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Output Feedback Control of a Class of Stochastic Hybrid Systems[★]

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Abstract

This paper deals with static output feedback control of a class of reconfigurable systems with Markovian Parameters and state-dependent noise. The main contribution is to formulate conditions for multi-performance design related to this class of stochastic hybrid systems. The specifications and objectives under consideration include stochastic stability, $\mathcal{H}_2$ and $\mathcal{H}_\infty$ performances. Another problematic related to a more general class of stochastic hybrid systems, known as Markovian Jump Linear Systems (MJLS), is also addressed. This problematic concerns the mode-independent output feedback control of MJLS. The obtained results are illustrated on a numerical example.

Key words: Reconfigurable Control; Markovian Jumping Parameters; $\mathcal{H}_2/\mathcal{H}_\infty$ Control; Static Output Feedback.

1 Introduction

Fault tolerant control systems (FTCS) aim at guaranteeing that a process keeps fulfilling its mission even in the presence of failures, although possibly in a degraded mode. For the design problem, two basic approaches have been studied: passive fault-tolerance rely on robust control principles where the faults are considered as modeling uncertainties. Active fault tolerant control (AFTC), on the other hand, involves automatically detecting and identifying the failed components using a fault detection and identification (FDI) scheme and then reconfiguring the control law on-line in response to the FDI algorithm decisions. Despite the evident interaction between FDI and reconfiguration algorithms, the research on FDI and reconfiguration methods has often evolved separately. Typically, in the reconfiguration literature, it is usually assumed that a perfect FDI device is available (*i.e.* no detection delays, no false alarms, no missed detection ...*etc.*). On the other hand, an FDI algorithm is judged satisfactory if it detects failures quickly and generates few false alarms, regardless of the performance of the closed loop reconfigured system. The first purpose of this paper is then to use a mathematical model that includes in the same analysis framework

the FDI and reconfiguration algorithms. Such a model was initially proposed by [18,19]. The author shows that when the FDI output is feedback into a reconfiguration closed-loop control, a short detection delay and a low false alarm rate do not guarantee stable operation of the reconfigurable system. This makes the synthesis of reconfigurable controllers that take explicitly into account the inherent imperfections of the FDI process of great importance.

The proposed model in [18] belongs to the class of MJLS. In MJLS, a single jump process is used to describe the random variations affecting the system parameters. This process is represented by a finite state Markov chain and is called the plant regime mode. The theory of stability, optimal control and $\mathcal{H}_2/\mathcal{H}_\infty$ control, as well as important applications of such systems, can be found in several papers in the current literature, for instance in [3,4,2,6,8,9,13,14,19]. In the class of stochastic hybrid systems introduced in [18,19], two random processes are defined: the first random process represents system components failures and the second random process represents the FDI process used to reconfigure the control law. The problem of stochastic stability of active fault tolerant control systems with Markovian parameters (AFTC-SMP) in the presence of noise, parameter uncertainties, detection errors, detection delays and actuator saturation limits as well as $\mathcal{H}_\infty$ performance analysis have been investigated in [1,16,17,21,22].

On the other hand, one of the most challenging open problems in control theory is the synthesis of fixed-order or static output feedback controllers that meet desired

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performances and specifications [25]. Among all variations of this problem, this note is concerned with the problem of static output feedback control of continuous time AFTCSMP with state-dependent noise. The main contribution is to formulate conditions for multi-performance design related to this class of stochastic hybrid systems. The specifications and objectives under consideration include stochastic stability and $\mathcal{H}_2/\mathcal{H}_\infty$ performances. Another problematic related to a more general class of stochastic hybrid systems, (MJLS), is also addressed. This problematic concerns the mode-independent static output feedback control of this class of systems. Most of the developed results in the field of MJLS control are obtained under the restrictive assumption of control with complete state observation (Markov and system states). However, the access to the system mode may not be possible in some circumstances, which limits the use of such controllers. For instance, if the changes in the Markov chain are associated to failures of components of non-critical significance, or more generally if some changes are difficult to measure, it is quite possible that the associated Markov states are not accessible to controller. This scenario impose serious limitations in the analytical front [10]. One can distinguish two approaches for the design of mode-independent controllers. The first one uses a constant Lyapunov function [5,7,11], which leads to a linear matrix inequality (LMI) formulation of the design problem. The second one uses mode-dependent Lyapunov function and therefore leads to less conservative results. This approach usually leads to bilinear matrix inequality (BMI) problem. All the aforementioned works are based on the critical assumption of the complete access to the state vector. Practically, this assumption may sometimes be hard to satisfy and therefore, the developed results are restrictive. To the best of our knowledge, to date the problem of static output-feedback $\mathcal{H}_2/\mathcal{H}_\infty$ control of continuous-time MJLS with a *non-accessible* jumping parameter has not yet been fully addressed. Our approach belongs to the class of methods that use mode-dependent Lyapunov functions.

This paper is organized as follows: Section 2 describes the dynamical model of the system. A brief summary of basic stochastic terms and results are given in Section 3. Section 4 addresses the stochastic stabilization problematic. Section 5 considers the $\mathcal{H}_2/\mathcal{H}_\infty$ control problem. In Section 6, a numerical algorithm is provided and its running is illustrated on a numerical example. Section 7 concludes this paper.

Notations. $\mathbb{R}^{m \times n}$ is the set of m -by- n real matrices and $\mathbb{S}^n$ is the subset of symmetric matrices in $\mathbb{R}^{n \times n}$. A' is the transpose of the matrix A . The notation $X \geq Y$ ($X > Y$, respectively), where X and Y are symmetric matrices, means that $X - Y$ is positive semi-definite (positive definite, respectively); $\mathbb{I}$ and $\mathbf{0}$ are identity and zero matrices of appropriate dimensions, respectively; $\mathcal{E}\{\cdot\}$ denotes the expectation operator with respect to some probability measure P ; $L^2[0, \infty)$ stands for the space of square-integrable vector functions over the in-

terval $[0, \infty)$; $\|\cdot\|$ refers to either the Euclidean vector norm or the matrix norm, which is the operator norm induced by the standard vector norm; $\|\cdot\|_2$ stands for the norm in $L^2[0, \infty)$; while $\|\cdot\|_{\mathcal{E}_2}$ denotes the norm in $L^2((\Omega, \mathcal{F}, P), [0, \infty))$; $(\Omega, \mathcal{F}, P)$ is a probability space. We also denote $\bullet(t) = \bullet_t$, and the initial conditions $\bullet(t_0) = \bullet_0$. In block matrices, $\star$ indicates symmetric

$$\text{terms: } \begin{bmatrix} A & B \\ B' & C \end{bmatrix} = \begin{bmatrix} A & \star \\ B' & C \end{bmatrix} = \begin{bmatrix} A & B \\ \star & C \end{bmatrix}.$$

2 Dynamical model

To describe the class of linear systems with Markovian jumping parameters that we deal with in this paper, let us fix a complete probability space $(\Omega, \mathcal{F}, P)$. The dynamical model of the reconfigurable system with Markovian process and state dependent noise, defined in the fundamental probability space $(\Omega, \mathcal{F}, P)$, is described by the following differential equations:

$$\varphi : \begin{cases} dx_t = A(\xi_t)x_t dt + B(\eta_t)u_t dt + E(\xi_t, \eta_t)w_t dt \\ \quad + \sum_{l=1}^v \mathbb{W}_l(\xi_t, \eta_t)x_t d\varpi_{lt} \\ y_t = C_2 x_t + D_2(\xi_t, \eta_t)w_t \\ z_t = C_1 x_t + D_1(\eta_t)u_t(y_t, \psi_t, t) \end{cases} \quad (1)$$

where $x_t \in \mathbb{R}^n$ is the system state, $u_t \in \mathbb{R}^r$ is the system input, $y_t \in \mathbb{R}^q$ is the system measured output, $z_t \in \mathbb{R}^p$ is the controlled output, $w_t \in \mathbb{R}^m$ is the system external disturbance, ξ_t , η_t and ψ_t represent the plant component failure process, the actuator failure process and the FDI process, respectively. ξ_t , η_t and ψ_t are separable and measurable Markov processes with finite state spaces $Z = \{1, 2, \dots, z\}$, $S = \{1, 2, \dots, s\}$ and $R = \{1, 2, \dots, r\}$, and transition rate matrices $\Pi = [\pi_{ij}, i \in Z, j \in Z]$, $\Upsilon = [\nu_{ij}, i \in S, j \in S]$, $\Lambda = [\lambda_{iv}^{kl}, i \in R, v \in R, k \in Z, l \in S]$ respectively. π_{ij} is the plant components failure rate, and ν_{kl} is the actuator failure rate. Here, λ_{iv}^{kl} represents the transition rate from i to v for the Markov process $\psi(t)$ conditioned on $\xi = k \in Z$ and $\eta = l \in S$. Depending on the values of $i, v \in R, k \in Z$ and $l \in S$, various interpretations, such as *rate of false detection and isolation*, *rate of correct detection and isolation*, *false alarm recovery rate*, etc, can be given to λ_{iv}^{kl} [17,24]. $\varpi_t = [\varpi_{1t} \dots \varpi_{vt}]'$ is a v -dimensional standard Wiener process on a given probability space $(\Omega, \mathcal{F}, P)$, that is assumed to be independent of the Markov processes.

In AFTC, we consider that the control law is only a function of the measurable FDI process $\psi(t)$. Therefore, we introduce a static output feedback compensator (φ_s) of the form:

$$\varphi_s : \begin{cases} u_t = \mathcal{K}(\psi_t)y_t \end{cases} \quad (2)$$

Applying the controller φ_s to φ , we obtain the following closed loop system:

$$\varphi_{cl} : \begin{cases} dx_t = \bar{A}(\xi_t, \eta_t, \psi_t)x_t dt + \bar{E}(\xi_t, \eta_t, \psi_t)w_t dt \\ \quad + \sum_{l=1}^v \mathbb{W}_l(\xi_t, \eta_t)x_t d\varpi_{lt} \\ y_t = C_2 x_t + D_2(\xi_t, \eta_t)w_t \\ z_t = \bar{C}_1(\eta_t, \psi_t)x_t + \bar{D}_1(\xi_t, \eta_t, \psi_t)w_t \end{cases} \quad (3)$$

where

$$\begin{bmatrix} \bar{A}(\xi_t, \eta_t, \psi_t) & \bar{E}(\xi_t, \eta_t, \psi_t) \\ \bar{C}_1(\eta_t, \psi_t) & \bar{D}_1(\xi_t, \eta_t, \psi_t) \end{bmatrix} = \begin{bmatrix} A(\xi_t) & E(\xi_t, \eta_t) \\ C_1 & 0 \end{bmatrix} + \begin{bmatrix} B(\eta_t) \\ D_1(\eta_t) \end{bmatrix} \mathcal{K}(\psi_t) \begin{bmatrix} C_2 & D_2(\xi_t, \eta_t) \end{bmatrix}$$

For notational simplicity, we will denote $\bullet(\xi_t) = \bullet_i$ when $\xi_t = i \in Z$, $\bullet(\eta_t) = \bullet_j$, $\bullet(\xi_t, \eta_t) = \bullet_{ij}$ when $\xi_t = i \in Z, \eta_t = j \in S$ and $\bullet(\psi_t) = \bullet_k$ when $\psi_t = k \in R$.

3 Definitions and Basic Results

In this section, we will first give basic definitions related to stochastic stability notions and then we will summarize some results about stochastic stability of reconfigurable systems with Markovian processes.

3.1 Stochastic Stability

For system (1), when $u_t \equiv 0$ for all $t \geq 0$, we have the following definition.

Definition 1. System (1) is said to be internally exponentially stable in the mean square sense (IESS), if there exist positive constants α and β such that the solution of

$$dx_t = A(\xi_t)x_t dt + \sum_{l=1}^v \mathbb{W}_l(\xi_t, \eta_t)x_t d\varpi_{lt}$$

satisfies the following inequality

$$\mathcal{E} \{ \|x_t\|^2 \} \leq \beta \|x_0\|^2 \exp[-\alpha(t - t_0)] \quad (4)$$

for arbitrary initial conditions (x_0, ξ_0, η_0) .

Definition 2. System (1) is said to be internally exponentially stabilizable in the mean square sense by static output feedback, if there exists a control φ_s such that the closed loop system (3) is IEISS.

The following proposition gives a necessary and sufficient condition for IEISS of the system (3).

Proposition 1. System (3) is IEISS if and only if there exists symmetric positive-definite matrices $\mathcal{P}_{ijk}$, $i \in Z$, $j \in S$ and $k \in R$ such that:

$$\begin{aligned} \bar{A}'_{ijk}\mathcal{P}_{ijk} + \mathcal{P}_{ijk}\bar{A}_{ijk} + \sum_{l=1}^v \mathbb{W}'_{lij}\mathcal{P}_{ijk}\mathbb{W}_{lij} + \sum_{h \in Z} \pi_{ih}\mathcal{P}_{hjk} \\ + \sum_{l \in S} \nu_{jl}\mathcal{P}_{ilk} + \sum_{v \in R} \lambda_{kv}^{ij}\mathcal{P}_{ijv} = \mathfrak{I}_{ijk} < 0 \end{aligned} \quad (5)$$

where: $\bar{A}_{ijk} = A_i + B_j\mathcal{K}_kC_2$

Proof. Due to page limit, the proof of this proposition is not detailed here. It is based on classical tools from stochastic stability analysis [15]. ■

Proposition 2. If system (3) is IEISS, for every $w = \{w_t; t \geq 0\} \in L_2[0, \infty)$, we have that $x = \{x_t; t \geq 0\} \in L_2((\Omega, \mathcal{F}, P), [0, \infty))$, i.e., $\mathcal{E} \{ \int_0^\infty x_t^T x_t dt \} < \infty$, for any initial conditions.

The proof of proposition 2 is omitted. It follows the same arguments as for the proof of proposition 5.1 in [1].

3.2 Matrix Ellipsoids

Through this paper, a particular set of matrices is used. In the literature, these sets are referred to as matrix ellipsoids of $\mathbb{R}^{(m \times p)}$.

Definition 3. [20] Given three matrices $\mathbb{X} \in \mathbb{S}^q$, $\mathbb{Y} \in \mathbb{R}^{q \times r}$ and $\mathbb{Z} \in \mathbb{S}^r$, the $\{\mathbb{X}, \mathbb{Y}, \mathbb{Z}\}$ -ellipsoid of $\mathbb{R}^{r \times q}$ is the set of matrices $\mathcal{K}$ satisfying the following matrix inequalities:

$$\mathbb{Z} > 0; \quad \begin{bmatrix} \mathbb{I} & \mathcal{K}' \end{bmatrix} \begin{bmatrix} \mathbb{X} & \mathbb{Y} \\ \star & \mathbb{Z} \end{bmatrix} \begin{bmatrix} \mathbb{I} \\ \mathcal{K} \end{bmatrix} \leq 0 \quad (6)$$

By definition, $\mathcal{K}_0 = -\mathbb{Z}^{-1}\mathbb{Y}'$ is the center of the ellipsoid and $R = \mathcal{K}_0'\mathbb{Z}\mathcal{K}_0 - \mathbb{X}$ is the radius. Inequality (6) can also be written as

$$\mathbb{Z} > 0; \quad (\mathcal{K} - \mathcal{K}_0)'\mathbb{Z}(\mathcal{K} - \mathcal{K}_0) \leq R \quad (7)$$

This definition shows that matrix ellipsoids are special cases of matrix sets defined by quadratic matrix inequality. Some properties of these sets are:

- i) A matrix ellipsoid is a convex set;
- ii) $\{\mathbb{X}, \mathbb{Y}, \mathbb{Z}\}$ -ellipsoid is nonempty iff the radius ($R \geq 0$) is positive semi definite. This property can also be expressed as

$$\mathbb{X} \leq \mathbb{Y}\mathbb{Z}^{-1}\mathbb{Y}' \quad (8)$$

4 Stochastic Stabilization

In this section, we shall address the problem of finding all static compensators (φ_s), as defined in section 2, such that the closed loop system (φ_{cl}) becomes IEISS. To this end, we use proposition 1 and the idea of synthesizing ellipsoidal sets of controllers [20] to get the following necessary and sufficient conditions for the IEISS of the system (3).

Proposition 3. System (1) is internally exponentially stabilisable in the mean square sense by static output-feedback if and only if there exists matrices $\mathcal{P}_{ijk} = \mathcal{P}'_{ijk} > 0$, $\mathbb{X}_k \in \mathbb{S}^q$, $\mathbb{Y}_k \in \mathbb{R}^{q \times r}$ and $\mathbb{Z}_k = \mathbb{Z}'_k > 0$ that simultaneously satisfy the following LMI constraints

$$\begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_j \end{bmatrix}' \begin{bmatrix} \Theta_{ijk} & \mathcal{P}_{ijk} \\ \mathcal{P}_{ijk} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_j \end{bmatrix} < \begin{bmatrix} C_2 & \mathbf{0} \\ \mathbf{0} & \mathbb{I} \end{bmatrix}' \begin{bmatrix} \mathbb{X}_k & \mathbb{Y}_k \\ \star & \mathbb{Z}_k \end{bmatrix} \begin{bmatrix} C_2 & \mathbf{0} \\ \mathbf{0} & \mathbb{I} \end{bmatrix} \quad (9)$$

and the nonlinear inequalities constraints

$$\mathbb{X}_k \leq \mathbb{Y}_k\mathbb{Z}_k^{-1}\mathbb{Y}_k' \quad (10)$$

$\forall i \in Z, j \in S$ and $k \in R$, where

$$\begin{aligned} \Theta_{ijk} = \sum_{l=1}^v \mathbb{W}'_{lij}\mathcal{P}_{ijk}\mathbb{W}_{lij} + \sum_{h \in Z} \pi_{ih}\mathcal{P}_{hjk} + \sum_{l \in S} \nu_{jl}\mathcal{P}_{ilk} \\ + \sum_{v \in R} \lambda_{kv}^{ij}\mathcal{P}_{ijv} \end{aligned}$$

Let $\{\mathcal{P}_{ijk}, \mathbb{X}_k, \mathbb{Y}_k, \mathbb{Z}_k\}$ be a solution, then the nonempty $\{\mathbb{X}_k, \mathbb{Y}_k, \mathbb{Z}_k\}$ -ellipsoids are sets of stabilizing gains.

Proof.

Sufficiency. Assume that (9)-(10) are satisfied for some $\{\mathcal{P}_{ijk}, \mathbb{X}_k, \mathbb{Y}_k, \mathbb{Z}_k\}$ matrices. Due to the properties of matrix ellipsoids, the $\{\mathbb{X}_k, \mathbb{Y}_k, \mathbb{Z}_k\}$ -ellipsoids are nonempty. Take any element $\mathcal{K}_k$. The LMI (9) implies that for all

$$\begin{pmatrix} x'_t & u'_t \end{pmatrix} \neq 0$$

$$\begin{pmatrix} x_t \\ A_i x_t + B_j u_t \end{pmatrix}' \begin{bmatrix} \Theta_{ijk} & \mathcal{P}_{ijk} \\ \mathcal{P}_{ijk} & \mathbf{0} \end{bmatrix} \begin{pmatrix} x_t \\ A_i x_t + B_j u_t \end{pmatrix} < \begin{pmatrix} C_2 x_t \\ u_t \end{pmatrix}' \begin{bmatrix} \mathbb{X}_k & \mathbb{Y}_k \\ \star & \mathbb{Z}_k \end{bmatrix} \begin{pmatrix} C_2 x_t \\ u_t \end{pmatrix} \quad (11)$$

Definition 3 implies that for all nonzero trajectories

$$x'_t \mathbb{I}_{ijk} x_t < y'_t \begin{bmatrix} \mathbb{I} & \mathcal{K}'_k \end{bmatrix} \begin{bmatrix} \mathbb{X}_k & \mathbb{Y}_k \\ \star & \mathbb{Z}_k \end{bmatrix} \begin{bmatrix} \mathbb{I} \\ \mathcal{K}_k \end{bmatrix} y_t \leq 0 \quad (12)$$

Then, the closed-loop exponential stochastic stability follows from proposition 1 for the quadratic stochastic Lyapunov function $\vartheta(\xi_t, \eta_t, \psi_t) = x'_t \mathcal{P}(\xi_t, \eta_t, \psi_t) x_t$.

Necessity. Assume $\mathcal{K}_k$ are stabilizing static output feedback gains and $\vartheta(\xi_t, \eta_t, \psi_t) = x'_t \mathcal{P}(\xi_t, \eta_t, \psi_t) x_t$ is a stochastic Lyapunov function. Then from proposition 1, we have

$$\begin{aligned} & \begin{bmatrix} \mathcal{K}_k C_2 & -\mathbb{I} \end{bmatrix} \begin{pmatrix} x_t \\ u_t \end{pmatrix} = 0 \Rightarrow \\ & \begin{pmatrix} x_t \\ u_t \end{pmatrix}' \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_j \end{bmatrix}' \begin{bmatrix} \Theta_{ijk} & \mathcal{P}_{ijk} \\ \mathcal{P}_{ijk} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_j \end{bmatrix} \begin{pmatrix} x_t \\ u_t \end{pmatrix} < 0 \end{aligned} \quad (13)$$

Based on the Finsler Lemma [23], there exist scalars τ_{ijk} such that

$$\begin{aligned} & \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_j \end{bmatrix}' \begin{bmatrix} \Theta_{ijk} & \mathcal{P}_{ijk} \\ \mathcal{P}_{ijk} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_j \end{bmatrix} < \\ & \tau_{ijk} \begin{bmatrix} \mathcal{K}_k C_2 & -\mathbb{I} \end{bmatrix}' \begin{bmatrix} \mathcal{K}_k C_2 & -\mathbb{I} \end{bmatrix} \leq \varepsilon_k \begin{bmatrix} \mathcal{K}_k C_2 & -\mathbb{I} \end{bmatrix}' \begin{bmatrix} \mathcal{K}_k C_2 & -\mathbb{I} \end{bmatrix} \end{aligned} \quad (14)$$

where $\varepsilon_k = \max_{i,j}(\tau_{ijk})$. The inequality (9) is obtained

with $\mathbb{X}_k = \varepsilon_k \mathcal{K}'_k \mathcal{K}_k$, $\mathbb{Y}_k = -\varepsilon_k \mathcal{K}'_k$, $\mathbb{Z}_k = \varepsilon_k \mathbb{I}$. (14) implies $\mathbf{0} < \mathbb{Z}_k$. Hence the proof is complete. ■

Mode-Independent Stabilization of MJLS: Consider the following continuous-time MJLS:

$$\begin{cases} dx_t = A(\phi_t)x_t dt + B(\phi_t)u_t dt + E(\phi_t)w_t dt \\ \quad + \sum_{l=1}^v \mathbb{W}_l(\phi_t)x_t d\varpi_{lt} \\ y_t = C_2(\phi_t)x_t + D_2(\phi_t)u_t \\ z_t = C_1(\phi_t)x_t + D_1(\phi_t)u_t \end{cases} \quad (15)$$

$\{\phi_t, t \geq 0\}$ is a continuous-time homogeneous Markov process with finite state-space $\Xi = \{1, \dots, \sigma\}$ and transition rate matrix $\Lambda = [\lambda_{ij}, i \in \Xi, j \in \Xi]$. The set Ξ comprises the operation modes of system (15) and for each possible value of $\theta_t = i$, $i \in \Xi$, we denote the matrices associated with the " i -th mode" by: $\mathcal{M}_i = \mathcal{M}(\theta_t = i)$. Most of the developed results in the field of MJLS control are obtained under the restrictive assumption of control with complete state observation (Markov and system states). However, the access to the system mode and system state vector may not be possible in some circumstances, which limits the use of such controllers. The design of mode-independent output feedback controller is of great interest in this case. The mode-independent controller used in the rest of the paper is assumed to be of the form: $u_t = \mathcal{K}y_t$.

We will design the mode-independent static output feedback controller using mode-dependent Lyapunov function approach. This leads to less conservative results when using mode-independent Lyapunov functions.

Proposition 4. System (15) is internally exponentially stabilisable in the mean square sense by static output-feedback if and only if there exists matrices $\mathcal{P}_i = \mathcal{P}'_i > 0$, $\mathbb{X} \in \mathbb{S}^q$, $\mathbb{Y} \in \mathbb{R}^{q \times r}$ and $\mathbb{Z} = \mathbb{Z}' > 0$ that simultaneously satisfy the following LMI constraints

$$\begin{aligned} & \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_i \end{bmatrix}' \begin{bmatrix} \Theta_i & \mathcal{P}_i \\ \mathcal{P}_i & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_i \end{bmatrix} < \\ & \begin{bmatrix} C_{2i} & \mathbf{0} \\ \mathbf{0} & \mathbb{I} \end{bmatrix}' \begin{bmatrix} \mathbb{X} & \mathbb{Y} \\ \star & \mathbb{Z} \end{bmatrix} \begin{bmatrix} C_{2i} & \mathbf{0} \\ \mathbf{0} & \mathbb{I} \end{bmatrix} \end{aligned} \quad (16)$$

and the nonlinear inequality constraint

$$\mathbb{X} \leq \mathbb{Y} \mathbb{Z}^{-1} \mathbb{Y}' \quad (17)$$

$\forall i \in \Xi$, where $\Theta_i = \sum_{l=1}^v \mathbb{W}'_{li} \mathcal{P}_i \mathbb{W}_{li} + \sum_{v \in \Xi} \lambda_{iv} \mathcal{P}_v$. Let $\{\mathcal{P}_i, \mathbb{X}, \mathbb{Y}, \mathbb{Z}\}$ be a solution, then the nonempty $\{\mathbb{X}, \mathbb{Y}, \mathbb{Z}\}$ -ellipsoid is a set of stabilizing gains.

5 The Control Problem

5.1 $\mathcal{H}_\infty$ Control

Let us consider the system (1) with

$$z_t = z_{\infty t} = C_{\infty 1} x_t + D_{\infty 1}(\eta_t) u(y_t, \psi_t, t)$$

$z_{\infty t}$ stands for the controlled output related to $\mathcal{H}_\infty$ performance.

In this section, we deal with the design of controllers that stochastically stabilize the closed-loop system and guarantee the disturbance rejection, with a certain level $\gamma_\infty > 0$. Mathematically, we are concerned with the characterization of compensators φ_s that stochastically stabilize the system (3) and guarantee the following for all $w \in L^2[0, \infty)$:

$$\|z_\infty\|_{\mathcal{E}_2} = \mathcal{E} \left\{ \int_0^\infty z'_{\infty t} z_{\infty t} dt \right\}^{1/2} < \gamma_\infty \|w\|_2 \quad (18)$$

where $\gamma_\infty > 0$ is a prescribed level of disturbance attenuation to be achieved. To this end, we need the auxiliary result given by the following proposition.

Proposition 5. If there exists symmetric positive-definite matrices $\mathcal{P}_{\infty ijk}$, $i \in Z$, $j \in S$ and $k \in R$ such that

$$\begin{bmatrix} \mathfrak{J}_{\infty ijk} & \bar{C}'_{1jk} \bar{D}_{1ijk} + \mathcal{P}_{\infty ijk} \bar{E}_{ijk} \\ \star & \bar{D}'_{1ijk} \bar{D}_{1ijk} - \gamma_\infty^2 \mathbb{I} \end{bmatrix} < 0 \quad (19)$$

where $\mathfrak{J}_{\infty ijk}$ is obtained from $\mathfrak{J}_{ijk}$ by replacing $\mathcal{P}_{ijk}$ by $\mathcal{P}_{\infty ijk}$, then the system (3) is stochastically stable and satisfies (18).

Proof. It follows the same arguments as for the proof of proposition 5.2 in [1]. ■

Using the previous proposition, the following $\mathcal{H}_\infty$ control result can be stated.

Proposition 6. If there exists matrices $\mathcal{P}_{\infty ijk} = \mathcal{P}'_{\infty ijk} > 0$, $\mathbb{X}_k \in \mathbb{S}^q$, $\mathbb{Y}_k \in \mathbb{R}^{q \times r}$ and $\mathbb{Z}_k = \mathbb{Z}'_k > 0$ that simultaneously satisfy the following LMI constraints

$$\begin{aligned} \mathbb{M}'_{1ij} \begin{bmatrix} \Theta_{ijk} & \mathcal{P}_{\infty ijk} \\ \mathcal{P}_{\infty ijk} & \mathbf{0} \end{bmatrix} \mathbb{M}_{1ij} &< \mathbb{M}'_{2j} \begin{bmatrix} -\mathbb{I} & \mathbf{0} \\ \mathbf{0} & \gamma_\infty^2 \mathbb{I} \end{bmatrix} \mathbb{M}_{2j} \\ &+ \mathbb{M}'_{3ij} \begin{bmatrix} \mathbb{X}_k & \mathbb{Y}_k \\ \star & \mathbb{Z}_k \end{bmatrix} \mathbb{M}_{3ij} \end{aligned} \quad (20)$$

and the nonlinear constraints (10), where $\mathbb{M}_{1ij} =$

$$\begin{bmatrix} \mathbb{I} & \mathbf{0} & \mathbf{0} \\ A_i & E_{ij} & B_j \end{bmatrix}, \quad \mathbb{M}_{2j} = \begin{bmatrix} C_{\infty 1} & \mathbf{0} & D_{\infty 1j} \\ \mathbf{0} & \mathbb{I} & \mathbf{0} \end{bmatrix}, \quad \mathbb{M}_{3ij} = \begin{bmatrix} C_2 & D_{2ij} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbb{I} \end{bmatrix},$$

then the $\{\mathbb{X}_k, \mathbb{Y}_k, \mathbb{Z}_k\}$ -ellipsoids are sets

of stabilizing gains such that (18) is verified.

Proof. The proof of this proposition is based on proposition 5 and uses the same arguments as for the proof of proposition 3. ■

The mode-independent static output feedback $\mathcal{H}_\infty$ control result is summarized by the following proposition.

Proposition 7. If there exists matrices $\mathcal{P}_{\infty i} = \mathcal{P}'_{\infty i} > 0$, $\mathbb{X} \in \mathbb{S}^q$, $\mathbb{Y} \in \mathbb{R}^{q \times r}$ and $\mathbb{Z} = \mathbb{Z}' > 0$ that simultaneously satisfy the following LMI constraints

$$\begin{aligned} \mathbb{M}'_{1i} \begin{bmatrix} \Theta_i & \mathcal{P}_{\infty i} \\ \mathcal{P}_{\infty i} & \mathbf{0} \end{bmatrix} \mathbb{M}_{1i} &< \mathbb{M}'_{2i} \begin{bmatrix} -\mathbb{I} & \mathbf{0} \\ \mathbf{0} & \gamma_\infty^2 \mathbb{I} \end{bmatrix} \mathbb{M}_{2i} \\ &+ \mathbb{M}'_{3i} \begin{bmatrix} \mathbb{X} & \mathbb{Y} \\ \star & \mathbb{Z} \end{bmatrix} \mathbb{M}_{3i} \end{aligned} \quad (21)$$

and the nonlinear inequality constraint (17), where

$$\mathbb{M}_{1i} = \begin{bmatrix} \mathbb{I} & \mathbf{0} & \mathbf{0} \\ A_i & E_i & B_i \end{bmatrix}, \quad \mathbb{M}_{2i} = \begin{bmatrix} C_{\infty 1i} & \mathbf{0} & D_{\infty 1i} \\ \mathbf{0} & \mathbb{I} & \mathbf{0} \end{bmatrix}, \quad \mathbb{M}_{3i} = \begin{bmatrix} C_{2i} & D_{2i} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbb{I} \end{bmatrix},$$

then the $\{\mathbb{X}, \mathbb{Y}, \mathbb{Z}\}$ -ellipsoid is a set of stabilizing gains such that (18) is satisfied.

5.2 $\mathcal{H}_2$ Control

Consider the system (1) with

$$z_t = z_{2t} = C_{21}x_t + D_{21}(\eta_t)u(y_t, \psi_t, t)$$

and $D_2(\xi_t, \eta_t) = \mathbf{0}$. z_{2t} stands for the controlled output related to $\mathcal{H}_2$ performance.

Before introducing the $\mathcal{H}_2$ control result, let us consider the following definition.

Definition 4. We define the $\mathcal{H}_2$ -norm of the IESS system (φ_{cl}) as

$$\|\varphi_{cl}\|_2^2 = \sum_{d=1}^m \sum_{i,j,k} \mu_{ijk} \|z_{d,i,j,k}\|_{\mathcal{E}_2}^2$$

where $z_{d,i,j,k}$ is the controlled output of the system with initial conditions $\xi_0 = i$, $\eta_0 = j$, $\psi_0 = k$, disturbed by $w_t = e_d \delta_t$. e_d is a m -dimensional unitary vector with its d th entry as 1 and other entries as 0's. δ_t is an impulse function and μ_{ijk} is the initial probability distribution for $\xi_0 = i$, $\eta_0 = j$, $\psi_0 = k$.

From the definition above and using the same arguments as in [6,26,27], we can state the following proposition.

Proposition 8. Assume that (φ_{cl}) is IESS then

- i) $\|\varphi_{cl}\|_2^2 = \sum_{i,j,k} \mu_{ijk} \text{tr}(E'_{ij} \mathcal{P}_{oijk} E_{ij})$, where $\mathcal{P}_o = \{\mathcal{P}_{o111}, \dots, \mathcal{P}_{oszzr}\}$ denotes the observability Gramian, i.e., $\mathcal{P}_{oijk}$ are the unique positive semi definite solutions of the following equations:

$$\mathfrak{J}_{oijk} + \bar{C}'_{21} \bar{C}_{21} = 0 \quad (22)$$

- ii) $\|\varphi_{cl}\|_2^2 < \sum_{i,j,k} \mu_{ijk} \text{tr}(E'_{ij} \mathcal{P}_{2ijk} E_{ij})$, where $\mathcal{P}_{2ijk}$ is a positive definite solution of the following matrix inequality

$$\mathfrak{J}_{2ijk} + \bar{C}'_{21} \bar{C}_{21} < 0 \quad (23)$$

- iii) If there exists positive definite matrices $\mathcal{P}_{2ijk}$ and matrices $\mathcal{K}_k$ such that

$$\begin{aligned} \sum_{i,j,k} \mu_{ijk} \text{tr}(E'_{ij} \mathcal{P}_{2ijk} E_{ij}) &< \gamma_2^2 \\ \mathfrak{J}_{2ijk} + \bar{C}'_{21} \bar{C}_{21} &< 0 \end{aligned} \quad (24)$$

Then $\mathcal{K}_k$ are stabilizing gains such that $\|\varphi_{cl}\|_2 < \gamma_2$.

Using the same framework as for the stochastic stabilization and the $\mathcal{H}_\infty$ control problematic, similar $\mathcal{H}_2$ control results are obtained and summarized as follows:

Proposition 9. If there exists matrices $\mathcal{P}_{2ijk} = \mathcal{P}'_{2ijk} > 0$, $\mathbb{X}_k \in \mathbb{S}^q$, $\mathbb{Y}_k \in \mathbb{R}^{q \times r}$ and $\mathbb{Z}_k = \mathbb{Z}'_k > 0$ that simultaneously satisfy the constraints (10), (24) and

$$\mathbb{N}'_{1ij} \begin{bmatrix} \Theta_{ijk} & \mathcal{P}_{2ijk} \\ \mathcal{P}_{2ijk} & \mathbf{0} \end{bmatrix} \mathbb{N}_{1ij} < -\mathbb{N}'_{2j} \mathbb{N}_{2j} + \mathbb{N}'_3 \begin{bmatrix} \mathbb{X}_k & \mathbb{Y}_k \\ \star & \mathbb{Z}_k \end{bmatrix} \mathbb{N}_3 \quad (25)$$

$$\mathbb{N}_{1ij} = \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_j \end{bmatrix}, \quad \mathbb{N}_{2j} = \begin{bmatrix} C_{21} & D_{21j} \end{bmatrix}, \quad \mathbb{N}_3 = \begin{bmatrix} C_2 & \mathbf{0} \\ \mathbf{0} & \mathbb{I} \end{bmatrix},$$

then the $\{\mathbb{X}_k, \mathbb{Y}_k, \mathbb{Z}_k\}$ -ellipsoids are sets of

stabilizing gains such that $\|\varphi_{cl}\|_2 < \gamma_2$.

We conclude this section by giving, as in the previous sections, results relating to mode-independent $\mathcal{H}_2$ control of MJLS.

Proposition 10. If there exist matrices $\mathcal{P}_{2i} = \mathcal{P}'_{2i} > 0$, $\mathbb{X} \in \mathbb{S}^q$, $\mathbb{Y} \in \mathbb{R}^{q \times r}$ and $\mathbb{Z} = \mathbb{Z}' > 0$ that simultaneously satisfy the constraints (17), $\sum_i \mu_i \text{tr}(E'_i \mathcal{P}_{2i} E_i) < \gamma_2^2$ and

$$\mathbb{N}'_{1i} \begin{bmatrix} \Theta_i & \mathcal{P}_{2i} \\ \mathcal{P}_{2i} & \mathbf{0} \end{bmatrix} \mathbb{N}_{1i} < -\mathbb{N}'_{2i} \mathbb{N}_{2i} + \mathbb{N}'_{3i} \begin{bmatrix} \mathbb{X} & \mathbb{Y} \\ \star & \mathbb{Z} \end{bmatrix} \mathbb{N}_{3i} \quad (26)$$

$\mathbb{N}_{1i} = \begin{bmatrix} \mathbb{I} & \mathbf{0} \\ A_i & B_i \end{bmatrix}$, $\mathbb{N}_{2i} = \begin{bmatrix} C_{21i} & D_{21i} \end{bmatrix}$, $\mathbb{N}_{3i} = \begin{bmatrix} C_{2i} & \mathbf{0} \\ \mathbf{0} & \mathbb{I} \end{bmatrix}$, then the $\{\mathbb{X}, \mathbb{Y}, \mathbb{Z}\}$ -ellipsoid is a set of stabilizing gains such that $\|\varphi_{cl}\|_2 < \gamma_2$.

5.3 $\mathcal{H}_2/\mathcal{H}_\infty$ Synthesis

The $\mathcal{H}_2/\mathcal{H}_\infty$ synthesis problem consists in finding a common controller that stochastically stabilizes the system and ensures the different performance levels. This can be stated as follows:

For two given levels on the $\mathcal{H}_\infty$ and $\mathcal{H}_2$ norms, γ_∞ and γ_2 respectively, find stabilizing static output feedback gains $\mathcal{K}_k$ such that

$$\begin{aligned} \|z_\infty\|_{\mathcal{E}_2} &< \gamma_\infty \|\mathbf{w}\|_2 \\ \|\varphi_{cl}\|_2 &< \gamma_2 \end{aligned}$$

Corollary 1. If there exists matrices $\mathcal{P}_{\infty ijk}$, $\mathcal{P}_{2ijk}$, $\mathbb{X}_k \in \mathbb{S}^q$, $\mathbb{Y}_k \in \mathbb{R}^{q \times r}$ and $\mathbb{Z}_k \in \mathbb{S}^r$ that simultaneously satisfy the constraints (10), (20) and (24)-(25), then the $\{\mathbb{X}_k, \mathbb{Y}_k, \mathbb{Z}_k\}$ -ellipsoids are sets of stabilizing gains such that the performances levels are satisfied.

6 Computational Issues and Examples

6.1 A Cone Complementary Algorithm (CCL)

In this section, an iterative algorithm is proposed to solve the different synthesis conditions. It is based on a cone complementary technique [12]. The proposed CCL algorithm is based on the following lemma which is obtained using the same arguments as in [20].

Lemma 1. The $\mathcal{H}_2/\mathcal{H}_\infty$ control problem is feasible if and only if zero is the global optimum of the optimisation problem

$$\begin{cases} \min & \text{tr}(\mathbb{T}\mathbb{S}) \\ \text{s.t.} & (20), (24), (25) \\ & \mathbb{X}_k \leq \hat{\mathbb{X}}_k \quad \mathcal{S}_k = \begin{bmatrix} \hat{\mathbb{X}}_k & \mathbb{Y}_k \\ \star & \mathbb{Z}_k \end{bmatrix} \geq 0 \\ & \mathcal{T}_{1k} \geq \mathbb{I} \quad \mathcal{T}_k = \begin{bmatrix} \mathcal{T}_{1k} & \mathcal{T}_{2k} \\ \star & \mathcal{T}_{3k} \end{bmatrix} \geq 0 \end{cases} \quad (27)$$

where $\mathbb{S} = \text{diag}\{\mathcal{S}_1, \dots, \mathcal{S}_r\}$, $\mathbb{T} = \text{diag}\{\mathcal{T}_1, \dots, \mathcal{T}_r\}$.

CCL algorithm. For two given levels γ_∞ , γ_2

- i) Find a feasible solution $\mathbb{X}_{k0}, \mathbb{Y}_{k0}, \mathbb{Z}_{k0}, \hat{\mathbb{X}}_{k0}, \mathcal{P}_{2ijk0}, \mathcal{P}_{\infty ijk0}, \mathbb{T}_0, \mathbb{S}_0$. If there is no solution, STOP, the algorithm failed. $h = 0$;
- ii) set $\mathbf{V}_h = \mathbb{S}_h$, $\mathbf{W}_h = \mathbb{T}_h$, and find $\mathbb{X}_{k(h+1)}, \mathbb{Y}_{k(h+1)}, \mathbb{Z}_{k(h+1)}, \hat{\mathbb{X}}_{k(h+1)}, \mathcal{P}_{2ijk(h+1)}, \mathcal{P}_{\infty ijk(h+1)}, \mathbb{T}_{h+1}, \mathbb{S}_{h+1}$, solutions of the LMI problem

$$\begin{cases} \min & \text{tr}(\mathbf{V}_h \mathbb{T} + \mathbf{W}_h \mathbb{S}) \\ \text{s.t.} & (20), (24), (25) \\ & \mathbb{X}_k \leq \hat{\mathbb{X}}_k \quad \mathcal{S}_k = \begin{bmatrix} \hat{\mathbb{X}}_k & \mathbb{Y}_k \\ \star & \mathbb{Z}_k \end{bmatrix} \geq 0 \\ & \mathcal{T}_{1k} \geq \mathbb{I} \quad \mathcal{T}_k = \begin{bmatrix} \mathcal{T}_{1k} & \mathcal{T}_{2k} \\ \star & \mathcal{T}_{3k} \end{bmatrix} \geq 0 \end{cases} \quad (28)$$

- iii) if $\text{tr}(\mathcal{T}_{h-1} \mathcal{S}_{h-1} - \mathcal{T}_h \mathcal{S}_h) < \varepsilon$, then STOP, the algorithm failed (slow progress).
- iv) if $\mathbb{X}_k \leq \mathbb{Y}_k \mathbb{Z}_k^{-1} \mathbb{Y}_k'$, STOP, matrix ellipsoids are found. Otherwise, set $h = h + 1$ and go to step ii).

6.2 Numerical Example

We applied the proposed mode-independent output feedback $\mathcal{H}_2/\mathcal{H}_\infty$ control to a VTOL helicopter model adapted from [8]. The dynamics can be written as:

$$\begin{cases} dx_t = A(\phi_t)x_t dt + B(\phi_t)u_t dt + Ew_t dt + \mathbb{W}_1 x_t d\varpi_t \\ y_t = C_2 x_t \\ z_{\infty t} = C_{\infty 1} x_t + D_{\infty 1} u_t \\ z_{2t} = C_{21} x_t + D_{21} u_t \end{cases}$$

where ϕ_t indicates the airspeed. The different parameters are given in [8]. The behavior of ϕ_t is modelled as a Markov chain with three different states, corresponding to airspeeds of 135 (nominal value), 60, 170 Knots. The transition matrix is given by:

$$\Xi = \begin{bmatrix} -0.0907 & 0.0671 & 0.0236 \\ 0.0671 & -0.0671 & 0 \\ 0.0236 & 0 & -0.0236 \end{bmatrix}$$

The central controller corresponding to $\gamma_\infty = \gamma_2 = 1$ (center of the ellipsoid) is given by

$$\mathcal{K}'_0 = \begin{bmatrix} 0.5143 & -0.4358 \end{bmatrix}$$

The state trajectories of the closed loop system resulting from the obtained controller are shown in Figure 1. Figure 2 represents the evolution of the controlled outputs $z_{\infty t}$. It can be seen that the closed-loop system is stochastically stable and that the disturbance attenuation is achieved.

Fig. 1. States of the closed loop system: single sample path simulation

Fig. 2. Evolution of the controlled outputs $z_{\infty t}$: single sample path simulation

7 Conclusion

In this paper, the static output feedback $\mathcal{H}_2/\mathcal{H}_\infty$ control of continuous time reconfigurable systems with Markovian processes and state-dependent noise was considered. We have proposed a synthesis method of reconfigurable controllers that take explicitly into account the inherent imperfections of the FDI process. Another problematic related to MJLS control was also addressed. This problematic concerns the mode-independent static output feedback $\mathcal{H}_2/\mathcal{H}_\infty$ control of MJLS with multiplicative noise. The numerical resolution of the obtained results was done using a cone complementary algorithm and its running was illustrated on a numerical example.

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