Robust $H_r$ Output Tracking Control for a Class of Nonlinear Systems with Time-Varying Delays

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Robust $H_\infty$ output tracking control for a class of nonlinear systems with time-varying delays

Luis F. C. Figueredo · João Y. Ishihara · Geovany A. Borges · Adolfo Bauchspiess

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Abstract This paper addresses the $H_\infty$ output tracking problem for a class of nonlinear systems subjected to model uncertainties and with interval time-varying delay. The stability of the nonlinear time-delay system is analyzed with a novel delay-interval-dependent Lyapunov-Krasovskii functional. Compared to state-of-the-art criteria for linear and nonlinear time-delay systems, less conservative stability conditions are derived with the introduction of new delay-interval-dependent terms and the exploitation of the delay subintervals size. The proposed analysis considers that the delay derivative is either upper and lower bounded, bounded above only, or unbounded, i.e., no restrictions are cast upon the derivative. Numerical examples are provided to enlighten the importance and advantages of the present criterion which outperforms previous criteria in time-delay systems literature. Also, an additional example is provided to highlight the effectiveness of the proposed $H_\infty$ output tracking control design technique for complex nonlinear systems with time-varying delay.

Keywords Nonlinear time-delay systems · Delay-dependent stability · $H_\infty$ control · Output tracking

1 Introduction

Time-delay systems belong to a class of infinite-dimensional systems often described by functional differential equations. The phenomena are encountered in various practical systems, e.g., biological, chemical systems, networked control systems, etc, [1,10,29]. They are usually employed in the description of propagation and transport phenomena, arising as feedback delays in control loops [19]. Since their existence can degrade systems performance and even cause instability, time-delay systems modeling, stability, and stabilization problems have emerged as a topic of significant interest to the control community, which is highlighted by several surveys and studies on the subject, see, e.g., [1,10,19,29]. Among recent results, the following should be mentioned due to their contribution to time-delay systems analysis [5,7,8,24,36,38]. Nonetheless, although being a fundamental issue in control theory,
tracking performance and control have received little attention in time-delay systems literature, especially if we regard nonlinear time-delay systems. In this context, we investigate the $H_{\infty}$ output tracking problem for a class of nonlinear systems subjected to model uncertainties and with input time-varying delay.

It is well recognized that the tracking problem is more general and challenging than stability and stabilization problems [8]. The main objective of the tracking control is to synthesize feedback controllers to make the output of a given plant asymptotically tracks a desired reference whereas ensuring disturbances attenuation properties. The importance of tracking is reflected by the extensive coverage with numerous applications in the areas of robot control, flight control, dynamic processes in industry, economics, etc, see, e.g., [2,16], and the references there in.

Nonetheless, existing results on time-delay systems rarely focus on tracking control problems. Indeed, time-delay systems literature contain several works on control design, e.g., [9,22,27,35], however very few regard the tracking problem. Among these works, the following should be mentioned for their important contributions. In [15], the tracking for switched linear systems with delayed states have been investigated, but with no regard to the time-delay effects on the feedback-loop. The authors in [34] were the first to investigate the tracking problem with constant feedback delays, and their work have been extended to the $H_{\infty}$ tracking control with time-varying delays in [5,8]. The tracking control problem for nonlinear time-delay systems has been addressed in [13,37], however the results are only valid for state-delayed systems, i.e., the time-delay effects on the feedback-loop were not considered. Since time-delay phenomena often arise as feedback delays in the control loop [19], the much more general and realistic scenario regarding the tracking of nonlinear systems with feedback delays still needs to be considered. To the best of the authors’ knowledge, this scenario has never been considered and remains challenging. In this context, the introduction of less conservative stability techniques with the solution of this open problem are the major motivation of the present study.

During the last decade, various methods have been taken for deriving stability conditions for linear time-delay systems using different Lyapunov–Krasovskii functionals (LKFs). Particularly, a recent Lyapunov-based technique must be stressed for its significant contributions to delay-dependent stability analysis: the piecewise analysis method (PAM). The method has similar concepts to the discretized Lyapunov functionals technique (DLF) [10], although applied to time-varying delays, and has been successfully employed in recent literature, see, e.g., [6,7,24]. Still, we believe its potential has not yet been fully exploited. Therefore, with novel less-restricted delay-interval-dependent LKFs, and by introducing the interval size information to the analysis, we improve the piecewise analysis method and considerably amend the stability results for time-delay systems.

In this context, the present paper brings an important contribution to the $H_{\infty}$ output tracking analysis and control design for time-delay systems. The development of a novel delay-interval-dependent Lyapunov-Krasovskii functional, with the improved PAM, provides the conditions under which the prescribed $H_{\infty}$ output tracking performance for a class of nonlinear uncertain time-delay system is achieved. The time-varying nonlinearities are assumed to be norm-bounded, satisfying a quadratic constraint. Moreover, it should be mentioned that the proposed tracking criterion, if particularized to stability analysis, also yields considerably superior results compared with state-of-the-art criteria for linear or nonlinear time-delay systems. The analysis is enriched with numerical examples that illustrates the advantages of our criteria, which outperform previous criteria in the literature, and with an additional example that shows the effectiveness of the proposed $H_{\infty}$ output tracking control for nonlinear time-delay systems.

**Notations.** Throughout the paper the superscript ‘T’ stands for matrix transposition, $\mathbb{R}^n$ denotes the $n$-dimensional Euclidean space, and $\mathbb{R}^{n \times p}$ defines the set of all $n \times p$ real matrices. The notation $\text{diag}\{\cdots\}$ stands for a diagonal matrix, $P>0$ means that $P$ is symmetric and positive definite, and the symmetric term in a matrix is denoted by $\ast$. The notation $A|_{s \to b}$ stands for the limit of a $s$-dependent matrix $A$ as $s \to b$. Matrices, if not explicitly stated, are assumed to have compatible dimensions.
2 Problem formulation and preliminaries

Consider a class of continuous-time nonlinear uncertain systems with time-varying delay:

\[
\begin{aligned}
\dot{x}_p(t) &= (A_p + \Delta A_p)x_p(t) + (B_p + \Delta B_p)u_p(t-d(t)) + g(t, x_p(t), x_p(t-d(t))) + B_{pw}\omega(t), \\
y_p(t) &= (C_p + \Delta C_p)x_p(t) + (D_p + \Delta D_p)u_p(t-d(t)), \quad t > 0 \\
x_p(t) &= \rho(t), \quad t \in [-\tau_{max}, 0]
\end{aligned}
\]

where \(x_p(t) \in \mathbb{R}^{r_x}, u_p(t) \in \mathbb{R}^{r_u}, y_p(t) \in \mathbb{R}^{r_y}\) denote the plant’s state, control input and output vectors, respectively, \(\omega(t) \in \mathbb{R}^{r_{\omega}}\) denotes the exogenous disturbance signal which is assumed to belong to \(L_2[0, \infty)\). \(A_p, B_p, B_{pw}, C_p, D_p\) are constant matrices with appropriate dimensions, \(\rho(t)\) describes the state’s initial condition, and \(g(t, x_p(t), x_p(t-d(t))) : \mathbb{R}_+ \times \mathbb{R}^{r_x} \times \mathbb{R}^{r_u} \to \mathbb{R}^{r_y}\) denotes a class of piecewise-continuous nonlinear functions in \(t, x_p(t), x_p(t-d(t))\), which are assumed to satisfy the quadratic condition:

\[
g^T(t, x_p(t), x_p(t-d(t)))g(t, x_p(t), x_p(t-d(t))) \leq \alpha_1 x_p^T(t)H_1 x_p(t) + \alpha_2 x_p^T(t-d(t))H_2 x_p(t-d(t)),
\]

where \(\alpha_1, \alpha_2\) are positive known bounding parameters of \(g(t, x_p(t), x_p(t-d(t)))\), and \(H_1\) and \(H_2\) are known constant matrices. The system uncertainties are assumed to be time-varying matrices,

\[
\begin{align*}
[\Delta A_p, \Delta B_p] &= \Xi_x \Delta(t) [\Xi_A, \Xi_B], \\
[\Delta C_p, \Delta D_p] &= \Xi_y \Delta(t) [\Xi_C, \Xi_D]
\end{align*}
\]

where \(\Xi_x, \Xi_A, \Xi_B, \Xi_y, \Xi_C, \Xi_D\) are known matrices with appropriate dimensions, and \(\Delta(t)\) is an unknown time-varying matrix, which is Lebesgue measurable in \(t\) and satisfies \(\Delta(t) \leq I\).

We consider the reference signal, \(y_r(t) \in \mathbb{R}^{r_y}\), to be the output of the given linear system:

\[
\begin{align}
\dot{x}_r(t) &= A_r x_r(t) + r(t), \\
y_r(t) &= C_r x_r(t),
\end{align}
\]

where \(x_r(t), r(t) \in \mathbb{R}^{r_y}\) are the reference’s state vector and the energy bounded reference input, respectively, \(A_r\) is a Hurwitz matrix, and \(C_r\) is a constant matrix with appropriate dimensions.

Finally, the continuous function \(d(t)\) denotes the time-varying delay which satisfies

\[
\begin{align}
\tau_{min} \leq d(t) \leq \tau_{max}, \\
d_{min} \leq d(t) \leq d_{max},
\end{align}
\]

where the constants \(0 \leq \tau_{min} \leq \tau_{max}\) and \(d_{min} \leq d_{max}\) denote the bounding parameters of \(d(t)\) and \(\dot{d}(t)\), respectively. In this paper, we also consider the case when \(d_{min}\) is unknown, and when no restrictions are cast upon the delay derivative, i.e., when it is assumed to be fast-varying.

Considering (1)-(5) with a state feedback control law \(u_p(t) = \bar{K} [x_p^T(t) x_r^T(t)]^T\), we obtain the augmented closed-loop nonlinear system with time-varying delay

\[
\begin{aligned}
\dot{x}(t) &= (\bar{A} + \Delta \bar{A})x(t) + (\bar{B} + \Delta \bar{B})\bar{K}x(t-d(t)) + \bar{g}(t, x(t), x(t-d(t))) + \bar{B}_\omega \bar{\omega}(t), \\
e(t) &= (\bar{C} + \Delta \bar{C})x(t) + (\bar{D} + \Delta \bar{D})\bar{K}x(t-d(t)),
\end{aligned}
\]

where \(x^T(t) := [x_p^T(t) x_r^T(t)] \in \mathbb{R}^{r_x}, \bar{\omega}^T(t) := [\omega^T(t) r^T(t)] \in \mathbb{R}^{r_{\omega}}\),

\[
\begin{align*}
\bar{A} := \begin{bmatrix} A_p & 0 \\ 0 & A_r \end{bmatrix}, \\
\bar{B} := \begin{bmatrix} B_p \\ 0 \end{bmatrix}, \\
\bar{B}_\omega := \begin{bmatrix} B_{pw} & 0 \\ 0 & I \end{bmatrix}, \\
\bar{C} := \begin{bmatrix} C_p & -C_r \end{bmatrix}, \\
\bar{D} := D_p,
\end{align*}
\]

\[
\begin{align*}
[\Delta \bar{A}, \Delta \bar{B}] &= \Xi_x \Delta(t) [\Xi_{\bar{A}}, \Xi_{\bar{B}}], \\
[\Delta \bar{C}, \Delta \bar{D}] &= \Xi_y \Delta(t) [\Xi_{\bar{C}}, \Xi_{\bar{D}}],
\end{align*}
\]

\[
[\bar{g}(t, x(t), x(t-d(t)))] := \begin{bmatrix} I \\ 0 \end{bmatrix}^T \bar{g}(t, x(t), x(t-d(t))).
\]

The matrix \(\bar{K}\) is the state-feedback controller, and \(e(t) := y_p(t) - y_r(t)\) denotes the output tracking error.
Tracking problem: We desire the plant’s output $y_p(t)$ to asymptotically track a given reference signal $y_r(t)$. Our purpose is therefore to design a robust state-feedback controller $K$ such that the output tracking performance $\gamma$ is ensured in the $H_{\infty}$ sense.

Definition 1 For a prescribed scalar $\gamma > 0$, the nonlinear time-delay system (6) achieves $H_{\infty}$ output tracking performance, if for any realization of the uncertainties $\Delta A$, $\Delta B$, $\Delta C$, $\Delta D$, the following hold

1. The augmented closed-loop nonlinear system (6) with $\dot{\omega}(t) \equiv 0$ is asymptotically stable;
2. Under the assumption of zero initial condition, the disturbance effect on the tracking error is attenuated below a prescribed level $\gamma$, $\|e(t)\|_2 < \gamma \|w(t)\|_2$, for all nonzero $w \in L_2[0, \infty)$.

3 $H_{\infty}$ output tracking control design

This section presents the main results of this paper. First, we divide the delay range $[\tau_{\min}, \tau_{\max}]$ into two equally spaced subintervals: $[\tau_1, \tau_2]$ and $[\tau_2, \tau_3]$, where $\tau_1=\tau_{\min}$, $\tau_3=\tau_{\max}$, and $\tau_2=\frac{1}{2} (\tau_{\max}+\tau_{\min})$. Note that one can consider different partitioning strategies (e.g., in [23], $\tau_2$ is defined to be anywhere between $\tau_{\min}$ and $\tau_{\max}$). Still, choosing equally subintervals, $\tau_3=\tau_2=\tau_1$, adds more information to the analysis which is used to obtain less conservative criteria. In this context, we also define the auxiliary variable

$$\tau_{\sigma} := \tau_2 - \tau_1,$$

and the delay-interval-dependent indicator function $\chi_{[\tau_1, \tau_2]}: \mathbb{R}_+ \to \{0, 1\}$, which is assumed to be 1, if $d(t) \in [\tau_1, \tau_2]$, and $\chi_{[\tau_1, \tau_2]} = 0$, otherwise.

The indicator function enlightens the piecewise analysis method main contribution: the establishment of different linear matrix inequalities (LMIs) for each subinterval, reducing the conservatism which arises from the analysis of the delay range $[\tau_{\min}, \tau_{\max}]$. In this context, it is proposed the following delay-interval-dependent LKF candidate

$$V(t) = \sum_{i=1}^{3} V_i(t),$$

$$V_1(t) = \chi_{[\tau_1, \tau_2]} x^T(t) \hat{P}_1 (d(t)) x(t) + (1-\chi_{[\tau_1, \tau_2]}) x^T(t) \hat{P}_2 (d(t)) x(t),$$

$$V_2(t) = \int_{t-d(t)}^{t-\tau_1} x^T(s) Q x(s) ds + \int_{t-\tau_1}^{t-\tau_2} x^T(s) [x^T(s) x^T(s-\tau_2)]_N x^T(s) x^T(s-\tau_2) ds + \left(\tau_2 - \tau_1\right) \int_{t-\tau_1}^{t-\tau_2} x^T(s) N_2 x(s) ds + \left(\tau_2 - \tau_1\right) \int_{t-\tau_1}^{t-\tau_2} x^T(s) N_3 x(s) ds + \int_{t-\tau_2}^{t} \varphi^T(s) N_4 \varphi(s) ds,$$

$$V_3(t) = \sum_{k=1}^{2} \int_{t-d(t) + \frac{\tau_1 - \tau_2}{2}}^{t} \xi^T(k) x(s) \varphi(s) ds + \left(\tau_2 - \frac{\tau_1 - \tau_2}{2}\right) \int_{t}^{t} \xi^T(s) S_{k+2} \xi(s) ds + \int_{t}^{t} \xi^T(s) (R_1 + R_2) \xi(s) ds + \int_{t}^{t} \xi^T(s) (R_3 + R_4) \xi(s) ds,$$

where $\varphi^T(s) = [x^T(s) x^T(s-\tau_1) x^T(s-\tau_2)]$, and the function matrices in $V_i(t)$ are defined as follows

$$\hat{P}_1 (d(t)) := \frac{d(t) - \tau_1}{\tau_2 - \tau_1} P_1 + \frac{\tau_2 - d(t)}{\tau_2 - \tau_1} P_2,$$

$$\hat{P}_2 (d(t)) := \frac{d(t) - \tau_2}{\tau_3 - \tau_2} P_1 + \frac{\tau_3 - d(t)}{\tau_3 - \tau_2} P_2.$$
It is interesting to highlight that the novel delay-dependent, delay-derivative-dependent, and delay-interval-dependent LKF terms that are introduced in the Lyapunov candidate (9) lead to less-restricted stability conditions. The novel and enhanced stability analysis presented in the next Subsection and the improvements illustrated in Section 4 stem mostly from these new set of Lyapunov terms.

Remark 1 Different strategies for the delay range partitioning have been previously considered, leading to different results. In [23], the dividing parameter \( \tau_2 \) is manually defined at any point within the delay range, whereas in [39], it is proposed an iterative optimization procedure based on the Nelder-Mead simplex algorithm [21] to define the partitioning points. Still, by simply choosing equally spaced subintervals, we add more information to the analysis. Indeed, the exactly new partitioning subinterval size together with the knowledge regarding the relationship between the system’s states \( x(t), x(t-\tau_i) \), \( i=1,2,3 \) deem the new auxiliary variable \( \tau_\sigma \) essential for the construction of Lyapunov candidate, as we can switch among these states solely by adding/subtracting a delay equal to \( \tau_\sigma \), e.g., \( V_2(t) \). The amount of information and the relationship described solely by \( \tau_\sigma \) can only stem from a good partitioning technique, and that leads to an improved exploitation of the delayed states during the design of the Lyapunov candidate (10), e.g., by adding a delay \( \tau_\sigma \) to \( \varphi^T(t) \) we obtain \( \varphi^T(t-\tau_\sigma)=[x(t-\tau_2) x^T(s-\tau_2) x^T(s-\tau_3)] \). Furthermore, since we are adding non-diagonal terms to relate the states \( x(t), x(t-\tau_\sigma), x(t-\tau_i) \), \( i=1,2,3 \), we esteem to obtain less restrictive LMI constraints which in turn leads to less conservative results.

3.1 Robust \( H_\infty \) output tracking performance analysis

In this subsection, we derive conditions under which the closed-loop uncertain nonlinear system (6) achieves \( H_\infty \) output tracking performance \( \gamma \), namely, the augmented closed-loop system is asymptotically stable and satisfies the performance conditions described in Definition 1. The following result stems from the proposed Lyapunov candidate (9), and describes a novel robust criterion for the output tracking in the \( H_\infty \) sense.

Theorem 1 For a prescribed \( \gamma>0 \), and given scalars \( \tau_{\min}, \tau_{\max}, d_{\min}, d_{\max} \) such that \( 0 \leq \tau_{\min} \leq \tau_{\max} \) and \( d_{\min} < d_{\max} \), and given controller gain \( K \), the augmented closed-loop nonlinear system (6) with time-varying delay satisfying (5), parameter uncertainties and nonlinearities described in (7) and (2), respectively, achieves \( H_\infty \) output tracking performance \( \gamma \), if there exist positive scalars \( \epsilon_{11}, \epsilon_{12}, \epsilon_{21}, \epsilon_{22}, \eta_1, \eta_2 \) matrices \( P_1, P_2, Q, Z_0, Z_1, Z_2 N_i, S_i, R_i \), \( i=1,2,3,4 \), with appropriate dimensions, satisfying (11) and free-weighting matrices \( F_1, F_2, V_1, V_2 \in \mathbb{R}^{r_x \times r_x} \), such that the following hold for \( k=\{1,2\} \), where

\[
\Omega_{lk} = \begin{bmatrix} \Pi_1 + \psi^{(1)}(d_t) & \tau_\sigma V_1 J_1 \ & Y_1 \\ * & -\tau_\sigma A_1 \ & \Sigma_1 \\ * & * & \end{bmatrix}, \quad \Omega_{lk} = \begin{bmatrix} \Pi_2 + \psi^{(2)}(d_t) & \tau_\sigma V_2 J_2 \ & Y_2 \\ * & -\tau_\sigma A_2 \ & \Sigma_2 \\ * & * & \end{bmatrix},
\]

with \( J_1 = \begin{bmatrix} 0 & 1 \end{bmatrix}^T \), \( J_2 = \begin{bmatrix} I & 0 \end{bmatrix}^T \) and

\[
\begin{align*}
A_{11} &= \tau_\sigma Z_1 + R_1 + R_4, \\
A_{12} &= \tau_\sigma Z_1 + U_R, \\
A_{21} &= \tau_\sigma Z_2 + R_1 + R_4, \\
A_{22} &= \tau_\sigma Z_2 + U_R + R_3 - 1, \\
\Sigma_t &= \Sigma_t \Xi_t \Xi_t^T - \text{diag} \{ \gamma_1^2 I; \epsilon_{11}; \epsilon_{12}; \epsilon_{21}; \epsilon_{22}; \eta_1 \},
\end{align*}
\]

\[
\begin{align*}
F_1 &= J_2 (I_2 - L_5) + J_4 (I_6 - L_2), \\
F_2 &= J_2 (I_2 - L_6) + J_3 (I_7 - L_2), \\
\Sigma_t &= \Sigma_t \Xi_t \Xi_t^T + \text{diag} \{ \gamma_2^2 I; \epsilon_{11}; \epsilon_{12}; \epsilon_{21}; \epsilon_{22}; \eta_1 \},
\end{align*}
\]

\[
\begin{align*}
F_1 &= F_1 T_s + G_1, \\
F_2 &= F_2 T_s + G_1, \\
R_t &= R_t + d(t) R_4 + (1 - d(t)) R_2, \\
U_t &= \frac{d}{d_t} U, \text{ if } \tau_3 \neq 0, \text{ or } U_0 = 0, \text{ otherwise, }
\end{align*}
\]

\[
\begin{align*}
\psi^{(1)}(d_t) = y_u - I_2 (I_2 - T_s) & \psi^{(1)} A_1 (I_6 - I_T) + I_3 ((T_s - d(t)) R_4 + d(t) R_2) T_s^T + 2 \tilde{P}_1 (d(t)) T_s^T + \eta_2 (d(t)) T_s^T,
\]

where the matrices \( \Pi_1, \Pi_2, \Sigma_1, \Sigma_2 \) are defined by

\[
\begin{align*}
\Pi_1 &= \Pi_1 + \psi^{(1)}(d_t), \\
\Pi_2 &= \Pi_2 + \psi^{(2)}(d_t),
\end{align*}
\]

\[
\begin{align*}
\eta_2 &= \eta_2 (d(t)), \\
\psi^{(1)} &= \psi^{(1)}(d_t).
\end{align*}
\]
The matrices $I_i$, $i = \{1, 2, ..., 8\}$, are block entry matrices with eight elements, e.g., $I_1^T = [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$.  

Remark 2 It is also interesting to consider two particular cases regarding the delay and its derivative information: the case when the time-delay derivative lower bound is unknown, and the case when there exist no information concerning the delay derivative, i.e., fast-varying delays. Theorem 1 can be easily adapted to deal with both cases. For the first case, if we take the conditions $P_2 > P_1$, $R_2 > R_4$ instead of (12b), then Theorem 1 becomes suitable for the analysis when the lower bound, $d_{\text{min}}$, is unknown. Note that, if the above conditions and (12a) hold, then (12b) will be satisfied regardless $d_{\text{min}}$. An evident consequence is the needlessness of the derivative lower bound information for the resulting performance conditions. For the later case, by assuming $P_1 = P_2$, and null $Q, R_2, R_4$ matrices, all the time-delay derivative information is removed from Theorem 1, and the criterion will thus be suitable for the analysis with fast-varying delays. Moreover, it should be mentioned that Theorem 1 can also be applied for nonlinear/time-delay systems if one simply takes $B_u, C, D$ to be null matrices.

Theorem 1 presents conditions which guarantee the $H_\infty$ output tracking performance for nonlinear time-delay systems. The results stem from a novel delay-dependent Lyapunov-Krasovskii functional that enhances the delay fractioning and the piecewise analysis. With interval-dependent terms and by further exploiting the delay partitioning information, we have weakened the positiveness constraints upon new functional terms and matrices, whereas maintaining (9) definite positive and continuously differentiable. The proposed method therefore increases the flexibility analysis upon some matrices and relaxes resulting LMIs conditions, yielding in a considerably reducing of conservatism, even if compared with state-of-the-art results for linear time-delay systems stability analysis.

3.2 Robust $H_\infty$ output tracking controller design

For the $H_\infty$ output tracking control, we seek conditions for the design of a state-feedback gain $K$ which leads the nonlinear time-delay systems output to asymptotically track a desired reference whereas ensuring disturbance attenuation properties in the $H_\infty$ sense. The next theorem provides a solution for the above-mentioned problem, which is non-convex due to the existence of the variable $K$. The main idea is to transform the non-convex problem into a rank minimization problem which may be approximated by a sequence of semi-definite problems involving the trace minimization of certain variables. The $H_\infty$ output tracking control problem is then solved through the use of the cone complementarity linearization algorithm (CCLA) from [4].

Theorem 2 For a prescribed $\gamma > 0$, and given $\tau_{\text{min}}, \tau_{\text{max}}, d_{\text{min}}, d_{\text{max}}$, there exist a feedback gain $K$ such that the resulting closed-loop nonlinear system (6) with input-varying time-delay satisfying (5), uncertainties and nonlinearities described in (7) and (2), respectively, achieves $H_\infty$ output tracking performance $\gamma$, if there exist positive scalars $\xi_{11}, \xi_{12}, \xi_{21}, \xi_{22}$, matrices $P_i, P_2, Q, Z_0, Z_1, Z_2, N_i, S_i, R_i$, $i = \{1, 2, 3, 4\}$, satisfying (11); free-weighting matrices $\Psi_i, \Phi_0 \in \mathbb{R}^{r_x \times r_x}$, $X \in \mathbb{R}^{r_x \times r_x}$; and definite positive matrices $X, F_j, N_i, N_j \in \mathbb{R}^{r_x \times r_x}$, $j = \{1, 2\}$, such that the global minimum of the optimization problem

$$
\min \ tr \left\{ XX + \Psi_1 \Psi_1 + \Psi_2 \Psi_2 + \Psi_1 \Psi_1 + \Psi_2 \Psi_2 \right\},
$$

subject to

$$
\begin{array}{c}
XX \geq 0, \quad [XX \quad I \quad K] \geq 0, \quad [XX \quad I \quad M_k] \geq 0, \quad [XX \quad F_k \quad \Psi_k] \geq 0, \quad [FF_k \quad \Psi_k] \geq 0,
\end{array}
$$
Algorithm 1 \(H_\infty\) output tracking controller design procedure

1. Find a feasible solution for the convex LMI conditions in Theorem 2 (without the optimization problem). If none are found, exit. Else, set \(X^0=X, \bar{X}^0=X, \bar{\eta}^0=\bar{\eta}, \bar{B}^0=\bar{B}, \bar{W}^0=\bar{W}, i=\{1,2\}; \) and \(k=1\).

2. For \(k<k_{\text{lim}}\), find \(X^{k+1} = X, \bar{X}^{k+1} = X, \bar{\eta}^{k+1} = \bar{\eta}, \bar{B}^{k+1} = \bar{B}, \bar{W}^{k+1} = \bar{W}, i=\{1,2\}\). Then, solve the LMIs conditions in Theorem 2 with the following linear minimization:

\[
O_k := \tr \left\{ e^{-1} (X^k X + \bar{X}^k \bar{X} + \bar{\eta}^k \bar{\eta} + \bar{B}^k \bar{B} + \bar{W}^k \bar{W} + \bar{W}^k \bar{W}_i) \right\}, i=\{1,2\}.
\]

3. If \(||O_k - O_{k-1}|| < \epsilon_{\text{lim}}\), where \(\epsilon_{\text{lim}}>0\) is a predefined parameter, exit. Else, set \(k=k+1\) and go to step 2.

4. Stopping criterion: reconstruct the stabilizing controller \(K = X^{-1}\). If (12) is feasible, exit. Otherwise, set \(k=k+1\), reduce \(\epsilon_{\text{lim}}\), and go back to step 2.

The first and every Step 2 of Algorithm 1 are simple LMI problems. Hence, interior point based algorithms can solve the set of convex problems in polynomial time. The predefined constants \(k_{\text{lim}}\) and \(\epsilon_{\text{lim}}\) denote the maximum number of iterations and the threshold for the convergence rate, respectively.
The sequence $O_k$ is monotonically decreasing and, according to [3, 4, 12], the algorithm shows excellent search performance and converges for a wide set of problems when properly set. Therefore, we expect (15) to converge to $5r_x$, when feasible. Indeed, numerical experience reported shows that it extremely efficient and fails to compute the global optimum in a very few cases [3], usually due to a small number of iterations [12].

Remark 3 The results from Theorem 2 with proper modifications, as stressed in Remark 2, are also valid when the delay derivative lower bound is unknown and for fast-varying delays.

4 Numerical examples

This section presents different benchmark examples$^1$ that illustrate the effectiveness of the proposed criteria. First, we investigate the advantages of applying Theorem 1 for the stability analysis of linear time-delay systems, i.e., when $\omega(t)\equiv0$, and the reference’s output, (4), is null. In the second, we show the improvements from the proposed criteria for $H_\infty$ performance analysis. Finally, we present a simulation to illustrate the effectiveness of the proposed $H_\infty$ output tracking control criterion for a class of nonlinear time-delay systems.

Example 1 Consider the following linear system with time-varying delay

$$\dot{x}(t) = \begin{bmatrix} -2 & 0 \\ 0 & -0.9 \end{bmatrix} x(t) + \begin{bmatrix} -1 & 0 \\ -1 & -1 \end{bmatrix} x(t-d(t)).$$

Assuming $\tau_{min}=0$, the maximum allowable upper bound for $\tau_{max}$ from Theorem 1 and from the literature [7, 14, 25, 39] are listed in Table 1. Particularly, the results from [39] are only feasible for $\tau_{min}=0$ and full knowledge about the delay derivative. Hence, the derivative lower bound for [39] is set to $d_{min}=-3$ instead of regarded to be unknown. From Table 1, it is clear that Theorem 1 provides much less conservative results for larger (or unknown) delay derivative bounds. The results from the proposed criterion are only outperformed by [39], and only for very slow-varying delays and a larger number of delay partitions ($m=2$ partitions compared to 1 from Theorem 1). Indeed, for all other conditions, Theorem 1 provides considerably superior results than [39], e.g., for $|d(t)|\leq0.9$, the improvement over [39] is higher than 40%. This illustrates the importance of the proposed method for the analysis of linear time-delay systems.

Now, assuming fast-varying delays, the maxima values for $\tau_{max}$ which maintain the time-delay system stability are listed in Table 2. The results compared to state-of-the-art criteria in the literature enlighten the advantages of Theorem 1, when particularized to the stability analysis of linear time-delay systems. Moreover, compared to the results from different authors [7, 11, 17, 28, 30–33], the improvements from the proposed method become even more evident, e.g., for $\tau_{min}=1$, the delay interval size is 16% larger than the results from [7] (Theorem 1), and 21% larger than the results from [11] and [7], Theorem 2.

\textit{Table 1} Admissible $\tau_{max}$ value for $\tau_{min}=0$ and given $d_{min}$ and $d_{max}$ (Ex. 1)

<table>
<thead>
<tr>
<th>Method</th>
<th>unknown $d_{min}$</th>
<th>$d_{min}=0.3$</th>
<th>$d_{min}=0.9$</th>
<th>$d_{min}=0.5$</th>
<th>$d_{min}=0.9$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>Thm 2</td>
<td>2.337</td>
<td>1.872</td>
<td>2.337</td>
</tr>
<tr>
<td></td>
<td></td>
<td>$m=2$</td>
<td>2.37</td>
<td>1.50</td>
<td>2.590</td>
</tr>
<tr>
<td>Theorem 1</td>
<td>Theorem 1</td>
<td>2.410</td>
<td>2.120</td>
<td>2.501</td>
<td>2.188</td>
</tr>
</tbody>
</table>

$^+$ The notation ‘m’ stands for the number of delay range ($[\tau_{min}, \tau_{max}]$) partitions.

$^1$ All numerical tests have been performed with an Intel Core i7, CPU 870@2.93GHz, 8 GB RAM, using Matlab with SeDuMi [26] and YALMIP [18]. The configuration constants in Algorithm 1 have been set to $k_{im}=300$ and $e_{im}=10^{-3}$. 
Table 2: Allowable upper bound value of $\tau_{\text{max}}$ for fast-varying delays and various $\tau_{\text{min}}$ (Ex. 1)

<table>
<thead>
<tr>
<th>Method</th>
<th>$\tau_{\text{min}}$</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
</tr>
</thead>
<tbody>
<tr>
<td>Tang et al. (2012) [33]</td>
<td></td>
<td>2.045</td>
<td>2.605</td>
<td>3.310</td>
<td>4.088</td>
<td>−</td>
</tr>
<tr>
<td>Souza (2013) [31]</td>
<td></td>
<td>−</td>
<td>3.418</td>
<td>4.210</td>
<td>5.044</td>
<td>−</td>
</tr>
<tr>
<td></td>
<td>Thm 2</td>
<td>2.169</td>
<td>2.646</td>
<td>3.321</td>
<td>4.090</td>
<td>−</td>
</tr>
<tr>
<td>Fridman et al. (2009) [7]</td>
<td></td>
<td>2.120</td>
<td>2.724</td>
<td>3.458</td>
<td>4.257</td>
<td>5.097</td>
</tr>
<tr>
<td><strong>Theorem 1</strong></td>
<td></td>
<td>2.359</td>
<td>2.800</td>
<td>3.482</td>
<td>4.266</td>
<td>5.101</td>
</tr>
</tbody>
</table>

Table 3: Minimum value of $\gamma$ for different values of $\alpha_1$ (Ex. 2)

<table>
<thead>
<tr>
<th>Methods</th>
<th>$\alpha_1^2=0.05$</th>
<th>$\alpha_1^2=0.10$</th>
<th>$\alpha_1^2=0.15$</th>
<th>$\alpha_1^2=0.2$</th>
<th>$\alpha_1^2=0.25$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Orihuela et al. (2011) [24]</td>
<td>0.92</td>
<td>1.20</td>
<td>1.60</td>
<td>2.66</td>
<td>7.70</td>
</tr>
<tr>
<td><strong>Theorem 1</strong></td>
<td><strong>0.816</strong></td>
<td><strong>0.906</strong></td>
<td><strong>1.038</strong></td>
<td><strong>1.253</strong></td>
<td><strong>1.706</strong></td>
</tr>
<tr>
<td>Improvements:</td>
<td>(13%)</td>
<td>(33%)</td>
<td>(54%)</td>
<td>(112%)</td>
<td>(351%)</td>
</tr>
</tbody>
</table>

Table 4: Minimum achievable value of $\gamma$ for different values of $\alpha_1$ (Ex. 3)

<table>
<thead>
<tr>
<th>$\alpha_1$</th>
<th>$\gamma_{\text{min}}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\alpha_1=1$</td>
<td>0.395</td>
</tr>
<tr>
<td>$\alpha_1=5$</td>
<td>0.429</td>
</tr>
<tr>
<td>$\alpha_1=10$</td>
<td>0.470</td>
</tr>
<tr>
<td>$\alpha_1=20$</td>
<td>0.575</td>
</tr>
<tr>
<td>$\alpha_1=30$</td>
<td>0.705</td>
</tr>
<tr>
<td>$\alpha_1=40$</td>
<td>0.990</td>
</tr>
<tr>
<td>$\alpha_1=50$</td>
<td>1.921</td>
</tr>
</tbody>
</table>

**Example 2** Consider the following nonlinear time-delay systems

\[
\dot{x}(t) = \begin{bmatrix} 1 & 1 \\ 0 & 0.99 \end{bmatrix} x(t) + \begin{bmatrix} 0 \\ -3.715 \end{bmatrix} x(t-d(t)) + \begin{bmatrix} 0.1 \\ 0 \end{bmatrix} \omega(t) + \tilde{g}(t, x(t), x(t-d(t))),
\]

\[
y(t) = \begin{bmatrix} 0 \\ 1 \end{bmatrix} x(t) + \begin{bmatrix} -0.03715 \\ -0.03514 \end{bmatrix} x(t-d(t)),
\]

and with nonlinearity $\tilde{g}(t, x(t), x(t-d(t)))$ satisfying (2) with $H_1 = \begin{bmatrix} 1 & 0 \end{bmatrix}$ and $H_2 = 0$. To allow comparison with existing methods, the reference signal $y_r(t)$ is considered to be null.

Assuming fast-varying delay and $d(t) \in [0, 0.2509]$, Table 3 presents the values for the noise to error attenuation, $\gamma$, for different values of the bounding parameter $\alpha_1$. From Table 3, it can be seen that the results from Theorem 1 are considerably less conservative than the ones from the state-of-the-art criterion, given by [24]. Moreover, in order to compare with different criteria, we also consider the case when there are no external disturbances, $\omega(t) = 0$. In this particular case, the maximum bounding parameter obtained with [27] and [24] are $\alpha_1^2 = 0.164$ and $\alpha_1^2 = 0.276$, respectively, whereas using Theorem 1, we find a maximum bound $\alpha_1^2 = 0.365$.

**Example 3** Now, consider the example of a satellite system, [8], modeled by rigid bodies joined through a link with torque 0.09 N·m and yaw angles denoted by $\theta_1$ and $\theta_2$. Differently from [8], a more realistic scenario is presented with a nonlinear viscous damping $f=0.04 + g(t, \theta(\dot{t}), \theta(t-d(t)))$ N·s/m. Taking the angular position $\theta_2$ as the system’s output $y_r(t)$, the state-space representation is derived as

\[
\begin{bmatrix} \dot{\theta}_1(t) \\ \dot{\theta}_2(t) \\ \dot{\theta}_1(t) \\ \dot{\theta}_2(t) \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -0.09 & 0.09 & -0.04 & 0.04 \\ 0.09 & -0.09 & 0.4 & -0.04 \end{bmatrix} \begin{bmatrix} \theta_1(t) \\ \theta_2(t) \\ \theta_1(t) \\ \theta_2(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} u(t-d(t))+ \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix} \omega(t)+ \begin{bmatrix} 0 \\ 0 \end{bmatrix} g(t, \theta, \dot{\theta}_d),
\]

\[
y_r(t) = \begin{bmatrix} 0 & 1 & 0 & 0 \end{bmatrix} [\theta_1(t) \theta_2(t) \dot{\theta}_1(t) \dot{\theta}_2(t)]^T,
\]

with the nonlinearity $g(t, \theta, \dot{\theta}_d)=\sin(\theta_1(t))\text{sgn}(\dot{\theta}(\dot{t}))\sqrt{\alpha_1|h(\theta)|}$, $h(\theta)=\theta_1^2(t)+2\dot{\theta}_1(t)\dot{\theta}_2(t)+\frac{1}{2}\dot{\theta}_2^2(t)$, satisfying the quadratic constraint (2) for $H_1=[0 \ 0.1 \ 0.1 \ 0.1]$, $H_2=0$. The reference model is defined as

\[
\dot{x}_r(t) = -x_r(t) + r(t), \quad y_r(t) = 0.5x_r(t).
\]
The prescribed delay is achieved using uniform distribution random delay. Moreover, considering zero initial conditions and numerically computing otherwise. The prescribed delay is achieved using uniform distribution random delay. The plot (a) shows the system’s output tracking the reference’s signal, despite the nonlinearity, is unable of tracking the desired reference, whereas \((y^\text{nonl}_p(t))\). Note the linear controller fails in tracking the reference, whereas \((y^\text{nonl}_p(t))\) successfully tracks \(y_r(t)\).

In this context, we assume a fast-varying delay with \(\tau_{\text{min}}=0.08\ s, \tau_{\text{max}}=0.1\ s\). Our purpose is to find a stabilizing controller \(\hat{K}\), which makes the system’s output, \(y_p(t)\), asymptotically tracks the signal \(y_r(t)\), whereas minimizing the upper bound of the disturbance attenuation, \(\gamma\). From Algorithm 1, the minima values of \(\gamma\) for different values of the bounding parameter \(\alpha_1\) are listed in Table 4.

**Simulation:**
In the first scenario, regard a fast-varying delay with \(d(t)\in[0.08,0.1]\), and \(\alpha_1=1\). From Theorem 2 and Algorithm 1, we find the stabilizing controller \(\hat{K}=-[18.79 \ 939.80 \ 10.60 \ 688.47 - 213.13]\), which yields an \(H_\infty\) performance with \(\gamma=0.395\). The disturbances are assumed \(\omega(t)=0.5\sin(0.3t)\), for \(6\leq t\leq 24\) or \(t>37\), and \(\omega(t)=2.75\sin(0.7t)\), otherwise; and \(r(t)=0.3\), for \(10\leq t\leq 20\) or \(t>33\), and \(r(t)=2\sin(0.9t)\), otherwise. The prescribed delay is achieved using uniform distribution random delay.

The system’s and the reference’s output are presented in Figure 1(a). It is clear that \(y_p(t)\) successfully tracks the reference’s signal, despite the nonlinearity, \(g(t,\theta(t),\theta(t-d(t)))\), and the time-varying delay. Moreover, considering zero initial conditions and numerically computing \(\|e(t)\|_2\) and \(\|\epsilon(t)\|_2\), we find the simulated disturbance attenuation, \(\|e(t)\|_2=0.314\), which is smaller than the minimum upper bound \(\gamma=0.395\) obtained with Theorem 2, highlighting the effectiveness of the proposed method.

To further enlighten the importance of Theorem 2, we propose a second scenario with a higher nonlinear influence, i.e., we will increase the value of \(\alpha_1\) to \(\alpha_1=50\). From Theorem 2 and Algorithm 1 for \(\alpha_1=50\), we find a stabilizing controller that yields an output \((y^\text{nonl}_p(t))\) with an \(H_\infty\) performance of \(\gamma=1.921\). In this scenario, we also define a linear controller yielding an output \((y^\text{lin}_p(t))\) that satisfies the \(H_\infty\) performance of \(\gamma=1.921\), however ignoring the existence of the nonlinear term \(g(t,\theta,\theta_d)\). The disturbances are assumed \(\omega(t)=9\sin(1.25t)\), \(r(t)=10\sin(0.5t)\). The results, presented in Figure 1(b), clearly demonstrates that \((y^\text{lin}_p(t))\) from the linear controller (ignoring the existence of nonlinearities) is unable of tracking the desired reference, whereas \((y^\text{nonl}_p(t))\) from Theorem 2 with \(\alpha_1=50\) successfully tracks the reference with disturbance attenuation below the prescribed \(\gamma\). The analysis illustrates the advantages and effectiveness of explicitly considering nonlinearities in the synthesis of controllers.

**5 Conclusion**
In this paper, the \(H_\infty\) output tracking problem for nonlinear uncertain time-delay systems were investigated, and novel criteria derived for the performance analysis and control design. With a novel Lyapunov-Krasovskii functional based on a delay partition, we improved the piecewise analysis method.
and introduced delay-interval-dependent terms exploiting the delay partitioning subintervals size. The resulting $H_\infty$ output tracking control for nonlinear time-delay systems, also yielded superior results compared to state-of-the-art criteria in the literature. These advantages in terms of conservatism reduction were further illustrated with numerical examples. Two different simulation scenarios were also provided to demonstrate the effectiveness and the importance of the proposed $H_\infty$ output tracking control criterion for nonlinear time-delay systems.

Acknowledgements

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Appendix: Proof of Theorem 1

This appendix presents the proof of Theorem 1. First, we shall take the time derivative of (9) with respect to $t$ along the trajectory of $x(t)$,

$$V_1(t) = d(t) \frac{1}{2} x^T(t) (P_2 - P_1) x(t) + 2x^T(t) (x(t) - \chi(t) \phi_2(t) + P_2(t) d(t)) + (1 - \chi(t) \phi_2(t)) P_2(t) d(t) x(t)$$

$$V_2(t) = x^T(t) (t - t_1) Q x(t - t_1) - \left[ \int_0^{t_1} x^T(t - \eta) Q x(t - \eta) d\eta \right] + \left[ x(t) \frac{T}{2} N_1 x(t) \right]^T N_1 x(t) + \left[ x(t) \frac{T}{2} N_1 x(t) \right]^T N_1 x(t)$$

$$+ \left[ x^T(t - t_1) N_1 x(t - t_1) \right]^T N_1 x(t - t_1) + \left[ x^T(t - t_1) N_1 x(t - t_1) \right]^T N_1 x(t - t_1)$$

$$+ \left[ x^T(t - t_1) N_1 x(t - t_1) \right]^T N_1 x(t - t_1) + \left[ x^T(t - t_1) N_1 x(t - t_1) \right]^T N_1 x(t - t_1)$$

$$- x^T(t - t_1) \left[ (\tau_0 - \frac{t_1}{2}) N_2 x(t - t_1) \right] x(t - t_1) - x^T(t - t_1) \left[ (\tau_0 - \frac{t_1}{2}) N_2 x(t - t_1) \right] x(t - t_1)$$

$$- \sum_{\kappa = 1}^{2} \sum_{\frac{t_1}{2} - \frac{\tau_0}{2} - \frac{k}{2}} \tau_0 x^T(s) S x(s) d\tau - \sum_{\kappa = 0}^{2} \sum_{\frac{t_1}{2} - \frac{\tau_0}{2} - \frac{k}{2}} \tau_0 x^T(s) S x(s) d\tau$$

$$- \int_{-t_1}^{t_1} \dot{x}^T(s) R_1 \dot{x}(s) ds - \int_{-t_1}^{t_1} \dot{x}^T(s) R_2 \dot{x}(s) ds - \chi(t - \tau_1) \int_{-t_1}^{t_1} \dot{x}^T(s) R_1 \dot{x}(s) ds$$

$$- (1 - \chi(t - \tau_1)) \int_{-t_1}^{t_1} \dot{x}^T(s) R_2 \dot{x}(s) ds$$

where $P_1(t)$, $P_2(t)$, $N_1$, and $N_2$ are defined in (10), and $R_1$ in (14). Considering the first subinterval, $\chi(t - \tau_1) = 1$, suppose we expand the integral terms, taking the fact that $\tau_1 \leq t \leq \tau_2$. Applying Jensen’s inequality, [10], and after some manipulation, we obtain

$$V(t)_{|t| < t_2} \leq \xi(t) \left\{ \begin{array}{ll}
\psi(1), & (d(t) - \tau_1) A_{12}; \\
- (d(t) - \tau_2) A_{11}; & (\tau_2 - d(t)) A_{11};
\end{array} \right\} \xi(t),$$

where $\xi(t) := [x^T(t) x^T(t - d(t)) x^T(t - d(t) - \tau_2) x^T(t - d(t) - \tau_1) x^T(t - d(t) - \tau_1)]^T$, $\tau_1 := [\xi(t) \xi_{1d} \xi_{2d}]^T$, $\xi_{1d} := \frac{1}{d(t - \tau_1)} \int_{-t_1}^{t_1} \dot{x}(s) ds$, $\xi_{2d} := \frac{1}{d(t - \tau_2)} \int_{-t_1}^{t_1} \dot{x}(s) ds$ with $\lim_{d(t) - \tau_1} \xi_{1d}, \xi_{2d} = \dot{x}(t - \tau_1)$, and $\lim_{d(t) - \tau_2} \xi_{1d}, \xi_{2d} = \dot{x}(t - \tau_2)$.

Now, from Leibniz-Newton formula for definite integrals with (6), we introduce the following null expressions

$$2 \xi(t) F_1 (x(t) + (B + \Delta B) \dot{x}(t) + \dot{x}_\omega(t) + \dot{\gamma}(t, x(t), x(t - d(t)))) = 0;$$

$$2 \xi(t) F_1 (x(t - t_1) - \gamma(t, t - d(t))) = 0;$$

where $F_1$ is defined in (14). Moreover, the inequality

$$2 \xi(t) F_1 (x(t, t - d(t))) \leq \eta_1 \xi(t) F_1 (x(t, t - d(t))) + \eta_1 \xi(t) F_1 (x(t, t - d(t))) \leq \gamma(t, t - d(t))) = 0,$$

which arises from [20], with (2), and adding the expression:

$$- \dot{T}^T(t) \omega(t) - \gamma^2 \omega^T(t) \omega(t) + \xi(t) F_1 (x(t, t - d(t))) \leq 0,$$

with $F_1 := I_1 (C + \Delta C) + \xi_2 (D + \Delta D)$, we have

$$V(t)|_{|t| < t_2} + e^T(t) e(t) - \gamma^2 \omega^T(t) \omega(t) = \left[ \xi(t) \omega^T(t) \right] \Omega_1 \left[ \xi(t) \omega^T(t) \right]^T,$$

(22)
Lemma, we have the matrices
\[ \dot{V}(t) = \phi(t)+\dot{P}_1+2\dot{F}_1(\Delta A_1 + \Delta B_2R_2) + R_2^T \dot{F}_1 \dot{P}_1 \\right\} \]

Now, we shall consider the matrices that arise from the analysis of \( \Omega_1 \) for \( d(t) \rightarrow \tau_1 \) and \( d(t) \rightarrow \tau_2 \). It is straightforward to conclude that \( \Omega_1 \) may be written as
\[ \begin{cases} \Omega_1 = C_1^T \Omega_1 C_1 \geq \Omega_2 \end{cases} \]
with \( \Omega_1 \) satisfying the conditions in Theorem 1 are satisfied, then
\[ \dot{V}(t) < 0 \]
holds for \( \chi_{1}, \chi_{2} \). Furthermore, using exactly the same arguments of the former case, we may prove that analogous results can be derived for \( \chi_{1}, \chi_{2} \). In this context, it is easy to conclude that if the conditions in the Theorem 1 are satisfied, then \( \dot{V}(t) < 0 \) and \( V(t) \rightarrow 0 \). Since (9) is continuously differentiable, the matrix is still asymptotically stable for \( \omega(t) \equiv 0 \). Moreover, we have that \( \dot{V}(t) = e^T(\dot{t}) - \gamma^2 \omega^T(t) \dot{\omega}(t) < 0 \).

Thus, integrating the inequality, from 0 to \( \infty \), yields
\[ V(\infty) - V(0) = \int_0^\infty ||e(t)||^2 dt < 0 \]
Assuming zero initial conditions, and the positivity of \( V(t) \), \( t \in [0, \infty) \), it is easy to see that \( \int_0^\infty [e^T(t) \dot{e}(t) - \gamma^2 \omega^T(t) \dot{\omega}(t)] dt < 0 \), and thus \( ||e(t)||^2 < \gamma ||\dot{\omega}(t)||^2 \). Therefore, the conditions in Definition 1 are satisfied.

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Robust $H_{\infty}$ output tracking control for nonlinear time-delay systems