Stabilization of Systems with Asynchronous Sensors and Controllers [★]

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Abstract

We study the stabilization of networked control systems with asynchronous sensors and controllers. Offsets between the sensor and controller clocks are unknown and modeled as parametric uncertainty. First we consider multi-input linear systems and provide a sufficient condition for the existence of linear time-invariant controllers that are capable of stabilizing the closed-loop system for every clock offset in a given range of admissible values. For first-order systems, we next obtain the maximum length of the offset range for which the system can be stabilized by a single controller. Finally we illustrate the results with a numerical simulation.

Key words: Networked control systems, clock offsets, parametric uncertainty.

1 Introduction

In networked and embedded control systems, the outputs of plants are often sampled in a nonperiodic fashion and sent to controllers with time-varying delays. To address robust control with such imperfections, various techniques have been developed, for example, the input-delay approach [10,20], the gridding approach [7,11,23], and the impulsive systems approach based on Lyapunov functionals [21], on looped functionals [4], and on clock-dependent Lyapunov functions [3]; see also the surveys [16,17]. In contrast to the references mentioned above, here we assume that time-stamps are used to provide the controller with information about the sampling times and the communication delays incurred by each mea-

surement. In this approach, sensors send measurements to controllers together with time-stamps, and the controllers exploit this information to mitigate the effect of variable delays and sampling periods [12,14,22]. However, when the local clocks at the sensors and at the controllers are not synchronized, the time-stamps and the true sampling instants do not match. Protocols to establish synchronization have been actively studied as surveyed in [26], and synchronization by the global positioning system (GPS) or radio clocks has been utilized in some systems. Nevertheless, synchronizing clocks over networks has fundamental limits [9], and a recent study [18] has shown that synchronization based on GPS signals is vulnerable against attacks.

In this paper, we study the stabilization problem of systems with asynchronous sensing and control. We assume that the controller can use the time-stamps but does not know the offset between the sensor and controller clocks, but we do assume that this offset is essentially constant over the time scales of interest. Our objective is to find linear time-invariant (LTI) controllers that achieve closed-loop stability for every clock offset in a given range.

We formulate the stabilization of systems with clock offsets as the problem of stabilizing systems with para-

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metric uncertainty, which can be regarded as the simultaneous stabilization of a family of plants, as studied in [30, Sec. 5.4] and [31]. However, we had to overcome a few technical difficulties that distinguish the problem considered here from previously published results:

Infinitely many plants: We consider a family of plant models that is indexed by a continuous-valued parameter. Such a family includes infinitely many plants, but the approaches for simultaneous stabilization e.g., in [28] exploit the property that the number of plant models is finite.

Nonlinearity of the uncertain parameter: In this work, the uncertain parameter appears in a non-linear form. Therefore, it is not suitable to use the techniques based on linear matrix inequalities (LMIs) in [6] for the robust stabilization of systems with polytopic uncertainties. Although the robust stability analysis based on continuous paths of systems with respect to the ν -gap metric was developed in [5], controller designs based on this approach have not been fully investigated.

Common unstable poles and zeros: Earlier studies on simultaneous stabilization consider a restricted class of plants. For example, the sufficient condition in [2] is obtained for a family of plants with no common unstable zeros or poles. The set of plants in [19] has common unstable zeros (or poles) but all the plants are stable (or minimum-phase). These assumptions are not satisfied for the systems in the present paper.

We make the following technical contributions for multi-input systems and first-order systems: First we consider multi-input systems and obtain a sufficient condition for stabilization with asynchronous sensing and control. We construct a stabilizing controller from the solution of an appropriately defined $\mathcal{H}^\infty$ control problem. The above mentioned difficulties found in the simultaneous stabilization problem we consider is circumvented by exploiting geometric properties on $\mathcal{H}^\infty$. For first-order systems, we obtain an explicit formula for the exact bound on the clock offset that can be allowed for stability. This result is based on the stabilization of interval systems [13, 25], to which our problem can be reduced for first-order plants. We start by formulating the problem in the context of state feedback without disturbances and noise, but we show in Section 3.2 that the above results also apply for output feedback with disturbances and noise.

The authors in the previous study [24] have considered systems with time-varying clock offsets and have proposed a stabilization method with causal controllers, based on the analysis of data rate limitations in quantized control. The stability analysis and the $\mathcal{L}^2$ -gain analysis of systems with variable clock offsets have been investigated in [33] and [35], respectively. The major difference with respect to those studies is that here we consider only constant offsets but design stabilizing LTI

controllers. This paper is based on the conference paper [34], but here we extend the preliminary results for single-input systems to the multi-input case.

The remainder of the paper is organized as follows. Section 2 introduces the closed-loop system we consider and presents the problem formulation. Section 3 is devoted to the discretization of the closed-loop system. In Section 4, we obtain a sufficient condition for the stabilizability of general-order systems. In Section 5, we derive the exact bound on the permissible clock offset for first-order systems. A numerical example is presented in Section 6.

Notation and definitions: We denote by $\mathbb{Z}_+$ the set of non-negative integers. The symbols $\mathbb{D}$ and $\mathbb{T}$ denote the open unit disc $\{z \in \mathbb{C} : |z| < 1\}$ and the unit circle $\{z \in \mathbb{C} : |z| = 1\}$, respectively. We denote by $\mathbb{D}^c$ the complement of the open unit disc $\{z \in \mathbb{C} : |z| \geq 1\}$.

A square matrix F is said to be *Schur stable* if all its eigenvalues lie in the unit disc $\mathbb{D}$. We say that a discrete-time LTI system $\xi_{k+1} = F\xi_k + Gu_k$, $y_k = H\xi_k$ is stabilizable (detectable) if there exists a matrix K (L) such that $F - GK$ ($F - LH$) is Schur stable. We also use the terminology (F, G) is stabilizable (respectively, (F, H) is detectable) to denote this same concept.

We denote by $\mathcal{RH}^\infty$ the space of all bounded holomorphic real-rational functions in $\mathbb{D}$. The field of fractions of $\mathcal{RH}^\infty$ is denoted by $\mathcal{RF}^\infty$. For a commutative ring R , $\mathbf{M}(R)$ denotes the set of matrices with entries in R , of any order. For $M \in \mathbf{M}(\mathbb{C})$, $\|M\|$ denotes the induced 2-norm. For $G \in \mathbf{M}(\mathcal{RH}^\infty)$, the $\mathcal{H}^\infty$ -norm is defined as $\|G\|_\infty = \sup_{z \in \mathbb{D}} \|G(z)\|$. For $G = \begin{bmatrix} G_{11} & G_{12} \\ G_{21} & G_{22} \end{bmatrix} \in \mathbf{M}(\mathcal{RF}^\infty)$ and $Q \in \mathbf{M}(\mathcal{RF}^\infty)$, we define a lower linear fractional transformation of G and Q as $\mathcal{F}_\ell(G, Q) := G_{11} + G_{12}Q(I - G_{22}Q)^{-1}G_{21}$.

A pair (N, D) in $\mathbf{M}(\mathcal{RH}^\infty)$ is said to be *right coprime* if the Bezout identity $XN + YD = I$ holds for some $X, Y \in \mathbf{M}(\mathcal{RH}^\infty)$. $P \in \mathbf{M}(\mathcal{RF}^\infty)$ admits a *right coprime factorization* if there exist $D, N \in \mathbf{M}(\mathcal{RH}^\infty)$ such that $P = ND^{-1}$ and the pair (N, D) is right coprime. Similarly, a pair $(\tilde{D}, \tilde{N})$ in $\mathbf{M}(\mathcal{RH}^\infty)$ is *left coprime* if the Bezout identity $\tilde{N}\tilde{X} + \tilde{D}\tilde{Y} = I$ holds for some $\tilde{X}, \tilde{Y} \in \mathbf{M}(\mathcal{RH}^\infty)$. $P \in \mathbf{M}(\mathcal{RF}^\infty)$ admits a *left coprime factorization* if there exist $\tilde{D}, \tilde{N} \in \mathbf{M}(\mathcal{RH}^\infty)$ such that $P = \tilde{D}^{-1}\tilde{N}$ and the pair $(\tilde{D}, \tilde{N})$ is left coprime. If P is a scalar-valued function, then we use the expressions *coprime* and *coprime factorization*.

2 Problem Statement

Consider the following LTI plant:

$$\Sigma_P : \dot{x}(t) = Ax(t) + Bu(t), \quad (1)$$

Fig. 1. Closed-loop system with a time-stamp aware estimator.

where $x(t) \in \mathbb{R}^n$ and $u(t) \in \mathbb{R}^m$ are the state and the input of the plant, respectively. As shown in Fig. 1, this plant is connected through a sampler and a zero-order hold (ZOH) to a time-stamp aware estimator and a controller, which will be described soon.

Let $s_1, s_2, \dots$ be sampling instants from the perspective of the controller clock. A sensor measures the state $x(s_k)$ and sends it to a controller together with a time-stamp. However, since the sensor and the controller may not be synchronized, the time-stamp determined by the sensor typically includes an unknown offset with respect to the controller clock. In this paper, we assume that the clock offset is constant. Although clock properties are affected by environment such as temperature and humidity, the change of such properties is slow for the time scales of interest. Furthermore, the difference of clock frequencies can be ignored. This is justified by noting that time synchronization techniques, like the one proposed in [15], can achieve asymptotic convergence of the clock frequencies (in the mean-square sense), even in the presence of random network delays. We thus assume that the time-stamp $\hat{s}_k$ reported by the sensor is given by

$$\hat{s}_k = s_k + \Delta \quad (k \in \mathbb{N}) \quad (2)$$

for some unknown constant $\Delta \in \mathbb{R}$.

Let $h > 0$ be the update period of the ZOH. The control signal $u(t)$ is assumed to be piecewise constant and updated periodically at times $t_k = kh$ ($k \in \mathbb{N}$) with values u_k computed by the controller: $u(t) = u_k$ for $t \in [t_k, t_{k+1})$. We place a basic assumption for stabilization of sampled-data systems.

Assumption 1 (*Stabilizability and non-pathological control update*) The plant (A, B) is stabilizable and the update period h is non-pathological, that is, $(\lambda_p - \lambda_q)h \neq 2\pi j\ell$ ($\ell = \pm 1, \pm 2, \dots$) for each pair (λ_p, λ_q) of eigenvalues of A .

While the ZOH updates the control signal $u(t)$ periodically, the true sampling times s_k and the reported sampling times $\hat{s}_k$ may not be periodic. However, we do assume that both s_k and $\hat{s}_k$ do not fall behind t_k by more than the ZOH update period h . This assumption is formally stated as follows.

Fig. 2. Sampling instants s_k , reported time-stamps $\hat{s}_k$, and updating instants t_k of the zero-order hold.

Assumption 2 (*Bounded clock offset*) For every $k \in \mathbb{Z}_+$, $s_k, \hat{s}_k \in [t_k, t_{k+1})$.

This assumption implies that the clock offset Δ is smaller than the control update period h , which holds in most mechatronics systems. In fact, control update periods for mechatronics systems generally take values from 100 μs to 10 ms, while recent clock synchronization algorithms such as the IEEE 1588 Precision Time Protocol (PTP) [1] make clock offsets smaller than a few tens of microseconds.

Fig. 2 shows the timing diagram of the sampling instants s_k , the reported time-stamps $\hat{s}_k$, and updating instants t_k of the control inputs.

The controller side is comprised of a time-stamp aware estimator and a controller as in the model-based or emulation-based control of networked control systems [12]. The time-stamp aware estimator generates the state estimate $\hat{x}(t_{k+1}) \in \mathbb{R}^n$ from the data $(x(s_k), \hat{s}_k)$ according to the following dynamics:

$$\Sigma_E : \begin{cases} \dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) & (t_k < t \leq t_{k+1}) \\ \hat{x}(\hat{s}_k) = x(s_k), & (k \in \mathbb{Z}_+). \end{cases} \quad (3)$$

Note that if the time-stamp is correct, i.e., $s_k = \hat{s}_k$, then this estimator consistently produces $\hat{x}(t) = x(t)$ for all t , perfectly compensating transmission delays. Time-stamp aware estimators have been used to compensate for network-induced imperfections, e.g., in [12, 14, 22].

The controller is a discrete-time LTI system and generates the control input u_k based on the state estimate $\hat{x}_k := \hat{x}(t_k)$:

$$\Sigma_C : \begin{cases} \zeta_{k+1} = A_c \zeta_k + B_c \hat{x}_k \\ u_k = C_c \zeta_k + D_c \hat{x}_k, \end{cases} \quad (4)$$

where $\zeta_k \in \mathbb{R}^{n_c}$ is the state of the controller.

The objective of the present paper is to find a discrete-time LTI controller Σ_C as in (4) that achieves closed-loop stability for every clock offset in a given range of admissible values. Specifically, we want to solve the following problem:

Problem 3 *Given an offset interval $[\underline{\Delta}, \overline{\Delta}]$, determine if there exists a controller Σ_C as in (4) such that $x(t), \hat{x}(t) \rightarrow 0$ as $t \rightarrow \infty$ and $\zeta_k \rightarrow 0$ as $k \rightarrow \infty$ for every*

$\Delta \in [\underline{\Delta}, \overline{\Delta}]$ and for every initial states $x(0)$ and ζ_0 . Furthermore, if one exists, find such a controller Σ_C .

3 Discretization of the Closed-loop System

To solve Problem 3, we discretize the system comprised of the plant Σ_P , the estimator Σ_E , the ZOH, and the sampler. In this section, we obtain a realization for the discretized system and describe its basic properties related to stability, stabilizability, and detectability. Moreover, we extend the discretized system to scenarios with disturbances/noise and output feedback.

3.1 Discretized system and its basic properties

The following lemma provides a realization for the discretized system:

Lemma 4 *Define*

$$\xi_k := \begin{bmatrix} x(t_k) - \hat{x}(t_k) \\ \hat{x}(t_k) \end{bmatrix}.$$

The dynamics of the discretized system Σ_d comprised of the plant Σ_P , the estimator Σ_E , the ZOH, and the sampler can be described by the following equations:

$$\Sigma_d : \xi_{k+1} = F_\Delta \xi_k + G_\Delta u_k, \quad \eta_k = H_\Delta \xi_k, \quad (5)$$

where $\Lambda := e^{Ah}$, $\Theta := e^{-A\Delta} - I$, and

$$F_\Delta := \begin{bmatrix} -\Lambda\Theta & -\Lambda\Theta \\ \Lambda(I + \Theta) & \Lambda(I + \Theta) \end{bmatrix}, \quad H_\Delta := \begin{bmatrix} 0 & I \end{bmatrix}$$

$$G_\Delta := \begin{bmatrix} \Lambda \left(\int_0^h e^{-A\tau} d\tau - (I + \Theta) \int_0^{h-\Delta} e^{-A\tau} d\tau \right) B \\ \Lambda(I + \Theta) \int_0^{h-\Delta} e^{-A\tau} d\tau B \end{bmatrix}. \quad (6)$$

PROOF. Using $\Lambda = e^{Ah}$, we have from the state equation (1) that

$$x(t_{k+1}) = \Lambda x(t_k) + \Lambda \int_0^h e^{-A\tau} d\tau \cdot B u_k. \quad (7)$$

We compute $\hat{x}(t_{k+1})$ in terms of $x(t_k)$ and u_k . It follows from the dynamics of the estimator Σ_E in (3) that

$$\hat{x}(t_{k+1}) = e^{A(t_{k+1}-\hat{s}_k)} \hat{x}(\hat{s}_k) + \int_{\hat{s}_k}^{t_{k+1}} e^{A(t_{k+1}-\tau)} B d\tau \cdot u_k \quad (8)$$

and

$$\hat{x}(\hat{s}_k) = x(s_k) = e^{A(s_k-t_k)} x(t_k) + \int_{t_k}^{s_k} e^{A(s_k-\tau)} B d\tau \cdot u_k. \quad (9)$$

Since $t_{k+1} - t_k = h$ and $\hat{s}_k = s_k + \Delta$, it follows that

$$e^{A(t_{k+1}-\hat{s}_k)} \cdot e^{A(s_k-t_k)} = e^{A(h-\Delta)}, \quad (10)$$

and also that

$$e^{A(t_{k+1}-\hat{s}_k)} \int_{t_k}^{s_k} e^{A(s_k-\tau)} d\tau = \int_{t_k+\Delta}^{\hat{s}_k} e^{A(t_{k+1}-\tau)} d\tau. \quad (11)$$

Using $\Lambda = e^{Ah}$ and $\Theta = e^{-A\Delta} - I$, we conclude from (8)–(11) that

$$\hat{x}(t_{k+1}) = \Lambda(I + \Theta)x(t_k) + \Lambda(I + \Theta) \int_0^{h-\Delta} e^{-A\tau} B d\tau \cdot u_k. \quad (12)$$

From (7) and (12), we obtain the F_Δ and G_Δ in (6). Moreover, we have $H_\Delta = [0 \ I]$ by the definition of the extended state ξ_k . $\square$

Next we show that if the extended state ξ_k and the controller state ζ_k converge to the origin, then the intersample values of x and $\hat{x}$ also converge to the origin.

Proposition 5 *For the discreteized system Σ_d in Lemma 4, we have that $\xi_k, \zeta_k \rightarrow 0$ as $k \rightarrow \infty$ if and only if $x(t), \hat{x}(t) \rightarrow 0$ as $t \rightarrow \infty$ and $\zeta_k \rightarrow 0$ as $k \rightarrow \infty$.*

PROOF. The statement that $x(t), \hat{x}(t) \rightarrow 0$ as $t \rightarrow \infty$ and $\zeta_k \rightarrow 0$ as $k \rightarrow \infty$ imply $\xi_k, \zeta_k \rightarrow 0$ as $k \rightarrow \infty$, follows directly from the definition of ξ_k .

To prove the converse statement, assume that $\xi_k, \zeta_k \rightarrow 0$ as $k \rightarrow \infty$. Then $\hat{x}(t_k) = H_\Delta \xi_k \rightarrow 0$ and $x(t_k), u_k \rightarrow 0$ as $k \rightarrow \infty$. Since

$$\|x(t_k + \tau)\| \leq e^{\|A\|h} \|x(t_k)\| + \int_0^h e^{\|A\|h} \|B\| dt \cdot \|u_k\|$$

for all $k \in \mathbb{Z}_+$ and all $\tau \in [0, h)$, we derive $x(t) \rightarrow 0$ ($t \rightarrow \infty$). Similarly, we see from the dynamics of the estimator Σ_E that $\hat{x}(t) \rightarrow 0$ as $t \rightarrow \infty$. This completes the proof. $\square$

This proposition allows us to conclude Problem 3 can be solved by finding LTI controllers Σ_C achieving ξ_k ,

$\zeta_k \rightarrow 0$ ($k \rightarrow \infty$) for every $\Delta \in [\underline{\Delta}, \overline{\Delta}]$ and for every initial states ξ_0 and ζ_0 .

The following result allows us to conclude that the discretized system Σ_d is detectable and stabilizable for all Δ and almost all h if the plant (A, B) is stabilizable.

Proposition 6 *The discretized system Σ_d in (5) is detectable for all Δ and h . Moreover, Σ_d is stabilizable for all Δ if Assumption 1 holds.*

PROOF. Let us first obtain another realization $(\bar{F}_\Delta, \bar{G}_\Delta, \bar{H}_\Delta)$ of the discretized system Σ_d in (5). We can transform F_Δ into

$$\bar{F}_\Delta := T^{-1}F_\Delta T = \begin{bmatrix} \Lambda & 0 \\ 0 & 0 \end{bmatrix}, \text{ where } T := \begin{bmatrix} -\Theta & -I \\ I + \Theta & I \end{bmatrix}.$$

Furthermore, if we define

$$J_1 := \int_0^h e^{-A\tau} d\tau, \quad J_2 := \int_0^{h-\Delta} e^{-A\tau} d\tau,$$

then we obtain

$$\bar{G}_\Delta := T^{-1}G_\Delta = \begin{bmatrix} \Lambda J_1 B \\ -\Lambda(J_1 - (I + \Theta)J_2)B - \Theta \Lambda J_1 B \end{bmatrix}$$

and $\bar{H}_\Delta := H_\Delta T = [I + \Theta \quad I]$. We have thus another realization $(\bar{F}_\Delta, \bar{G}_\Delta, \bar{H}_\Delta)$ for Σ_d .

Next we check detectability and stabilizability by using the realization $(\bar{F}_\Delta, \bar{G}_\Delta, \bar{H}_\Delta)$. Define

$$L_\Delta := \begin{bmatrix} \Lambda(I + \Theta)^{-1} \\ 0 \end{bmatrix}. \quad (13)$$

Then we have that

$$\bar{F}_\Delta - L_\Delta \bar{H}_\Delta = \begin{bmatrix} 0 & -\Lambda(I + \Theta)^{-1} \\ 0 & 0 \end{bmatrix}, \quad (14)$$

and clearly $\bar{F}_\Delta - L_\Delta \bar{H}_\Delta$ is Schur stable. Therefore, the discretized system Σ_d is detectable for all h and Δ .

To show stabilizability, we use the well-known rank conditions (see, e.g., [38, Sec. 3.2]). We have that $[zI - \bar{F}_\Delta \quad \bar{G}_\Delta]$ is full row rank for all $z \in \mathbb{D}^c$ if and only if

$$\begin{bmatrix} zI - \Lambda & \Lambda J_1 B \end{bmatrix} = \begin{bmatrix} zI - e^{Ah} & \int_0^h e^{A\tau} B d\tau \end{bmatrix}$$

is full row rank for all $z \in \mathbb{D}^c$. Hence, the discretized system Σ_d is stabilizable for all Δ if Assumption 1 holds.

3.2 Extension to the output feedback case with disturbances and noise

Instead of Σ_P in (1), consider a plant Σ'_P with disturbances, noise, and output feedback:

$$\Sigma'_P : \begin{cases} \dot{x}(t) = Ax(t) + Bu(t) + d(t) \\ y(t) = Cx(t) + n(t), \end{cases}$$

where $d(t) \in \mathbb{R}^n$ and $n(t), y(t) \in \mathbb{R}^p$ are the disturbance, measurement noise, and output of the plant, respectively. As in [12, Chap. 3], [36], and the references therein, we assume that a smart sensor is co-located with the plant and that the sensor has the following observer to generate the state estimate, which is sampled and sent to the controller side:

$$\Sigma_O : \dot{\hat{x}}(t) = A\bar{x}(t) + Bu(t) + L(y(t) - C\bar{x}(t)),$$

where $\bar{x}(t) \in \mathbb{R}^n$ is the state estimate and L is an observer gain such that $A - LC$ is Hurwitz. The sampler sends the state estimate $\bar{x}$, and the resulting dynamics of the time-stamp aware estimator Σ'_E is provided by

$$\Sigma'_E : \begin{cases} \dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) & (t_k \leq t < t_{k+1}) \\ \hat{x}(\hat{s}_k) = \bar{x}(s_k) + w_k, & (k \in \mathbb{Z}_+), \end{cases}$$

where $w_k \in \mathbb{R}^n$ is the quantization noise. A calculation similar to the one performed in the proof of in Lemma 4 can be used to show that the dynamics of the discretized system Σ'_d is given by

$$\Sigma'_d : \xi_{k+1} = F_\Delta \xi_k + G_\Delta u_k + d_k, \quad \eta_k = H_\Delta \xi_k, \quad (15)$$

where $d_k := [d_{1,k}^\top \quad d_{2,k}^\top]^\top$ and

$$\begin{aligned} d_{1,k} &:= \int_{t_k}^{t_{k+1}} e^{A(t_{k+1}-\tau)} d(\tau) d\tau - d_{2,k} \\ d_{2,k} &:= -e^{A(t_{k+1}-\hat{s}_k)} \left(e^{(A-LC)(s_k-t_k)} e_k - w_k \right. \\ &\quad \left. + \int_{t_k}^{s_k} \left(e^{(A-LC)(s_k-\tau)} (d(\tau) - Ln(\tau)) \right. \right. \\ &\quad \left. \left. - e^{A(s_k-\tau)} d(\tau) \right) d\tau \right) \\ e_k &:= x(t_k) - \bar{x}(t_k). \end{aligned}$$

The only difference from the original idealized system Σ_d in (5) is that Σ'_d has the disturbance d_k . Hence, for the output feedback case with bounded disturbances and noise, solutions of Problem 3 achieve the boundedness of the closed-loop state.

Proposition 7 *Assume that $\xi_k, \zeta_k \rightarrow 0$ as $k \rightarrow \infty$ for the idealized system Σ_d in Lemma 4 (in the context of state feedback without disturbances and measurement*

noise). If $d(t)$, $n(t)$, and w_k are bounded for all $t \geq 0$ and all $k \in \mathbb{Z}_+$, then the states $x(t)$, $\bar{x}(t)$, $\hat{x}(t)$, and ζ_k are also bounded for all $t \geq 0$ and all $k \in \mathbb{Z}_+$. Moreover, if $d(t) = n(t) = w_k = 0$ for all $t \geq 0$ and all $k \in \mathbb{Z}_+$, then $x(t)$, $\bar{x}(t)$, $\hat{x}(t)$, and ζ_k converge to the origin.

PROOF. Since d_k is bounded for every $k \geq 0$ and every $s_k, \hat{s}_k \in [t_k, t_{k+1})$, it follows that ξ_k and ζ_k are also bounded for all $k \geq 0$. The rest of the proof follows the similar lines as that of Proposition 5, and hence it is omitted.

See also [35] for the $\mathcal{L}^2$ -gain analysis of systems with time-varying offsets.

4 Controller Design via Simultaneous Stabilization

4.1 Preliminaries

We first consider a general simultaneous stabilization problem not limited to the system introduced in Section 2.

The transfer function P of the system $\xi_{k+1} = F\xi_k + Gu_k$, $y_k = H\xi_k$ is usually defined by the Z-transform of the system's impulse response, i.e., $\tilde{H}(zI - F)^{-1}G$, but in this paper, we define the transfer function P by $P(z) := H(1/z \cdot I - F)^{-1}G$ for consistency of the Hardy space theory; see [30, Sec. 2.2] for details. Hence the transfer function of a causal system is not proper. We say that $C \in \mathbf{M}(\mathcal{RF}^\infty)$ stabilizes $P \in \mathbf{M}(\mathcal{RF}^\infty)$ if $(I + PC)^{-1}$, $C(I + PC)^{-1}$, and $(I + PC)^{-1}P$ belong to $\mathbf{M}(\mathcal{RH}^\infty)$. We recall that when these three transfer functions belong to $\mathbf{M}(\mathcal{RH}^\infty)$, they will have no poles in the closed unit disk.

Consider the family of plants $P_\theta \in \mathbf{M}(\mathcal{RF}^\infty)$ parameterized by $\theta \in S$, where S is a nonempty parameter set, and assume that we have a doubly coprime factorization of P_θ over $\mathcal{RH}^\infty$

$$\begin{bmatrix} Y_\theta & X_\theta \\ -\tilde{N}_\theta & \tilde{D}_\theta \end{bmatrix} \begin{bmatrix} D_\theta & -\tilde{X}_\theta \\ N_\theta & \tilde{Y}_\theta \end{bmatrix} = I, \quad (16)$$

where $P_\theta = N_\theta D_\theta^{-1}$ and $P_\theta = \tilde{D}_\theta^{-1} \tilde{N}_\theta$ are a right coprime factorization and a left coprime factorization, respectively. We explicitly construct the matrices in (16) using a stabilizable and detectable realization of P_θ ; see, e.g., [30, Theorem 4.2.1].

The following theorem provides a necessary and sufficient condition for simultaneous stabilization:

Theorem 8 ([30, 31]) *Given a nonempty set S , consider the plant P_θ having a doubly coprime factorization (16) for each $\theta \in S$. Fix $\theta_0 \in S$ and define*

$$\begin{bmatrix} U_\theta & V_\theta \end{bmatrix} := \begin{bmatrix} \tilde{D}_\theta & \tilde{N}_\theta \end{bmatrix} \begin{bmatrix} \tilde{Y}_{\theta_0} & -N_{\theta_0} \\ \tilde{X}_{\theta_0} & D_{\theta_0} \end{bmatrix} \quad (\theta \in S). \quad (17)$$

Then (V_θ, U_θ) is right coprime for every $\theta \in S$. Moreover, there exists a controller that stabilizes P_θ for every $\theta \in S$ if and only if there exists $Q \in \mathbf{M}(\mathcal{RH}^\infty)$ such that for all $\theta \in S$,

$$(U_\theta + V_\theta Q)^{-1} \in \mathbf{M}(\mathcal{RH}^\infty). \quad (18)$$

Such a stabilizing controller is given by

$$C := (\tilde{X}_{\theta_0} + D_{\theta_0} Q)(\tilde{Y}_{\theta_0} - N_{\theta_0} Q)^{-1}. \quad (19)$$

Remark 9 Although the simultaneous stabilization of a finite family of plants is considered in [30, Sec. 5.4] and [31], generalization to an arbitrary family of plants is readily apparent, as mentioned in the last paragraph of Section 3 in [31].

Remark 10 A left coprime factorization of stabilizing controllers is used in [30, Sec. 5.4] and [31], whereas we represent controllers by a right coprime factorization in (19). Therefore, Theorem 8 is slightly different from its counterpart in [30, Sec. 5.4] and [31].

4.2 Robust Controller Design

It is generally not easy to verify in a computationally efficient fashion that a transfer function Q satisfying (18) exists. In the next theorem, we develop a simple sufficient condition for (18) to hold, by exploiting geometric properties on $\mathcal{H}^\infty$ inspired by results on strong stabilization [37].

Theorem 11 *Given a nonempty set S , assume that each plant P_θ ($\theta \in S$) has a doubly coprime factorization (16) such that there exist $\theta_0 \in S$, $W \in \mathbf{M}(\mathcal{RH}^\infty)$, and $R(\theta) \in \mathbf{M}(\mathbb{R})$ satisfying $\tilde{D}_\theta = \tilde{D}_{\theta_0}$ and*

$$\tilde{N}_\theta(z) - \tilde{N}_{\theta_0}(z) = R(\theta)W(z), \quad (20)$$

for all $\theta \in S$. If there exists $Q \in \mathbf{M}(\mathcal{RH}^\infty)$ satisfying the following $\mathcal{H}^\infty$ -norm condition:

$$\|W(\tilde{X}_{\theta_0} + D_{\theta_0} Q)\|_\infty < \frac{1}{\sup_{\theta \in S} \|R(\theta)\|}, \quad (21)$$

then Q satisfies (18), and hence the controller C in (19) stabilizes P_θ for every $\theta \in S$.

PROOF. We define U_θ and V_θ as in (17). Since $\tilde{D}_\theta = \tilde{D}_{\theta_0}$, it follows from (17) and the Bezout identity

$\tilde{D}_{\theta_0}\tilde{Y}_{\theta_0} + \tilde{N}_{\theta_0}\tilde{X}_{\theta_0} = I$ in (16) that

$$U_\theta = \tilde{D}_\theta\tilde{Y}_{\theta_0} + \tilde{N}_\theta\tilde{X}_{\theta_0} = I + (\tilde{N}_\theta - \tilde{N}_{\theta_0})\tilde{X}_{\theta_0}.$$

Moreover, since $\tilde{D}_{\theta_0}^{-1}\tilde{N}_{\theta_0} = N_{\theta_0}D_{\theta_0}^{-1}$, we obtain

$$V_\theta = -\tilde{D}_\theta N_{\theta_0} + \tilde{N}_\theta D_{\theta_0} = (\tilde{N}_\theta - \tilde{N}_{\theta_0})D_{\theta_0}.$$

Hence $U_\theta + V_\theta Q = I + (\tilde{N}_\theta - \tilde{N}_{\theta_0})(\tilde{X}_{\theta_0} + D_{\theta_0}Q)$. Since $(I + \Phi)^{-1} \in \mathbf{M}(\mathcal{RH}^\infty)$ for all $\Phi \in \mathbf{M}(\mathcal{RH}^\infty)$ satisfying $\|\Phi\|_\infty < 1$, it follows that if

$$\|(\tilde{N}_\theta - \tilde{N}_{\theta_0})(\tilde{X}_{\theta_0} + D_{\theta_0}Q)\|_\infty < 1 \quad (\theta \in S), \quad (22)$$

then (18) holds for all $\theta \in S$. From the assumption (20),

$$\begin{aligned} & \|(\tilde{N}_\theta - \tilde{N}_{\theta_0})(\tilde{X}_{\theta_0} + D_{\theta_0}Q)\|_\infty \\ & \leq \|R(\theta)\| \cdot \|W(\tilde{X}_{\theta_0} + D_{\theta_0}Q)\|_\infty. \end{aligned}$$

Hence if Q satisfies (21) for all $\theta \in S$, then (22) holds, and consequently P_θ is simultaneously stabilizable by C in (19) from Theorem 8. $\square$

The proposition below shows that our discretized system Σ_d in (5) always satisfies the assumptions on $\tilde{D}_\theta$ and $\tilde{N}_\theta$ that appear in Theorem 11. This result also provides the matrices R and W in (20) without explicitly calculating a coprime factorization of P_θ for all $\theta \in S$.

Proposition 12 *Define the transfer function $P_\Delta(z) := H_\Delta(1/z \cdot I - F_\Delta)^{-1}G_\Delta$. For all $\Delta \in (-h, h)$, there exists a doubly coprime factorization (16) such that $\tilde{D}_\Delta(z) = \tilde{D}_0(z) = I - ze^{A_h}$, and (20) holds with*

$$R(\Delta) := \int_0^\Delta e^{A(h-\tau)} B d\tau \in \mathbb{R}^{n \times m} \quad (23)$$

$$W(z) := z(z-1) \in \mathcal{RH}^\infty. \quad (24)$$

PROOF. Consider the realization $(\bar{F}_\Delta, \bar{G}_\Delta, \bar{H}_\Delta)$ in the proof of Proposition 6. For every $\Delta \in (-h, h)$, the matrix L_Δ in (13) achieves the Schur stability of $\bar{F}_\Delta - L_\Delta \bar{H}_\Delta$ as shown in (14). From the realization of $\tilde{D}_\Delta$, e.g., in [30, Theorem 4.2.1], we can write $\tilde{D}_\Delta$ as

$$\tilde{D}_\Delta(z) = I - \bar{H}_\Delta(1/z \cdot I - (\bar{F}_\Delta - L_\Delta \bar{H}_\Delta))^{-1}L_\Delta = I - z\Lambda.$$

Noticing that the far right-hand side of the equation above does not depend on Δ , we have $\tilde{D}_\Delta(z) = \tilde{D}_0(z) = I - z\Lambda$.

It follows that $\tilde{N}_\Delta - \tilde{N}_0 = \tilde{D}_0(P_\Delta - P_0)$. From the realization $(\bar{F}_\Delta, \bar{G}_\Delta, \bar{H}_\Delta)$, we see that

$$\begin{aligned} P_\Delta(z) &= z((I + \Theta)(I - z\Lambda)^{-1}\Lambda J_1 \\ &\quad - \Lambda(J_1 - (I + \Theta)J_2) - \Theta\Lambda J_1)B. \end{aligned}$$

Since $\Theta = 0$ and $J_2 = J_1$ for $\Delta = 0$, it follows that

$$\begin{aligned} P_\Delta(z) - P_0(z) &= z(\Theta(I - z\Lambda)^{-1}\Lambda J_1 \\ &\quad - \Lambda(J_1 - (I + \Theta)J_2) - \Theta\Lambda J_1)B. \end{aligned} \quad (25)$$

On the other hand, we have $J_1 - (I + \Theta)J_2 = \int_0^\Delta e^{-A\tau} d\tau =: \bar{\Theta}$, and

$$\bar{\Theta}A = - \int_0^\Delta \left(\frac{d}{d\tau} e^{-A\tau} \right) d\tau = -\Theta. \quad (26)$$

Since $A(I - z\Lambda) = A - zAe^{A_h} = A - ze^{A_h}A = (I - z\Lambda)A$, it follows that $A(I - z\Lambda)^{-1} = (I - z\Lambda)^{-1}A$. Therefore we derive from (25)

$$P_\Delta - P_0 = z\Lambda\bar{\Theta}(I - z\Lambda)^{-1}(-I + z\Lambda(I - AJ_1))B.$$

Similarly to (26), we have $I - AJ_1 = \Lambda^{-1}$, and hence $P_\Delta - P_0 = z(z-1)\Lambda\bar{\Theta}(I - z\Lambda)^{-1}B$. Since Λ , $\bar{\Theta}$, and $(I - z\Lambda)^{-1}$ are commutative, we derive

$$P_\Delta - P_0 = z(z-1)(I - z\Lambda)^{-1}\bar{\Theta}\Lambda B,$$

and $\tilde{N}_\Delta - \tilde{N}_0 = z(z-1)\bar{\Theta}\Lambda B$. Thus (20) holds with R in (23) and W in (24) $\square$

Define

$$\gamma := \frac{1}{\max_{\Delta \in [\underline{\Delta}, \bar{\Delta}]} \left\| \int_0^\Delta e^{A(h-\tau)} B d\tau \right\|}. \quad (27)$$

From Theorem 11, to obtain a controller Σ_C as in (4), it is enough to solve the following suboptimal problem: Find $Q \in \mathbf{M}(\mathcal{RH}^\infty)$ satisfying $\|W(\tilde{X}_0 - D_0Q)\|_\infty < \gamma$. This problem is equivalent to a standard suboptimal $\mathcal{H}^\infty$ control problem [38, Chaps. 16, 17]: Find $Q \in \mathbf{M}(\mathcal{RH}^\infty)$ such that $\|\mathcal{F}_\ell(\Phi, Q)\|_\infty < \gamma$, where Φ is defined by

$$\Phi := \begin{bmatrix} W\tilde{X}_0 & WD_0 \\ -I & 0 \end{bmatrix}. \quad (28)$$

The results of this section can be summarized through the following controller design algorithm:

Algorithm 1 (1) Using the realization

$$(\bar{F}_0, \bar{G}_0, \bar{H}_0) = \left(\begin{bmatrix} e^{Ah} & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} \int_0^h e^{A\tau} B d\tau \\ 0 \end{bmatrix}, \begin{bmatrix} I & I \end{bmatrix} \right),$$

the matrix $L_0 = [e^{A^\top h} \ 0]^\top$, and an arbitrary matrix K_0 such that $\Phi := \bar{F}_0 - \bar{G}_0 K_0$ is Schur stable, set

$$\begin{aligned} D_0(z) &:= I - K_0(1/z \cdot I - \Phi)^{-1} \bar{G}_0 \\ N_0(z) &:= \bar{H}_0(1/z \cdot I - \Phi)^{-1} \bar{G}_0 \\ \tilde{X}_0(z) &:= K_0(1/z \cdot I - \Phi)^{-1} L_0 \\ \tilde{Y}_0(z) &:= I + H_0(1/z \cdot I - \Phi)^{-1} K_0 \\ W(z) &:= z(z-1). \end{aligned}$$

- (2) For a given offset interval $[\underline{\Delta}, \bar{\Delta}]$, set γ as in (27), and solve the $\mathcal{H}^\infty$ control problem [38, Chaps. 16, 17]: Find $Q \in \mathbf{M}(\mathcal{RH}^\infty)$ such that $\|\mathcal{F}_\ell(\Phi, Q)\|_\infty < \gamma$, where Φ is defined by (28).
(3) If the $\mathcal{H}^\infty$ control problem is not solvable, then the algorithm fails. Otherwise the transfer function C of the controller Σ_C is given by $C = (\tilde{X}_0 + D_0 Q)(\tilde{Y}_0 - N_0 Q)^{-1}$.

Remark 13 We have from Proposition 12 that $P_\Delta = P_0 + W D_0^{-1} R(\Delta)$ for constant Δ , where P_Δ is expressed as the nominal component P_0 plus the uncertainty block $W D_0^{-1} R(\Delta)$. If we obtain a similar formula for the case of time-varying offsets as studied for systems with aperiodic sampling in [11], we can deal with the stabilization problem of systems with time-varying offsets through a small gain theorem. Although the uncertainty part of the discretized system Σ_d may be non-causal, the small gain theorem for systems with non-causal uncertainty in [29] can be used. This extension is a subject for future research.

5 Exact Bound on Offsets for First-order Systems

In this section, the plant class is restricted to first-order systems, and we provide an explicit formula for the exact bound on the clock offset that LTI controllers can allow.

Consider an unstable first-order plant: $\dot{x} = ax + bu$ with a scalar $a > 0$. If $a < 0$, the stabilization problem is trivial because a zero control input $u(t) = 0$ leads to the stability of the closed-loop system. Hence in the remainder of this section, we focus our attention on the case $a > 0$.

The following theorem provides the exact bound on clock offsets for first-order systems:

Theorem 14 Assume $-h < \underline{\Delta} < 0 < \bar{\Delta} < h$. There exists a controller that stabilizes the discretized system

Σ_d in (5) for all $\Delta \in [\underline{\Delta}, \bar{\Delta}]$ if and only if

$$\bar{\Delta} - \underline{\Delta} < \frac{2(\log(e^{ah} + 1) - \log(e^{ah} - 1))}{a}. \quad (29)$$

PROOF. As in Section 3, taking the Z-transform and then mapping $z \mapsto 1/z$, we obtain the transfer function P_Δ of the discretized system Σ_d :

$$P_\Delta(z) = \frac{b}{a} \left(\frac{(e^{ah} - 1)z}{1 - e^{ah}z} - (e^{-a\Delta} - 1) \frac{e^{ah}z(z-1)}{1 - e^{ah}z} \right), \quad (30)$$

which belongs to the so-called interval system [13, 25]. From the stabilization results in [13, 25], there exists a controller stabilizing P_Δ for all $\Delta \in [\underline{\Delta}, \bar{\Delta}]$ if and only if the associated Pick matrix is positive definite, i.e.,

$$\begin{bmatrix} 1 & 1 \\ 1 & \frac{1-\phi^2}{1-1/e^{2ah}} \end{bmatrix} > 0, \quad (31)$$

where $\phi := (1 - e^{a(\bar{\Delta}-\underline{\Delta})/2}) / (1 + e^{a(\bar{\Delta}-\underline{\Delta})/2})$. From the Schur complement formula, (31) is equivalent to

$$\frac{1 - \phi^2}{1 - 1/e^{2ah}} > 1. \quad (32)$$

We see that (32) is

$$e^{a(\bar{\Delta}-\underline{\Delta})} < \left(\frac{e^{ah} + 1}{e^{ah} - 1} \right)^2.$$

Taking the logarithm function of both sides gives the desired conclusion. $\square$

Remark 15 Although here we use the results of [13, 25], an alternative proof based on Theorem 8 is presented in [32].

Remark 16 In the case $a = 0$, the transfer function of the discretized system, P_Δ , is given by

$$P_\Delta(z) = \frac{hz}{1-z} - \Delta z.$$

Similarly to the case $a > 0$, one can show that there exists a controller stabilizing P_Δ for all $\Delta \in (-h, h)$. This result is consistent with that in the case when $a \rightarrow 0$ in Theorem 14, but we omit the proof for brevity.

6 Numerical Example

Consider the unstable batch reactor studied in [27], where the system matrices A and B in (1) are given by

$$A := \begin{bmatrix} 1.38 & -0.2077 & 6.715 & -5.676 \\ -0.5814 & -4.29 & 0 & 0.675 \\ 1.067 & 4.273 & -6.654 & 5.893 \\ 0.048 & 4.273 & 1.343 & -2.104 \end{bmatrix}$$

$$B := \begin{bmatrix} 0 & 0 \\ 5.679 & 0 \\ 1.136 & -3.146 \\ 1.136 & 0 \end{bmatrix}.$$

This example has been developed over the years as a benchmark example for networked control systems, and its data were transformed by a change of basis and time scale [27].

Here we compare the proposed method with the robust stabilization method in [8] and [30, Chap. 7] based on the following fact: *Consider a family of plants $P_\Delta \in \mathbf{M}(\mathcal{RH}^\infty)$ with $\Delta \in [\underline{\Delta}, \bar{\Delta}]$. Assume that P_Δ has no poles on $\mathbb{T}$ and the same number of unstable poles for every $\Delta \in [\underline{\Delta}, \bar{\Delta}]$ and that a function $r \in \mathcal{RH}^\infty$ satisfies*

$$\|P_\Delta(e^{j\omega}) - P_0(e^{j\omega})\| < |r(e^{j\omega})| \quad (33)$$

for all $\Delta \in [\underline{\Delta}, \bar{\Delta}]$ and all $\omega \in [0, 2\pi]$. If the controller $C \in \mathbf{M}(\mathcal{RH}^\infty)$ stabilizes P_0 and satisfies

$$\|rC(I - P_0C)^{-1}\|_\infty \leq 1, \quad (34)$$

then C stabilizes P_Δ for all $\Delta \in [\underline{\Delta}, \bar{\Delta}]$. The order of such a controller is typically equal to the order of the following transfer function:

$$\begin{bmatrix} 0 & rI \\ I & P_0 \end{bmatrix}.$$

We compute the length of the allowable offset interval $[\underline{\Delta}, \bar{\Delta}]$ obtained from the sufficient condition (21) for each $h \in [0.2, 3.6]$, which is shown as the solid line in Fig. 3. On the other hand, the dashed line in the figure represents the length of the offset interval $[\underline{\Delta}, \bar{\Delta}]$ obtained from the robust control approach that leads to the condition (34) with an appropriate function $r \in \mathcal{RH}^\infty$ satisfying (33). For example, we use $r(z) = 0.1766(z-1)/(z-0.9389)$ for $h = 1$ and $[\underline{\Delta}, \bar{\Delta}] =$

$[-0.02, 0.02]$, and this r satisfies (33) and

$$\frac{1}{2\pi} \int_0^{2\pi} (|r(e^{j\omega})| - \|P_\Delta(e^{j\omega}) - P_0(e^{j\omega})\|) d\omega \leq 8.5 \times 10^{-3}$$

for all $\Delta \in [\underline{\Delta}, \bar{\Delta}] = [-0.02, 0.02]$. The solid line is obtained by finding the maximum and minimum of Δ that satisfies the condition

$$\left\| \int_0^\Delta e^{A(h-\tau)} B d\tau \right\| \leq \frac{1}{\min_{Q \in \mathcal{RH}^\infty} \|\mathcal{F}_\ell(\Phi, Q)\|_\infty},$$

whereas to derive the dashed line, we first calculate r satisfying (33) for a fixed $[\underline{\Delta}, \bar{\Delta}]$ and then check the existence of a controller C that stabilizes P_0 and achieves the $\mathcal{H}^\infty$ -norm condition (34). We see from Fig. 3 that the proposed sufficient condition (21) is less conservative than (34).

Consider the case $h = 1$, and let C_1 and C_2 be controllers that are obtained from the sufficient conditions (21) and (34) with the maximum offset length, respectively. The order of the controller C_1 is 7, but applying balanced model truncation [38, Chap. 6] to the controller C_1 , we can obtain an approximated controller C_{app} with order 5, which satisfies $\|C_{\text{app}} - C_1\|_\infty / \|C_1\|_\infty = 0.023$. From iterative calculations of the eigenvalues of the discretized closed-loop system for each Δ , we find that both C_1 and C_{app} stabilize the discretized system Σ_d in (5) for all $\Delta \in (-1, 1)$. The controller C_2 has order 5 and allows the offsets $\Delta \in [-0.054, 0.068]$ without compromising the closed-loop stability. Approximated controllers with any order obtained by applying balanced model truncation to C_2 do not achieve the closed-loop stability even in the case $\Delta = 0$. For comparison, a linear quadratic regulator whose state weighting matrix and input weighting matrix are identity matrices with appropriate dimension stabilizes the discretized system Σ_d in (5) only for $\Delta \in [-0.029, 0.062]$. From this numerical result, we see that the derived controller achieves better robust performance against clock offsets than a linear quadratic regulator designed without regard to the clock offset and also than the robust controller based on (34).

7 Concluding Remarks

We studied the problem of stabilizing systems in which the sensor and the controller have a constant clock offset. We formulated the problem as the stabilization problem for systems with parametric uncertainty. For multi-input systems, we derived a sufficient condition that is numerically testable, based on the results of simultaneous stabilization. For first-order systems, we obtained the maximum offset length that can be allowed by an LTI controller. However, a full investigation of the problem for general-order systems and systems with model uncertainty is still an open area for future research.

Fig. 3. Allowable offset length $\bar{\Delta} - \underline{\Delta}$ versus control update period h .

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