{ A^T X_1 \} - X_1 B R_1^{-1} B^T X_1 + Q_1 = 0$, we obtain

$$
X_1 = \begin{pmatrix} 2.23662 \times 10^{-1} & 1.04459 & 1.95383 \times 10^{-2} \\ * & 1.6170145 \times 10^1 & 1.25624 \\ * & * & 2.22160 \times 10^{-1} \end{pmatrix},
$$

(51)

$$
K_{\mathcal{LQ}} = \begin{pmatrix} -1.49191 \times 10^{-1} & -1.82593 & -1.29532 \times 10^{-1} \end{pmatrix},
$$

(52)

Next we design an output feedback gain matrix $K \in \mathbb{R}^{1 \times 2}$ by using the the nominal system of (9). We select the design parameter $\alpha \in \mathbb{R}^1$ such as $\alpha = 5.0$, then by applying the LMI-based design algorithm (see [29] and the Appendix section in [15]), we obtain the following solutions for LMI problems in Appendix in [15] and the following gain matrices, respectively:

$$
X = \begin{pmatrix} 1.26046 \times 10^3 & 1.25379 \times 10^2 & 3.19323 \times 10^2 \\ 8.97329 \times 10^1 & -1.13357 \times 10^2 & 5.45874 \times 10^2 \\ * & * & 5.45874 \times 10^2 \end{pmatrix},
$$

$$
y = \begin{pmatrix} 1.02905 \times 10^3 & -5.78548 \times 10^2 & -4.11077 \times 10^2 \end{pmatrix},
$$

(53)


\[ P = \begin{pmatrix}
1.34029 \times 10^1 & -4.46413 \times 10^1 & -2.31136 \\
* & 1.74644 \times 10^2 & 4.12700 \\
* & * & 1.03904 
\end{pmatrix}, \]

\[ \rho = 4.38355 \times 10^4, \]

\[ K = \begin{pmatrix}
2.18703 \\
-1.63634 \times 10^1 \\
-5.43048 
\end{pmatrix}, \]

\[ K_d = \begin{pmatrix}
3.17745 \times 10^{-1} \\
-1.20809 \times 10^1 
\end{pmatrix}. \]

Now, we use Theorem 5 to design the proposed adaptive gain robust output feedback controller; that is, we solve the LMIs of (14). By selecting the symmetric positive definite matrix \( Q \in \mathbb{R}^{3 \times 3} \) such as \( Q = 0.1 \times I_3 \), we have

\[ \delta = \begin{pmatrix}
1.67330 \\
-1.49618 \\
-6.50900 \times 10^{-3} 
\end{pmatrix}, \]

\[ \Xi = \begin{pmatrix}
4.30003 \\
2.32356 \\
2.04206 
\end{pmatrix}, \]

\[ \Psi_f = \begin{pmatrix}
1.07015 \\
8.119016 \times 10^{-1} \\
1.00293 
\end{pmatrix}, \]

\[ \Psi_e = \begin{pmatrix}
9.56125 \\
7.12983 \\
9.13154 
\end{pmatrix}, \]

\[ \gamma_1 = 7.14302, \quad \gamma_2 = 7.57592, \]

\[ \epsilon = 9.55061. \]

In this example, we consider the following two cases for the unknown parameter \( \Delta(t) \in \mathbb{R}^{1 \times 2} \) and its unknown upper bound \( \theta^* \) in (5):

**Case (1):**

\[ \theta^* = 5.0, \]

\[ \Delta(t) = \theta^* \times \left( 8.113249 \times 10^{-4} \right) \times 10^{-1}. \]

**Case (2):**

\[ \theta^* = 5.0 \times 10^{-1}, \]

\[ \Delta(t) = \theta^* \times \left( \sin(10\pi t) \cos(10\pi t) \right). \]

Furthermore, initial values for the uncertain system of (50), its target model, and the adjustable parameter \( \hat{\theta}(t) \) are selected as \( x(0) = (1.5 \ 2.0 \ -4.5)^T \), \( x_i(0) = (2.0 \ 2.0 \ -5.0)^T \), and \( \hat{\theta}(0) = 1.0 \), respectively. Besides, we choose \( \sigma(t) \in \mathbb{R}^+ \) in (16) and the design parameter \( \beta \in \mathbb{R}^1 \) in (17) such as \( \sigma(t) = 5.0 \times 10^4 \times \exp(-1.0 \times 10^{-4}t) \) and \( \beta = 2.5 \times 10^3 \), respectively.

The results of the simulation of this example are depicted in Figures 1, 2, 3, 4, and 5. In these figures, the time histories of the state variables \( x_1(t), x_2(t), \) and \( x_3(t) \), the control input \( u(t) \), and the adjustable parameter \( \theta(t) \in \mathbb{R}^1 \) are shown. “Desired” in these figures represents the trajectories for the state variables and the control input generated by the target model; that is, “Desired” for state variables shows a desirable response for the uncertain nonlinear systems.

From Figures 1, 2, 3, and 4, we find that the proposed adaptive gain robust output feedback controller stabilizes the controlled system of (50) in spite of plant uncertainties and nonlinearities. Besides, one can see that the proposed adaptive gain robust output feedback controller achieves good transient performance and the proposed control input is tuned by the measurable signals and the adjustable parameter. In particular the time responses of the uncertain nonlinear system of (50) for Case (2) close to the trajectory of the target model and the control input also achieves satisfactory trajectory. This result shows that the proposed adaptive robust output feedback controller reflects the effect of uncertainties and nonlinearities as online information.
Furthermore, we can see from Figure 5 that the parameter $\hat{\theta}(t)$ is tuned by the updating law and is not an estimate of the upper bound $\theta^*$. 

6. Conclusions

In this paper, we have proposed an adaptive gain robust output feedback controller for a class of uncertain Lipschitz nonlinear system of which upper bounds are unknown. For the uncertain Lipschitz nonlinear system, we have shown that the proposed adaptive gain robust output feedback controller can be obtained by solving LMIs. Besides, by numerical simulations, the effectiveness of the proposed adaptive robust controller has been presented. One can see that the crucial difference between the existing results [24] and our new one is that information is not required on the upper bound of the unknown parameter $\Delta(t)$ in the system matrix. The proposed controller design method is adaptable when some assumptions are satisfied and in cases where only the output signal of the system is available. Namely, if for uncertain Lipschitz nonlinear systems which satisfy the assumptions for the system parameters, the LMIs of (14) are feasible, then the proposed adaptive gain robust output feedback controller is applicable.

The future research subjects are extension of the proposed adaptive gain robust output feedback controller synthesis to such a broad class of systems as uncertain time-delay systems and uncertain large-scale interconnected systems. Furthermore in future work, we will examine the assumption of (7).

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