Robust and resilient state-dependent control of continuous-time nonlinear systems with general performance criteria

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A novel state-dependent control approach for continuous-time nonlinear systems with general performance criteria is presented in this paper. This controller is optimally robust for model uncertainties and resilient against control feedback gain perturbations in achieving general performance criteria to secure quadratic optimality with inherent asymptotic stability property together with quadratic dissipative type of disturbance reduction. For the system model, unstructured uncertainty description is assumed, which incorporates commonly used types of uncertainties, such as norm-bounded and positive real uncertainties as special cases. By solving a state-dependent linear matrix inequality at each time, sufficient condition for the control solution can be found which satisfies the general performance criteria. The results of this paper unify existing results on nonlinear quadratic regulator, $H_\infty$ and positive real control. The efficacy of the proposed technique is demonstrated by numerical simulations of the nonlinear control of the inverted pendulum on a cart system.

Keywords: robust control; nonlinear systems; optimal control

1. Introduction

Optimal control of nonlinear systems is traditionally characterized in terms of Hamilton Jacobi Equations (HJEs). The solution of the HJEs provides the necessary and sufficient optimal control condition for nonlinear systems. Furthermore, when the controlled system is linear time-invariant and the performance index is linear quadratic regulator (LQR), the HJEs reduced to Algebraic Riccati Equations (AREs). As for $H_\infty$ nonlinear control problem, the optimal control solution is equivalent to solving the corresponding Hamilton Jacobi Inequalities (HJIs) Basar and Bernhard (1995). However, HJEs and HJIs, which are first-order partial differential equations and inequalities, cannot be solved for more than a few state variables. In the past few years, it has been shown that the problems of quadratic regulation and $H_\infty$ nonlinear control can be approached by the state-dependent Riccati equation (SDRE) and nonlinear matrix inequality (NLMI) techniques (Cloutier, 1997; Cloutier, D’Souza, & Mracek, 1996; Huang & Lu, 1996). The state-dependent linear matrix inequality (LMI) control of nonlinear systems, as pointed out in Wang and Yaz (2009), Wang, Yaz, and Jeong (2010), and Wang, Yaz, and Yaz (2010, 2011) synthesizes a controller to achieve mixed nonlinear quadratic regulator (NLQR) and $H_\infty$ control objectives.

Dissipative control for linear systems has also received considerable attention over the past two decades. The concept of dissipative systems was first introduced by Willems (1972a, 1972b), and further generalized by Hill and Moylan (1975, 1976, 1980), playing an important role in systems, circuits and controls. The theory of dissipative systems generalizes the basic tools including the passivity theorem, bounded real lemma, Kalman–Yakubovich lemma and circle criterion. Dissipativity performance includes $H_\infty$ performance, passivity, positive realness and sector-bounded constraint as special cases. Research addressing the problems of $H_\infty$ and positive real control systems can be found in Zhou and Khargonekar (1988) Doyle, Glover, Khargonekar, and Francis (1989), Haddad and Bernstein (1991), Sun, Khargonekar, and Shim (1994), Safonov, Jonckheere, Verma, and Limebeer (1987) and Shim (1996).

Control of uncertain linear systems with $L_2$-bounded structured uncertainty satisfying $H_\infty$ and passivity criteria has been tackled in Khargonekar, Petersen, and Zhou (1990) and Petersen (1987). More recent development involving the quadratic dissipative control for linear systems problem has been tackled in Tan, Soh, and Xie (2000) and Xie, Xie, and De Souza (1998).

In this paper, we further consider the problem of optimal, robust and resilient LMI control of continuous-time
nonlinear systems with general performance criteria. The controller is robust for model uncertainties and resilient for control gain perturbations. As for uncertain nonlinear systems, we consider a general form of $L_2$-bounded uncertainty description, without any standard structure, incorporating commonly used types of uncertainty, such as norm-bounded and positive real uncertainties as special cases. The purpose behind this novel approach is to convert a nonlinear system control problem into a convex optimization problem which is solved by state-dependent LMI. The recent development in convex optimization provides very efficient means for solving LMIs. If a solution can be expressed in a LMI form, then there exist optimization algorithms providing efficient global numerical solutions (Boyd, Ghaoui, Feron, & Balakrishnan, 1994). Therefore, if the LMI is feasible, the LMI control technique provides asymptotically stable solutions satisfying various general performance criteria. We further propose to employ general performance criteria to design the controller guaranteeing the quadratic suboptimal property with inherent stability property in combination to design the controller guaranteeing the quadratic sub-
optimality with inherent stability property.

We further propose to employ general performance criteria for the system model and general performance criteria

2. System model and general performance criteria

The following notation is used in this work: $\mathbb{N}_+$ stands for the set of non-negative real numbers and $\mathbb{R}^n$ stands for the $n$-dimensional Euclidean space. $x \in \mathbb{R}^n$ denotes $n$-dimensional real vector with norm $\|x\|_2 = (x^T x)^{1/2}$, where $(\cdot)^T$ indicates transpose. $\mathbb{R}^{m \times n}$ is the set of $m \times n$ real matrices. $I_n$ is the $n \times n$ identity matrix. $A \succeq 0$ for a symmetric matrix denotes a positive semi-definite matrix. $L_2$ is the space of finite dimensional vectors with finite energy: $\int_0^\infty \|x(t)\|^2 dt < \infty$. Let $L_2^n$ be the extended space of $L_2$ space defined by

$$L_2^n = \{ f : f \text{ is a measurable function : } \mathbb{R}_+ \to \mathbb{R}^n, \text{ with } \times \text{ property that } F_T f \in L_2 \text{ for all finite } T \in \mathbb{R}_+ \}$$

where $F_T f(t) = \begin{cases} f(t), & 0 \leq t \leq T \\ 0, & T < t \end{cases}$ is called the truncation function on $\mathbb{R}_+$ with values in $\mathbb{R}^n$. The inner product in this space is defined as $(u(t), v(t))_T = \int_0^T u(t) v(t) dt$, for $u, v \in L_2^n$.

Consider the following nonlinear dynamical system equation and performance output equation

$$\dot{x} = f(x(t), u(t), w(t)), \quad u(t) = (A(x(t) + \Delta A(x, t)x(t) + (B(x, t) + \Delta B(x, t))u(t) + (E(x, t) + \Delta E(x, t))w(t),$$

$$= (A + \Delta A)x + (B + \Delta B)u + (E + \Delta E)w, \quad z(t) = g(x(t), w(t)) = C \cdot x + D \cdot w,$$

where $x(t) \in \mathbb{R}^n$ is the state variable of the dynamical system, $u(t) \in \mathbb{R}^m$ the applied input, $w(t) \in \mathbb{R}^p$ the $L_2$ type of disturbance, $z(t) \in \mathbb{R}^r$ the performance output function, $f, g$ the smooth real vector functions, $A \in \mathbb{R}^{n \times n}, B \in \mathbb{R}^{n \times m}, E \in \mathbb{R}^{r \times p}, C \in \mathbb{R}^{r \times n}$ and $D \in \mathbb{R}^{r \times p}$ the state-dependent coefficient matrices, and $\Delta A \in \mathbb{R}^{n \times n}, \Delta B \in \mathbb{R}^{n \times m}$ and $\Delta E \in \mathbb{R}^{r \times p}$ the time-varying uncertainty matrices.

It is assumed that the state feedback is available and the state feedback control input is given by

$$u(t) = (K(x(t) + \Delta K(x, t))x(t) = (K + \Delta K)x.$$ (3)

Introducing the following quadratic energy supply function $E: L_2^n \times L_2^n \times \mathbb{R}^+ \to \mathbb{R}$ associated with the system equations, defined in Hill and Moylan (1975, 1976, 1980):

$$E(z, w, T) = \langle z, Qz \rangle_T + 2\langle z, Sw \rangle_T + \langle w, Rw \rangle_T, \quad (4)$$

where $Q \in \mathbb{R}^{r \times r}, S \in \mathbb{R}^{r \times p}, R \in \mathbb{R}^{p \times p}$ are the chosen weighting matrices. Next, from the definition of dissipativity (Hill & Moylan 1975, 1976, 1980), we have

**Definition 1** Given matrices $Q \in \mathbb{R}^{r \times r}, S \in \mathbb{R}^{r \times p}, R \in \mathbb{R}^{p \times p}$ with $Q, R$ symmetric, the system (1) and (2) with energy function (4) is said to be $(Q, S, R)$ dissipative if for some real function $\beta(\cdot)$ with $\beta(0) = 0$. The physical meaning for $\beta(\cdot)$ is the stored energy at the initial time, given by the initial condition $x_0$.

$$E(z, w, T) + \beta(x_0) \geq 0, \forall w \in L_2, \forall T \geq 0.$$ (5)

Furthermore, if for some scalar $\alpha > 0$,

$$E(z, w, T) + \beta(x_0) \geq \alpha \langle w, w \rangle_T, \forall w \in L_2, \forall T \geq 0.$$ (6)

The system (1) and (2) is said to be strictly $(Q, S, R)$ dissipative.

**Theorem 1** Consider the quadratic function $V = x^T P x > 0$, matrices $Q \in \mathbb{R}^{r \times r}, S \in \mathbb{R}^{r \times p}, R \in \mathbb{R}^{p \times p}$ with $Q, R$ symmetric, $M \in \mathbb{R}^{m \times m}, M > 0, N \in \mathbb{R}^{m \times n}, N > 0$ with $M, N$ symmetric, the control of nonlinear system (1) and (2) will achieve mixed NLQR and dissipative performance if the following condition holds:

$$\dot{V} + x^T M x + u^T N u - (z^T Qz + 2z^T Sw + w^T Rw) < 0, \forall T \geq 0.$$ (7)
Proof By integrating Equation (7) from 0 to $T$, we have

$$\int_0^T (\varepsilon T Q z + 2 z^T S w + w^T R w) \, dt > 0,$$

$$\int_0^T x^T M x \, dt + \int_0^T u^T N u \, dt + V(x(T)) - V(x(0)), \forall T \geq 0.$$  \hfill (8)

Let $\beta(x_0) = V(x(0)), V(x) = x^T P x, V(x(T)) \geq 0$, Equation (8) implies

$$\int_0^T (\varepsilon T Q z + 2 z^T S w + w^T R w) \, dt + \beta(x_0) > 0, \forall T \geq 0,$$

which is the condition of $(Q,S,R)$ dissipativity. By adding the terms $x^T M x + u^T N u$, we include the NLQR control performance into the original $(Q,S,R)$-dissipative criteria. \hfill \blacksquare

Remark 1 Notice that both $H_\infty$ and passivity are special cases of $(Q,S,R)$ dissipativity. The special cases are summarized as follows:

Case 1 $Q = -I, S = 0, R = \gamma^2 I$, the strict $(Q, S, R)$ dissipativity reduces to $H_\infty$ design (Doyle et al., 1989). The overall control design satisfies mixed NLQR–$H_\infty$ performance.

Case 2 $Q = 0, S = I, R = 0$, the strict $(Q, S, R)$ dissipativity reduces to strict positive realness (Sun et al., 1994). The overall control design satisfies mixed NLQR–strict positive realness performance.

Case 3 $Q = -\theta I, S = (1 - \theta)I, R = \theta \gamma^2 I$, the strict $(Q, S, R)$ dissipativity reduces to mixed $H_\infty$ and positive real design, when $\theta \in (0,1)$. The overall control design satisfies mixed NLQR–$H_\infty$–positive real performance.

Case 4 $Q = -I, S = \frac{1}{2} (K_1 + K_2)^T, R = -\frac{1}{2} (K_1^T K_2 + K_2^T K_1)$, where $K_1$ and $K_2$ are constant matrices of appropriate dimensions, the strict $(Q, S, R)$ dissipativity reduces to a sector-bounded constraint (Gupta & Joshi, 1994). The overall control design satisfies mixed NLQR–sector-bounded constraint performance.

Before introducing the main result of the paper, the following model of unstructured uncertainties is introduced.

ASSUMPTION 1 The following general form of $L_2$-bounded unstructured uncertainties is considered:

$$\Delta_A \Delta_A^T \leq \gamma_A I,$$

$$\Delta_B \Delta_B^T \leq \gamma_B I,$$

$$\Delta_E \Delta_E^T \leq \gamma_E I,$$

$$\Delta_K \Delta_K^T \leq \gamma_K I,$$

\hfill (10)

for $\forall x \in \mathbb{R}^n$ and $t \geq 0$.

3. State-dependent LMI control

LEMMA 1

$$AB^T + BA^T \leq \alpha A A^T + \alpha^{-1} B B^T.$$ \hfill (11)

This can be proven easily by considering

$$(a^{1/2} A - \alpha^{-1/2} B)(a^{1/2} A - \alpha^{-1/2} B)^T \geq 0.$$ \hfill (12)

Also, by choosing $A$ and $B$ matrices as $A = \begin{bmatrix} 0 & 0 \\ b^T a & 0 \end{bmatrix}$, we have

$$\begin{bmatrix} 0 & a^T b \\ b^T a & 0 \end{bmatrix} \leq \begin{bmatrix} \xi a^T a & 0 \\ 0 & \xi^{-1} b^T b \end{bmatrix}.$$ \hfill (13)

LEMMA 2 Denote $X = P^{-1}$ for positive definite matrix $P > 0$, then the following equality always holds:

$$X P X = -\dot{X}.$$ \hfill (14)

This can be proven easily by considering

$$0 = \frac{d}{dt} (I) = \frac{d}{dt} (P P^{-1}) = \frac{d}{dt} (P) \cdot P^{-1} + P \cdot \frac{d}{dt} (P^{-1}).$$ \hfill (15)

Remark 2 Since in the presentation below, $P$ will be used to describe the energy content, which needs to decrease due to the asymptotic stability requirement, $X = P^{-1}$ matrix will be increasing in time. Therefore, we have $X \dot{P} X = -\dot{X} < 0$.

The following theorem summarizes the main results of the paper:

THEOREM 2 Given the system Equation (1), performance output Equation (2) and control Equation (3), if there exist matrices $X = P^{-1} > 0$ and $Y$ for all $t > 0$, such that the following state-dependent LMI holds:

$$\begin{bmatrix} \gamma_{11} & \gamma_{12} & X & Y^T \\ * & \gamma_{22} & 0 & 0 \\ * & * & \gamma_{33} & 0 \\ * & * & * & \gamma_{44} \end{bmatrix} < 0,$$ \hfill (16)

where

$$\gamma_{11} = X A^T + AX + Y^T B^T + BY + BB^T + [\gamma_A + 2 \gamma_B + \gamma_E I],$$

$$\gamma_{12} = E - X C^T Q D - X C^T S,$$

$$\gamma_{22} = -D^T Q D - 2 D^T S - R + I,$$

$$\gamma_{33} = -[I + [3 + \lambda_{\max}(N)] \gamma_k I + M + C^T Q C]^{-1},$$

$$\gamma_{44} = -[I + N^2 + N]^{-1}.$$ \hfill (17)

Then the inequality (7), which guarantees mixed NLQR and dissipative performance, is satisfied. The nonlinear feedback control gain is given by

$$K = Y \cdot P.$$ \hfill (18)
Applying Lemma 2 and Remark 2, we have formed as follows:

\[ x^T(A + \Delta_A + (B + \Delta_B)(K + \Delta_K))^TPx + w^T[E + \Delta_E]^TPx + x^TP[A + \Delta_A + (B + \Delta_B)(K + \Delta_K)]x + x^TP[E + \Delta_E]w + x^TPx + x^TMx + x^T[K + \Delta_K]^TN[K + \Delta_K]x - \{Cx + Dw\}^TQ[\{Cx + Dw\} - 2\{Cx + Dw\}^TSw - w^TRw < 0. \]  

(19)

Equivalently,

\[ [x^T \ w^T] \Psi \begin{bmatrix} x \ w \end{bmatrix} = [x^T \ w^T] \begin{bmatrix} \Psi_{11} & \Psi_{12} \\ \Psi_{21} & \Psi_{22} \end{bmatrix} \begin{bmatrix} x \\ w \end{bmatrix} < 0, \]

(20)

where

\[ \Psi_{11} = \hat{P} + [A + \Delta_A + (B + \Delta_B)(K + \Delta_K)]^TP + P[A + \Delta_A + (B + \Delta_B)(K + \Delta_K)] + M + [K + \Delta_K]^TN[K + \Delta_K] + C^TQC, \]

\[ \Psi_{12} = P[E + \Delta_E] - C^TQD - C^TS, \]

(21)

\[ \Psi_{22} = -D^TQD - 2D^TS - R. \]

Pre-multiplying and post-multiplying the matrix \( \Psi \) with the block \( \text{diag}\{X, I\} \), where \( X = P^{-1}, Y = K \cdot P^{-1} = KX \). Then the following matrix inequality holds:

\[ \begin{bmatrix} \Lambda_{11} & \Lambda_{12} \\ * & \Lambda_{22} \end{bmatrix} < 0, \]

(22)

where

\[ \Lambda_{11} = X[A + \Delta_A + (B + \Delta_B)(K + \Delta_K)]^T + [A + \Delta_A + (B + \Delta_B)(K + \Delta_K)]X + X\hat{P}X + XMX + X[K + \Delta_K]^TN[K + \Delta_K]X + XCTQCX, \]

\[ \Lambda_{12} = [E + \Delta_E] - XCTQD - XC^TS, \]

\[ \Lambda_{22} = -D^TQD - 2D^TS - R. \]

(23)

Applying Lemma 2 and Remark 2, we have \( X\hat{P}X < 0. \)

Denote \( W = XA^T + AX + Y^TB^T + BY. \)

The sufficient condition for matrix inequality (22) to be held is to change term \( \Lambda_{11} \) as follows:

\[ \Lambda_{11} = XA^T + AX + Y^TB^T + BY + X(\Delta_A + \Delta_BK + B\Delta_K) \]

\[ + \Delta_B\Delta_K^T + [\Delta_A + \Delta_BK + B\Delta_K + \Delta_B\Delta_K]X \]

\[ + XMX + Y^TN\Delta_KX + X^TNCX + X\Delta_K^T\Delta_KX \]

\[ = W + [\Delta_A + \Delta_BK]X + X(\Delta_A + \Delta_BK)^T \]

\[ + X(B\Delta_K + \Delta_B\Delta_K)^T + [\Delta_K + \Delta_B\Delta_K]X \]

\[ + [X\Delta_K^TNY + Y^TN\Delta_KX] + [XM + Y^TN] \]

\[ + X\Delta_K^TN\Delta_KX + X^TNCX, \]

(25)

By applying Lemma 1 to Equation (25) and using Assumption 1, we obtain

\[ \{\Delta_A + \Delta_BK\}X + X(\Delta_A + \Delta_BK)^T = X[I \ K^T][\Delta_A \ \Delta_B]\]

\[ \leq \alpha_1[\Delta_A \ \Delta_B] + \alpha_1^{-1}X[I \ K^T][I \ K]X \]

\[ \leq \alpha_1(\gamma_A + \gamma_B)I + \alpha_1^{-1}X[I \ K^T][I \ K]X, \]

\[ X\Delta_K^TB^T + B\Delta_KX \leq \alpha_2X\Delta_K^T\Delta_KX \]

\[ + \alpha_2^{-1}BB^T \leq \alpha_2\gamma_KX^2 + \alpha_2^{-1}BB^T, \]

\[ X\Delta_K^T\Delta_B^T + \Delta_B\Delta_KX \leq \alpha_3X\Delta_K^T\Delta_KX \]

\[ + \alpha_3^{-1}\Delta_B\Delta_K^T \leq \alpha_3\gamma_KX^2 + \alpha_3^{-1}\gamma_BI, \]

\[ X\Delta_K^TNY + Y^TN\Delta_KX \leq \alpha_4X\Delta_K^T\Delta_KX \]

\[ + \alpha_4^{-1}Y^TN^2Y \leq \alpha_4\gamma_KX^2 + \alpha_4^{-1}Y^TN^2Y, \]

\[ X\Delta_K^TN\Delta_KX \leq X\Delta_K^T\Delta_KX \cdot \lambda_{\text{max}}(N) \leq \lambda_{\text{max}}(N)\gamma_KX^2. \]

(26)

Therefore, we have

\[ \Lambda_{11} \leq W + \alpha_1(\gamma_A + \gamma_B)I + \alpha_1^{-1}X[I \ K^T][I \ K]X \]

\[ + \alpha_2\gamma_KX^2 + \alpha_2^{-1}BB^T + \alpha_3\gamma_KX^2 + \alpha_3^{-1}\gamma_BI \]

\[ + \alpha_4\gamma_KX^2 + \alpha_4^{-1}Y^TN^2Y + \{XM + Y^TN\} \]

\[ + \lambda_{\text{max}}(N)\gamma_KX^2 + X^TNCX, \]

(27)

\[ \Lambda_{12} = [E + \Delta_E] - XCTQD - XC^TS, \]

\[ \Lambda_{22} = -D^TQD - 2D^TS - R. \]

(28)

Using Lemma 1 and Assumption 1, we have

\[ \begin{bmatrix} 0 \ \Delta_E \\ \Delta_E^T \ 0 \end{bmatrix} \preceq \begin{bmatrix} \alpha_2\gamma_K \alpha_2^{-1} & \alpha_3 \alpha_3^{-1} \\ \alpha_3^{-1} \alpha_3 \ & \alpha_4 \alpha_4^{-1} \end{bmatrix} \preceq \begin{bmatrix} \alpha_5 \gamma_B & 0 \\ 0 & \alpha_6 \alpha_6^{-1} \end{bmatrix}. \]

(29)
Therefore, by applying the results of Equations (24)–(29) to Equation (22), we find:

$$\begin{bmatrix} \Theta_{11} & \Theta_{12} \\ \ast & \Theta_{22} \end{bmatrix} < 0,$$

where

$$\begin{align*}
\Theta_{11} &= W + [\alpha_1 \gamma_A + \alpha_1 \gamma_B + \alpha_3 \gamma_E + \alpha_3^{-1} \gamma_B]I \\
&\quad + \alpha_1^{-1} X [I \ K^T \ F] X + \alpha_4^{-1} Y T N^2 Y \\
&\quad + [\alpha_2 + \alpha_3 + \alpha_4 + \lambda_{\text{max}}(N)] \gamma_k X^2 \\
&\quad + \alpha_2^{-1} BB^T + \{XMX + Y^T NY + XC^T QCX\} \\
&= W + [\alpha_1 \gamma_A + \alpha_1 \gamma_B + \alpha_3 \gamma_E + \alpha_3^{-1} \gamma_B]I \\
&\quad + X[\alpha_1^{-1} I + [\alpha_2 + \alpha_3 + \alpha_4 + \lambda_{\text{max}}(N)] \gamma_k I \\
&\quad + M + C^T QC \} X \\
&\quad + Y^T \{\alpha_1^{-1} + \alpha_4^{-1} N^2 + N\} Y + \alpha_2^{-1} BB^T,
\end{align*}$$

$$\begin{align*}
\Theta_{12} &= E - XC^T QD - XC^T S, \\
\Theta_{22} &= -D^T QD - 2D^T S - R + \alpha_5^{-1} I.
\end{align*}$$

By applying the Schur complement, we conclude the final LMI solution as

$$\begin{bmatrix} \Upsilon_{11} & \Upsilon_{12} & X & Y^T \\ \ast & \Upsilon_{22} & 0 & 0 \\ \ast & \ast & \Upsilon_{33} & 0 \\ \ast & \ast & \ast & \Upsilon_{44} \end{bmatrix} < 0,$$

where

$$\begin{align*}
\Upsilon_{11} &= W + [\alpha_1 \gamma_A + \alpha_1 \gamma_B + \alpha_3 \gamma_E + \alpha_3^{-1} \gamma_B]I + \alpha_2^{-1} BB^T \\
&= XA^T + AX + YB^T + BY + \alpha_2^{-1} BB^T \\
&\quad + [\alpha_1 \gamma_A + \alpha_1 \gamma_B + \alpha_3 \gamma_E + \alpha_3^{-1} \gamma_B]I, \\
\Upsilon_{12} &= E - XC^T QD - XC^T S, \\
\Upsilon_{22} &= -D^T QD - 2D^T S - R + \alpha_5^{-1} I, \\
\Upsilon_{33} &= -[\alpha_1^{-1} I + [\alpha_2 + \alpha_3 + \alpha_4 + \lambda_{\text{max}}(N)] \gamma_k I \\
&\quad + M + C^T QC]^{-1}, \\
\Upsilon_{44} &= -[\alpha_1^{-1} + \alpha_4^{-1} N^2 + N]^{-1}.
\end{align*}$$

Since positive constants $\alpha_1, \ldots, \alpha_5$ are arbitrary, choosing all of them as 1, we obtain Equation (16). Therefore, if LMI (16) holds, the inequality (7) is satisfied. This concludes the proof of the theorem.

**Remark 3** At this point, it is to be noted that other choices of positive constants $\alpha_1, \ldots, \alpha_5$ are possible and can be tried if the value 1 for all these constants does not work.

4. **Application to the inverted pendulum on a cart**

The inverted pendulum on a cart problem (Wang & Yaz, 2010) is used for testing the novel robust and resilient state-dependent LMI approach to compare the performance. Using the Euler–Lagrange equation technique, the complete equations of motion for the inverted pendulum on a cart can be reached as

$$(M + m)\ddot{x} + b\ddot{x} + mL\dot{\theta} \cos(\theta) - mL\dot{\theta} \sin(\theta) = F,$$

$$(I + mL^2)\ddot{\theta} + mgL \sin(\theta) + mL\dot{\theta} \cos(\theta) = 0.$$  (34)

The following system parameters are assumed:

- $M = 0.5$ kg, $m = 0.5$ kg, $b = 0.1N \frac{s}{m}$,
- $L = 0.3$ m, $I = 0.06$ kg m$^2$,
- Sampling time: $T = 0.01$ s.

Denote the following state variables:

$$x_1 = x(t), x_2 = \dot{x}(t), x_3 = \theta(t), x_4 = \dot{\theta}(t).$$

The following initial conditions are assumed:

$$x_1 = 1, x_2 = 0, x_3 = \pi/4, x_4 = 0.$$

The following design parameters are chosen to satisfy different mixed performance criteria:

- Mixed NLQR–$H_\infty$ design (predominant NLQR)

$$C = \begin{bmatrix} 0.01 & 0.01 & 0.01 & 0.01 \end{bmatrix}, D = [0.01], M = I_4, N = 1, Q = 1, S = 0, R = -1.$$

- Mixed NLQR–$H_\infty$ design (predominant $H_\infty$)

$$C = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}, D = [1], M = 0.01 \times I_4, N = 0.01, Q = 1, S = 0, R = -10.$$

- Mixed NLQR–$H_\infty$–positive real design (NLQR passivity)

$$C = \begin{bmatrix} 1 & 1 & 1 \end{bmatrix}, D = [1], M = I_4, N = 1, Q = 0.01, S = 1, R = 0.01.$$

All of the above mixed criteria control performance results are shown in Figures 1–5, in comparison with the traditional LQR technique based on linearization. From these figures, we find that the novel state-dependent LMI control has better performance compared with the traditional LQR technique based on linearization. Especially, Figures 3 and 4 show that the traditional LQR technique loses control of the angle and angular velocities of the pendulum, respectively. Figure 5 shows that the highest magnitude of control is needed by the predominant $H_\infty$ control and the lowest control magnitude is needed by the linearization-based LQR technique.
5. Conclusions

This paper addresses nonlinear system control design with general NLQR and quadratic dissipative criteria to achieve asymptotic stability, quadratic optimality and strict quadratic dissipativeness. For systems with unstructured but bounded uncertainty, the LMI-based sufficient conditions are derived for the solution of general performance criteria control. Our results unify the existing results on SDRE control, robust $H_{\infty}$, positive real control and sector-bounded control. The relative weighting matrices of these criteria can be achieved by choosing different weighing coefficient matrices. The optimal control can be obtained by solving LMI at each time step. The inverted pendulum on a cart control, which is a benchmark under-actuated nonlinear control system, is used as an example to demonstrate its effectiveness and robustness of the proposed method. The simulation studies show that the proposed method provides a satisfactory alternative to the existing nonlinear control approaches.
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