

\mathcal{H}_∞ Filtering of Uncertain LPV Systems with Time-Delay

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- Stability of Uncertain LPV Systems with Delays
- The filtering Problem
- Conclusion and Future Works

Introduction

- Considered Systems
- Filters Structures

- Uncertain LPV Systems with Delay

$$\begin{bmatrix} \dot{x}(t) \\ z(t) \\ y(t) \end{bmatrix} = \Sigma(\rho(t), \delta) \begin{bmatrix} x(t) \\ x(t-h(t)) \\ w(t) \end{bmatrix}$$

$$x(\theta) = \phi(\theta), \theta \in [-h_M, 0]$$

$$\rho \in U_\rho$$

$$\dot{\rho} \in \text{hull}[U_\nu]$$

$$\delta \in U_\delta$$

$$h(t) \in [0, h_M]$$

$$\dot{h}(t) < \mu < 1$$

Considered Systems (2)

- System Matrix :

$$\Sigma = \begin{bmatrix} A(\rho) + \Delta A(\rho, \delta) & A_h(\rho) + \Delta A_h(\rho, \delta) & E(\rho) + \Delta E(\rho, \delta) \\ C(\rho) & C_h(\rho) & F(\rho) \\ C_y(\rho) + \Delta C_y(\rho, \delta) & C_{yh}(\rho) + \Delta C_{yh}(\rho, \delta) & F_y(\rho) + \Delta F_y(\rho, \delta) \end{bmatrix}$$

- where the uncertain part obeys

$$\begin{bmatrix} \Delta A & \Delta A_h & \Delta E \\ \Delta C_y & \Delta C_{yh} & \Delta F_y \end{bmatrix}(\rho, \delta) = \begin{bmatrix} H_0 \\ H_1 \end{bmatrix}(\rho) \Delta(\delta) \begin{bmatrix} F_0 & F_1 & F_2 \\ F_3 & F_4 & F_5 \end{bmatrix}(\rho)$$

with $\|\Delta\|_2 \leq 1$

- Filters with memory

$$\begin{bmatrix} \dot{x}_F(t) \\ z_F(t) \end{bmatrix} = \begin{bmatrix} A_F(\rho) & A_{hF}(\rho) & B_F(\rho) \\ C_F(\rho) & C_{hF}(\rho) & D_F(\rho) \end{bmatrix} \begin{bmatrix} x_F(t) \\ x_F(t-h(t)) \\ y(t) \end{bmatrix}$$

- Memoryless filters

$$\begin{bmatrix} \dot{x}_F(t) \\ z_F(t) \end{bmatrix} = \begin{bmatrix} A_F(\rho) & B_F(\rho) \\ C_F(\rho) & D_F(\rho) \end{bmatrix} \begin{bmatrix} x_F(t) \\ y(t) \end{bmatrix}$$

- These matrices are aimed to be chosen such that

$$\|z - z_F\|_{\mathcal{L}_2} \leq \gamma \|w\|_{\mathcal{L}_2}$$

with a minimal \mathcal{L}_2 -gain $\gamma > 0$.

Stability of Uncertain LPV Systems with Delays

- Lyapunov-Krasovskii Functional
- Asymptotic Stability Theorem
- Relaxed Version

Lyapunov-Krasovskii Functional (1)

- Delay-Dependent LKF :

$$V(x_t, \dot{x}_t, \rho) = V_1(x_t, \rho) + V_2(x_t) + V_3(\dot{x}_t)$$

$$V_1(x_t, \rho) = x(t)^T P(\rho)x(t)$$

$$V_2(x_t) = \int_{t-h(t)}^t x(\theta)^T Qx(\theta)d\theta$$

$$V_3(\dot{x}_t) = h_M \int_{-h_M}^0 \int_{t+\theta}^t \dot{x}(\eta)^T R\dot{x}(\eta)d\eta d\theta$$

- whose derivative along the trajectories solution of the system satisfies :

$$\dot{V}_1 = \dot{x}(t)^T P(\rho)x(t) + x(t)^T P(\rho)\dot{x}(t) + x(t)^T \left(\sum_i \dot{\rho}_i \frac{\partial}{\partial \rho_i} P(\rho) \right) x(t)$$

$$\dot{V}_2 \leq x(t)^T Qx(t) - (1 - \mu)x(t - h(t))^T Qx(t - h(t))$$

$$\dot{V}_3 \leq h_M^2 \dot{x}(t)^T R\dot{x}(t) - h_M \int_{t-h(t)}^t \dot{x}(\theta)^T R\dot{x}(\theta)d\theta$$

Lyapunov-Krasovskii Functional (2)

- Using Jensen's Inequality :

$$-h_M \int_{t-h(t)}^t \dot{x}(\theta)^T R \dot{x}(\theta) d\theta \leq - \left(\int_{t-h(t)}^t \dot{x}(\theta) d\theta \right)^T R \left(\int_{t-h(t)}^t \dot{x}(\theta) d\theta \right)$$

- we get

$$\dot{V} \leq \chi(t)^T \begin{bmatrix} \Psi + h_M^2 A^T R A & P A_h + R + h_M^2 A^T R A_h \\ \star & -(1-\mu)Q - R + h_M^2 A_h^T R A_h \end{bmatrix} \chi(t)$$

with $\Psi = A^T P + P A + Q - R + \sum_i \dot{\rho}_i \frac{\partial}{\partial \rho_i} P(\rho)$ and
 $\chi(t) = \text{col}(x(t), x(t-h(t)))$.

Theorem

The uncertain LPV system is asymptotically stable for all $h(t) \in [0, h_M]$ such that $\dot{h}(t) < \mu$ if there exist $P : U_\rho \rightarrow \mathbb{S}_{++}^n$, $Q, R \in \mathbb{S}_{++}^n$ such that the LMI

$$\begin{bmatrix} \Psi(\rho, \nu) & P(\rho)A_h(\rho) + R & h_M A(\rho)^T R \\ \star & -(1 - \mu)Q - R & h_M A_h(\rho)^T R \\ \star & \star & -R \end{bmatrix} \prec 0$$

holds for all $(\rho, \nu) \in U_\rho \times U_\nu$ with

$$\Psi(\rho, \nu) = A(\rho)^T P(\rho) + P(\rho)A(\rho) + Q - R + \sum_i \nu_i \frac{\partial}{\partial \rho_i} P(\rho).$$

Bounded Real Lemma

Theorem

The uncertain LPV system is asymptotically stable for all $h(t) \in [0, h_M]$ such that $\dot{h}(t) < \mu$ if there exist $P : U_\rho \rightarrow \mathbb{S}_{++}^n$, $Q, R \in \mathbb{S}_{++}^n$ and $\gamma > 0$ such that the LMI

$$\begin{bmatrix} \Psi(\rho, \nu) & P(\rho)A_h(\rho) + R & P(\rho)E(\rho) & C(\rho)^T & h_M A(\rho)^T R \\ * & -(1 - \mu)Q - R & 0 & C_h(\rho)^T & h_M A_h(\rho)^T R \\ * & * & -\gamma I & F(\rho)^T & h_M E(\rho)^T R \\ * & * & * & -\gamma I & 0 \\ * & * & * & * & -R \end{bmatrix} \prec 0$$

holds for all $(\rho, \nu) \in U_\rho \times U_\nu$ with

$\Psi(\rho, \nu) = A(\rho)^T P(\rho) + P(\rho)A(\rho) + Q - R + \sum_i \nu_i \frac{\partial}{\partial \rho_i} P(\rho)$. Moreover, we have $\|z\|_{\mathcal{L}_2} \leq \gamma \|w\|_{\mathcal{L}_2}$.

Theorem

The uncertain LPV system is asymptotically stable for all $h(t) \in [0, h_M]$ such that $\dot{h}(t) < \mu$ if there exist $P : U_\rho \rightarrow \mathbb{S}_{++}^n$, $X : U_\rho \rightarrow \mathbb{R}^{n \times n}$, $Q, R \in \mathbb{S}_{++}^n$ and $\gamma > 0$ such that the LMI

$$\begin{bmatrix} -(X + X^T) & P + X^T A & X^T A_h & X^T E & 0 & X^T & h_M R \\ * & \Phi_1 & R & 0 & C^T & 0 & 0 \\ * & * & \Phi_2 & 0 & C_h^T & 0 & 0 \\ * & * & * & -\gamma I & F^T & 0 & 0 \\ * & * & * & * & -\gamma I & 0 & 0 \\ * & * & * & * & * & -P & -h_M R \\ * & * & * & * & * & * & -R \end{bmatrix} \prec 0$$

holds for all $(\rho, \nu) \in U_\rho \times U_\nu$ with $\Phi_1 = P + Q - R + \sum_i \nu_i \frac{\partial}{\partial \rho_i} P(\rho)$ and $\Phi_2 = -(1 - \mu)Q - R$. Moreover, we have $\|z\|_{\mathcal{L}_2} \leq \gamma \|w\|_{\mathcal{L}_2}$.

Filtering Problem

- Augmented System
- Relaxation
- Example

- Interconnection between the system and filter

$$\dot{x}_a(t) = \mathcal{A}x_a(t) + \mathcal{A}_h x_a(t - h(t)) + \mathcal{E}w(t)$$

$$z_e(t) = \mathcal{C}x_a(t) + \mathcal{C}_h x_a(t - h(t)) + \mathcal{F}w(t)$$

$$x_a(t) = \text{col}(x(t), x_F(t))$$

$$z_e(t) = z(t) - z_F(t)$$

- with

$$\begin{aligned} \mathcal{A} &= \begin{bmatrix} A & 0 \\ A - B_F C_y & A_F \end{bmatrix} & \mathcal{A}_h &= \begin{bmatrix} A_h & 0 \\ A_h - B_F C_{yh} & A_{Fh} \end{bmatrix} & \mathcal{E} &= \begin{bmatrix} E \\ E - B_F F_y \end{bmatrix} \\ \mathcal{C} &= \begin{bmatrix} C - D_F C_y & C_F \\ C_F & C_F \end{bmatrix} & \mathcal{C}_h &= \begin{bmatrix} C_h - D_F C_{yh} & C_{Fh} \\ C_{Fh} & C_{Fh} \end{bmatrix} \\ \mathcal{F} &= F - D_F F_y \end{aligned}$$

Relaxation of Bilinear Terms

- Bilinear Terms $X^T \mathcal{A}$, $X^T \mathcal{A}_h$, $X^T \mathcal{E}$

$$\begin{aligned} X^T \mathcal{A} &= \begin{bmatrix} X_1^T & X_3^T \\ X_2^T & X_4^T \end{bmatrix} \begin{bmatrix} A & 0 \\ A - B_F C_y & A_F \end{bmatrix} \\ &= \begin{bmatrix} (X_1 + X_3)^T A - X_3^T B_F C_y & X_3^T A_F \\ (X_2 + X_4)^T A - X_4^T B_F C_y & X_4^T A_F \end{bmatrix} \end{aligned}$$

- Set $X_4 = X_3$ (both system and filter have the same order)
- Linearization

$$[\tilde{A}_F \quad \tilde{A}_{hF} \quad \tilde{B}_F] = X_3^T [A_F \quad A_{hF} \quad B_F]$$

- we get a LMI problem

Example 1

- Let

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

$$z(t) = \begin{bmatrix} 1 & 2 \end{bmatrix} x(t)$$

$$y(t) = \begin{bmatrix} 1 & 0 \end{bmatrix} x(t)$$

- We set $h_M = 1$ and we study γ w.r.t. μ using a memoryless filter

μ	0	0.4	0.8
Fridman [2003]	1.4086	1.8311	15.8414
This result	0.06484	0.10651	0.48661

Example 2(1)

- We consider the LPV system

$$\dot{x}(t) = \begin{bmatrix} 0 & 1 + 0.2\rho \\ -2 & -3 + 0.1\rho \end{bmatrix} x(t) + \begin{bmatrix} 0.2\rho & 0.1 \\ -0.2 + 0.1\rho & -0.3 \end{bmatrix} x(t - h(t)) \\ + \begin{bmatrix} -0.2 \\ -0.2 \end{bmatrix} w(t)$$

$$z(t) = \begin{bmatrix} 0.3 & 1.5 \\ -0.45 & 0.75 \end{bmatrix} x(t) + \begin{bmatrix} 0.5\rho \\ -0.5\rho \end{bmatrix} w(t)$$

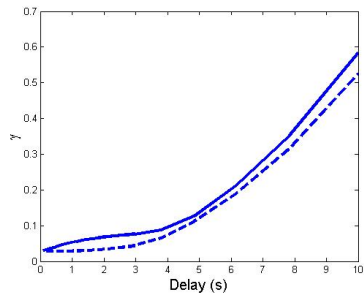
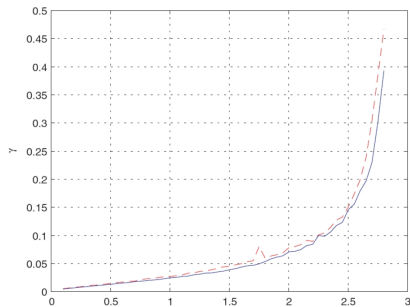
$$y(t) = \begin{bmatrix} 0 & 1 \\ 0.5 & 0 \end{bmatrix} x(t) + \begin{bmatrix} 0 \\ 1 + 0.1\rho \end{bmatrix} w(t)$$

$$\rho \in [-1, 1] \quad \dot{\rho} \in [-1, 1]$$

- we choose $P(\rho) = P_0 + P_1\rho$ and we study γ w.r.t. the delay bound h_M

Example 2(2)

- We get the following figures

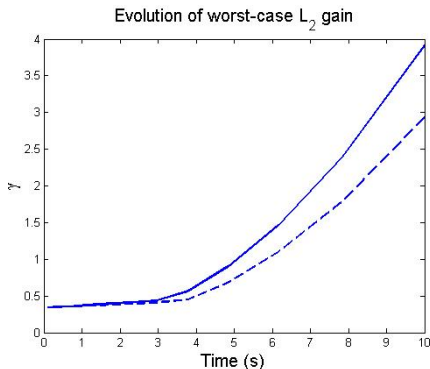


Example 2(3)

- Adding uncertainties

$$H_0 = H_1 = 0.1I, \quad F_0 = F_1 = F_3 = F_4 = I$$
$$F_2 = F_5 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

- We get



Conclusion et Future Works

Conclusion et Future Works

- Advantages (Stability/Performance Analysis) :
 - Simple and Fast
 - Interesting results but still conservative despite of the use of the Jensen's inequality.
- Use a more complex LKF, e.g.

$$V_2 = \sum_{i=1}^N \int_{t-ih_n(t)}^{t-(i-1)h_n(t)} x(\theta)^T Q_i x(\theta) d\theta, \quad h_n(t) = h(t)/N$$

$$V_3 = \bar{h} \sum_{i=1}^N \int_{t-i\bar{h}}^{t-(i-1)\bar{h}} \int_{t+\theta}^t \dot{x}(\eta)^T R_i \dot{x}(\eta) d\eta d\theta, \quad \bar{h} = h_M/N$$

- Tackle the delay knowledge uncertainty