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Article:

Selivanov, A. orcid.org/0000-0001-5075-7229 and Fridman, E. (2019) Boundary observers for a reaction–diffusion system under time-delayed and sampled-data measurements. IEEE Transactions on Automatic Control, 64 (8). pp. 3385-3390. ISSN 0018-9286

https://doi.org/10.1109/tac.2018.2877381

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Boundary observers for a reaction-diffusion system under time-delayed and sampled-data measurements

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Abstract—We construct finite-dimensional observers for a 1D reaction-diffusion system with boundary measurements subject to time-delays and data sampling. The system has a finite number of unstable modes approximated by a Luenberger-type observer. The remaining modes vanish exponentially. For a given reaction coefficient, we show how many modes one should use to achieve a desired rate of convergence. The finite-dimensional part is analyzed using appropriate Lyapunov–Krasovskii functionals that lead to LMI-based convergence conditions feasible for small enough time-delay and sampling period. The LMIs can be used to find appropriate injection gains.

I. INTRODUCTION

Time-delays and data sampling are inevitable in practice due to finite speed of signal processing/transmission and digital nature of most controllers. Since the delay may lead to instability in the reaction-diffusion systems (see the examples in [1] and in Section IV below), these phenomena should be carefully studied.

Reaction-diffusion systems with various types of *in-domain* measurements/actuators subject to time-delays and sampling have been considered in [1]–[3]. These papers proposed observers/controllers that work if the delay, sampling period, and the distances between adjacent sensors/actuators are small enough. That is, the system should have enough high-frequency sensors/actuators.

The case of only one *boundary* sensor/actuator is more difficult to study. For diffusion-reaction systems, boundary controllers can be constructed using the backstepping approach [4], [5] or modal decomposition technique [6]–[9]. It has been shown in [10] that both approaches are robust to data sampling. In [11], modal decomposition technique was combined with a predictor to compensate a constant delay in the boundary controller. Robustness to small delays of general linear PDEs was studied in [12].

In this paper, we construct finite-dimensional observers for a 1D reaction-diffusion system with boundary measurements subject to time-delays and data sampling. Due to diffusion, there is a finite number of unstable modes, which we approximate by a Luenberger-type observer. The remaining modes vanish exponentially. For a given reaction coefficient, we show how many modes one should use to achieve a desired rate of convergence. Similar constructions have been proposed in [13], where a "lifting" technique and singular perturbation theory were used to obtain qualitative results. To obtain quantitative conditions, we use Lyapunov-Krasovskii functionals that lead to LMIs, which are feasible for small enough delay and sampling period and allow to find admissible upper bounds of these quantities.

Lemma 1 (Cauchy-Schwarz inequality): For $f \in L^2(0,1)$,

$$\left(\int_{0}^{1} f(x) \, dx\right)^{2} \le \int_{0}^{1} \left(f(x)\right)^{2} \, dx.$$
 (1)

Lemma 2 (Wirtinger inequality [14]): If $f \in \mathcal{H}^1(a, b)$ is such that f(a) = 0 or f(b) = 0 then

$$\|f\|_{L^2} \le \frac{2(b-a)}{\pi} \|f'\|_{L^2}.$$
(2)

II. TIME-DELAYED BOUNDARY MEASUREMENTS

Consider the reaction-diffusion system

$$z_t(x,t) = z_{xx}(x,t) + az(x,t),$$
 (3a)

$$z_x(0,t) = z(1,t) = 0,$$
 (3b)

$$z(x,0) = z_0(x) \tag{3c}$$

with the state $z: [0,1] \times [0,\infty) \to \mathbb{R}$, reaction coefficient $a \in \mathbb{R}$, and initial function $z_0: [0,1] \to \mathbb{R}$.

In this section, we construct an observer for the system (3) under the time-delayed boundary measurements

$$y(t) = \begin{cases} z(0, t - \tau(t)), & t - \tau(t) \ge 0, \\ 0, & t - \tau(t) < 0, \end{cases}$$
(4)

where $\tau(t) \in [\tau_m, \tau_M] \subset (0, \infty)$ is a known delay such that

$$\exists t_* \in [\tau_m, \tau_M]: \qquad \begin{cases} t - \tau(t) \ge 0, \quad t \ge t_*, \\ t - \tau(t) < 0, \quad t < t_*. \end{cases}$$
(5)

The condition $0 < \tau_m \le \tau(t)$ allows to use the step method for the well-posedness analysis (see Lemma 3). We perform robustness analysis with respect to the time delay, that is, the observer will converge to the system state for any $\tau(t) \le \tau_M$ with a small enough τ_M . Following [15], we require (5) to simplify the analysis on the interval where $t - \tau(t) < 0$.

Remark 1: The results of this paper can be extended to a more general system

$$\frac{\partial z}{\partial t}(x,t) = \frac{\partial}{\partial x} \left(p(x) \frac{\partial}{\partial x} z(x,t) \right) + q(x) z(x,t),$$

$$a_1 z(0,t) + a_2 z_x(0,t) = 0,$$

$$b_1 z(1,t) + b_2 z_x(1,t) = 0,$$
(6)

where $p \in C^1([0, 1]; (0, \infty))$, $q \in C([0, 1]; \mathbb{R})$, $a_2 \neq 0$, $|b_1| + |b_2| \neq 0$. We consider the simplified system (3) to avoid some technical details.

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A strong solution of (3) is a function

$$z \in L^{2}((0,\infty); \mathcal{H}^{2}(0,1)) \cap C([0,\infty); \mathcal{H}^{1}(0,1)),$$

$$z_{t} \in L^{2}((0,\infty); L^{2}(0,1))$$
(7)

that satisfies (3c) for t = 0 and (3a), (3b) for almost all t > 0. By [16, Theorem 7.7], (3) has a unique strong solution for

$$z_0 \in \mathcal{H}^1(0,1)$$
 s.t. $z_0(1) = 0.$ (8)

To construct a finite-dimensional observer, note that (3) has a finite number of unstable modes, while the remaining modes converge to zero. Namely, the system (3) can be presented as

$$\frac{dz}{dt} + \mathcal{A}z = 0, \quad z(0) = z_0, \tag{9}$$

where $z \colon [0,\infty) \to L^2(0,1)$ and

$$\mathcal{A}: D(\mathcal{A}) \subset L^2(0,1) \to L^2(0,1),$$

$$\mathcal{A}w = -w'' - aw$$
(10)

is a symmetric operator with the domain

$$D(\mathcal{A}) = \{ w \in \mathcal{H}^2(0,1) \, | \, w'(0) = w(1) = 0 \}$$
(11)

dense in $L^2(0,1)$. The eigenfunctions of \mathcal{A} , given by

$$\phi_n(x) = \sqrt{2} \cos\left(x\sqrt{\lambda_n} + a\right), \qquad n \in \mathbb{N}, \qquad (12)$$
$$\lambda_n = \frac{(2n-1)^2 \pi^2}{4} - a,$$

form an orthonormal basis in $L^2(0,1)$ [16, Corollary 3.26]. Thus, the solution of (3) can be presented as

$$z(\cdot,t) = \sum_{n=1}^{\infty} z_n(t)\phi_n(\cdot)$$
(13)

with $z_n(t) = \langle z(\cdot, t), \phi_n \rangle$. Using the symmetry of \mathcal{A} ,

$$\dot{z}_n(t) = \langle z_t(\cdot, t), \phi_n \rangle \stackrel{(9)}{=} - \langle \mathcal{A}z(\cdot, t), \phi_n \rangle = - \langle z(\cdot, t), \mathcal{A}\phi_n \rangle = -\lambda_n \langle z(\cdot, t), \phi_n \rangle = -\lambda_n z_n(t).$$
(14)

That is,

$$\dot{z}_n(t) = -\lambda_n z_n(t), \quad n \in \mathbb{N}.$$
 (15)

Let $\delta > 0$ be a desired decay rate of the observer estimation error. Since $\lim_{n\to\infty} \lambda_n = +\infty$, there exists $N \in \mathbb{N}$ such that

$$-\lambda_n \le -\delta, \quad \forall n > N.$$
 (16)

We will show that (16) implies the exponential convergence of $\sum_{n>N} z_n(t)\phi_n(\cdot)$ with the decay rate δ . Thus, it can be approximated by zero. The term $\sum_{n=1}^N z_n(t)\phi_n(\cdot)$ is approximated using the Luenberger-type observer

$$\hat{z}(x,t) = \sum_{n=1}^{N} \hat{z}_n(t)\phi_n(x),$$
(17a)

$$\frac{d}{dt}\hat{z}_{n}(t) = -\lambda_{n}\hat{z}_{n}(t) - l_{n}[\hat{z}(0, t - \tau(t)) - y(t)], \quad (17b)$$

$$\hat{z}_n(t) = 0, \quad t \le 0, \quad n = 1, \dots, N$$
 (17c)

with the injection gains $l_1, \ldots, l_N \in \mathbb{R}$.

Remark 2: Our results can be easily extended to arbitrary initial conditions $\hat{z}_n(t) = z_n^0$, n = 1, ..., N. We consider (17c) to avoid some technical details.

Introduce the estimation error

$$e(x,t) = \hat{z}(x,t) - z(x,t).$$
(18)

If $e(\cdot, t) \in L^2(0, 1)$, it can be presented as

$$e(\cdot,t) = \sum_{n=1}^{\infty} e_n(t)\phi_n(\cdot), \tag{19}$$

where, in view of (13) and (17a),

$$e_n(t) = \hat{z}_n(t) - z_n(t), \quad n \le N,$$
 (20a)

$$e_n(t) = -z_n(t), \qquad n > N.$$
(20b)

In view of (15) and (17b), relation (20a) implies

 \mathbf{T}

$$\dot{e}_n(t) = -\lambda_n e_n(t) - l_n e(0, t - \tau(t)), \quad n \le N,$$
 (21)

which can be presented as

$$\dot{\bar{e}}(t) = A\bar{e}(t) - LC\bar{e}(t-\tau(t)) + L\zeta(t-\tau(t))$$
(22)

$$e = (e_1, \dots, e_N)^T,$$

$$A = \text{diag}\{-\lambda_1, \dots, -\lambda_N\},$$

$$L = (l_1, \dots, l_N)^T,$$

$$C = (\phi_1(0), \dots, \phi_N(0)) = (\sqrt{2}, \dots, \sqrt{2}),$$

$$\zeta(t) = \sum_{n=1}^N e_n(t)\phi_n(0) - e(0, t).$$
(23)

Since $\lambda_1, \ldots, \lambda_N$ are different, the pair (A, C) is observable. Therefore, we can choose $L = (l_1, \ldots, l_N)^T \in \mathbb{R}^N$ such that

$$\exists P > 0: \quad P(A - LC) + (A - LC)^T P < -2\delta P. \quad (24)$$

If $\tau(t) \equiv 0$, then (24) guarantees ISS of (22) with respect to $\zeta(t)$, which decays exponentially (we show this below). Thus, (22) is exponentially stable for $\tau(t) \equiv 0$ and remains so for $\tau(t) \leq \tau_M$ with a small enough τ_M . The next theorem allows to find admissible τ_M .

Theorem 1: Consider the system (3) with the measurements (4) subject to (5) and the boundary observer (17) with λ_n , ϕ_n from (12), N satisfying (16) with an arbitrary decay rate $\delta > 0$, and $L = (l_1, \ldots, l_N)^T \in \mathbb{R}^N$. Let there exist matrices $P_2, P_3, G \in \mathbb{R}^{N \times N}$ and positive-definite matrices $P, S, R \in \mathbb{R}^{N \times N}$ such that¹

$$\Phi < 0 \quad \text{and} \quad \left[\begin{smallmatrix} R & G \\ G^T & R \end{smallmatrix} \right] \ge 0,$$
 (25)

where $\Phi = {\Phi_{ij}}$ is the symmetric matrix composed from

$$\Phi_{11} = A^T P_2 + P_2^T A + 2\delta P + S - e^{-2\delta\tau_M} R,
\Phi_{12} = P - P_2^T + A^T P_3, \quad \Phi_{13} = e^{-2\delta\tau_M} (R - G) - P_2^T LC,
\Phi_{14} = e^{-2\delta\tau_M} G, \quad \Phi_{22} = -P_3 - P_3^T + \tau_M^2 R,
\Phi_{23} = -P_3^T LC, \quad \Phi_{24} = 0, \quad \Phi_{33} = -e^{-2\delta\tau_M} (2R - G - G^T),
\Phi_{34} = e^{-2\delta\tau_M} (R - G), \quad \Phi_{44} = -e^{-2\delta\tau_M} (S + R)$$
(26)

with A and C from (23). Then there exists M > 0 such that

$$\|\hat{z}(\cdot,t) - z(\cdot,t)\|_{L^2} \le M e^{-\delta t} \|z_0\|_{\mathcal{H}^1}, \quad t \ge 0$$
(27)

for any initial function z_0 from (8).

Proof: Since ϕ_n and λ_n defined in (12) are eigenfunctions and eigenvalues of the operator \mathcal{A} defined in (10),

$$\hat{z}_{t}(x,t) \stackrel{(17a)}{=} \sum_{n=1}^{N} \frac{d}{dt} \hat{z}_{n}(t) \phi_{n}(x) \\
\stackrel{(17b)}{=} -\sum_{n=1}^{N} \lambda_{n} \hat{z}_{n}(t) \phi_{n}(x) \\
-\sum_{n=1}^{N} l_{n} [\hat{z}(0,t-\tau(t)) - z(0,t-\tau(t))] \phi_{n}(x) \\
= -\sum_{n=1}^{N} \hat{z}_{n}(t) \mathcal{A} \phi_{n} \\
-\sum_{n=1}^{N} l_{n} [\hat{z}(0,t-\tau(t)) - z(0,t-\tau(t))] \phi_{n}(x) \\
\stackrel{(10)}{=} \hat{z}_{xx}(x,t) + a\hat{z}(x,t) \\
-l(x) [\hat{z}(0,t-\tau(t)) - z(0,t-\tau(t))],$$
(28)

¹MATLAB codes for solving the LMIs are available at https://github.com/AntonSelivanov/TAC18a

where $l(x) = \sum_{n=1}^{N} l_n \phi_n(x)$. The latter, (3), and (18) imply

$$e_t(x,t) = e_{xx}(x,t) + ae(x,t) - l(x)e(0,t-\tau(t)),$$
 (29a)

$$e_x(0,t) = e(1,t) = 0,$$
 (29b)

$$e(\cdot, 0) = -z_0, \quad e(\cdot, t) = 0, \quad t < 0.$$
 (29c)

Lemma 3: There exists a unique strong solution of (29) for any initial function z_0 satisfying (8).

Proof is given in Appendix A.

The strong solution $e(\cdot, t)$ of (29) can be presented as the series (19) and, by Parseval's identity,

$$||e(\cdot,t)||_{L^2}^2 = \sum_{n=1}^N e_n^2(t) + \sum_{n>N} e_n^2(t).$$
 (30)

The second term can be bounded as

$$\sum_{n>N} e_n^2(t) \stackrel{(20b)}{=} \sum_{n>N} z_n^2(t) \stackrel{(15)}{=} \sum_{n>N} e^{-2\lambda_n t} z_n^2(0)$$

$$\stackrel{(16)}{\leq} e^{-2\delta t} \sum_{n>N} z_n^2(0) \leq e^{-2\delta t} \|z(\cdot,0)\|_{L^2}^2$$

$$\stackrel{(29c)}{=} e^{-2\delta t} \|e(\cdot,0)\|_{L^2}^2 \leq e^{-2\delta t} \frac{4}{\pi^2} \|e_x(\cdot,0)\|_{L^2}^2.$$
(31)

To bound the first summand of (30), i.e., the state of (22), we first show that $\zeta(t)$ exponentially converges to zero. Since $\phi_n(1) = e(1,t) = 0$ and $\|\phi'_n\|_{L^2}^2 = \lambda_n + a$, we have

$$\begin{aligned} \zeta^{2}(t) &= \left(\sum_{n=1}^{N} e_{n}(t)\phi_{n}(0) - e(0,t)\right)^{2} \\ &= \left(\int_{0}^{1} \left(\sum_{n=1}^{N} e_{n}(t)\phi_{n}'(x) - e_{x}(x,t)\right) dx\right)^{2} \\ &\leq \left\|\sum_{n=1}^{N} e_{n}(t)\phi_{n}'(\cdot) - e_{x}(\cdot,t)\right\|_{L^{2}}^{2} \\ &= \left\|\sum_{n>N} e_{n}(t)\phi_{n}'\right\|_{L^{2}}^{2} = \sum_{n>N}(\lambda_{n} + a)e_{n}^{2}(t) \\ &\leq e^{-2\delta t}\sum_{n=1}^{\infty}(\lambda_{n} + a)e_{n}^{2}(0) = e^{-2\delta t}\|e_{x}(\cdot,0)\|_{L^{2}}^{2}. \end{aligned}$$
(32)

The last inequality is obtained in a manner similar to (31). Consequently,

$$\begin{aligned} \zeta^{2}(t-\tau(t)) &\leq e^{-2\delta(t-\tau(t))} \|e_{x}(\cdot,0)\|_{L^{2}}^{2} \\ &\leq e^{2\delta\tau_{M}} e^{-2\delta t} \|e_{x}(\cdot,0)\|_{L^{2}}^{2}. \end{aligned}$$
(33)

Consider the functional $V_{\tau} = V_0 + V_S + V_R$ with

$$V_{0} = \bar{e}^{T}(t)P\bar{e}(t),$$

$$V_{S} = \int_{t-\tau_{M}}^{t} e^{-2\delta(t-s)}\bar{e}^{T}(s)S\bar{e}(s) ds,$$

$$V_{R} = \tau_{M} \int_{-\tau_{M}}^{0} \int_{t+\theta}^{t} e^{-2\delta(t-s)}\dot{\bar{e}}^{T}(s)R\dot{\bar{e}}(s) ds d\theta.$$
(34)

We consider $V_{\tau}(t)$ on $[t_*, \infty)$ with t_* from (5). On this interval, (22) does not depend on $\bar{e}(t)$ with t < 0. Thus, we formally set $\bar{e}(t) = \bar{e}(0)$ for t < 0 to define V_{τ} on $[t_*, \tau_M)$ (see [15]). We have

$$\dot{V}_0 + 2\delta V_0 = 2\bar{e}^T P \dot{\bar{e}} + 2\delta \bar{e}^T P \bar{e},$$

$$\dot{V}_S + 2\delta V_S = \bar{e}^T S \bar{e} - e^{-2\delta\tau_M} \bar{e}^T (t - \tau_M) S \bar{e} (t - \tau_M),$$

$$\dot{V}_R + 2\delta V_R = \tau_M^2 \dot{\bar{e}}^T R \dot{\bar{e}} - \tau_M \int_{t - \tau_M}^t e^{-2\delta(t - s)} \dot{\bar{e}}^T (s) R \dot{\bar{e}} (s) \, ds.$$
(35)

Using Jensen's inequality [17, Proposition B.8] and reciprocally convex approach [18, Theorem 1], we have

$$\begin{aligned} &-\tau_{M}\int_{t-\tau_{M}}^{t}e^{-2\delta(t-s)}\dot{e}^{T}(s)R\dot{e}(s)\,ds \leq -\tau_{M}e^{-2\delta\tau_{M}}\times\\ &\left[\int_{t-\tau(t)}^{t}\dot{e}^{T}(s)R\dot{e}(s)\,ds + \int_{t-\tau_{M}}^{t-\tau(t)}\dot{e}^{T}(s)R\dot{e}(s)\,ds\right]\\ \leq &-e^{-2\delta\tau_{M}}\frac{\tau_{M}}{\tau(t)}\left[\int_{t-\tau(t)}^{t}\dot{e}(s)\,ds\right]^{T}R\left[\int_{t-\tau(t)}^{t}\dot{e}(s)\,ds\right]\\ &-e^{-2\delta\tau_{M}}\frac{\tau_{M}}{\tau_{M}-\tau(t)}\left[\int_{t-\tau_{M}}^{t-\tau(t)}\dot{e}(s)\,ds\right]^{T}R\left[\int_{t-\tau_{M}}^{t-\tau(t)}\dot{e}(s)\,ds\right]\\ \leq &-e^{-2\delta\tau_{M}}\left[\frac{\bar{e}(t)-\bar{e}(t-\tau(t))}{\bar{e}(t-\tau(t))-\bar{e}(t-\tau_{M})}\right]^{T}\left[\begin{smallmatrix} R\\ G^{T} R\\ R\end{smallmatrix}\right]\left[\frac{\bar{e}(t)-\bar{e}(t-\tau(t))}{\bar{e}(t-\tau(t))-\bar{e}(t-\tau_{M})}\right].\end{aligned}$$
(36)

Similarly to [19], we use the descriptor representation of (22)

$$0 = 2[\bar{e}^T P_2^T + \dot{\bar{e}}^T P_3^T][-\dot{\bar{e}} + A\bar{e} - LC\bar{e}(t - \tau(t)) + L\zeta(t - \tau(t))].$$
(37)

Summing up (35) and (37), for $\gamma > 0$ we obtain

$$\dot{V}_{\tau}(t) + 2\delta V_{\tau}(t) - \gamma \zeta^2 (t - \tau(t)) \le \psi^T(t) \Psi \psi(t), \quad (38)$$

where $\psi = \operatorname{col}\{\bar{e}(t), \bar{e}(t), \bar{e}(t-\tau(t)), \bar{e}(t-\tau_M), \zeta(t-\tau(t))\},\$

$$\Psi = \begin{bmatrix} \Phi & P_2^T L \\ \Phi & P_3^T L \\ 0_{2N \times 1} \\ \bar{L}^T \bar{P}_2 \ \bar{L}^T \bar{P}_3 \ \bar{0}_{1 \times 2N} \\ \bar{-\gamma} \end{bmatrix}$$
(39)

Since $\Phi < 0$, the inequality $\Psi < 0$ holds for a large enough $\gamma \in \mathbb{R}$. Moreover, $\Phi < 0$ holds with δ replaced by $\delta + \epsilon$ if $\epsilon > 0$ is small enough. Thus,

$$\dot{V}_{\tau}(t) \leq -2(\delta+\epsilon)V_{\tau}(t) + \gamma\zeta^{2}(t-\tau(t)) \\
\leq -2(\delta+\epsilon)V_{\tau}(t) + \gamma e^{2\delta\tau_{M}}e^{-2\delta t} \|e_{x}(\cdot,0)\|_{L^{2}}^{2}.$$
(40)

The comparison principle implies:

$$V_{\tau}(t) \le e^{-2\delta(t-t_*)}V_{\tau}(t_*) + \frac{\gamma e^{2\delta\tau_M}}{2\epsilon} e^{-2\delta t} \|e_x(\cdot,0)\|_{L^2}^2.$$
(41)

Due to (5), $\dot{\bar{e}}(t) = A\bar{e}(t)$ for $t \in [0, t_*)$, thus, $|\bar{e}(t)| \le e^{\kappa t} |\bar{e}(0)|$ for $t \in [0, t_*)$ with some $\kappa > 0$. Therefore, for some C > 0,

$$V_{\tau}(t_{*}) \leq C \max_{t \in [t_{*} - \tau_{M}, t_{*}]} |\bar{e}(t)|^{2} \leq C e^{2\kappa t_{*}} |\bar{e}(0)|^{2} \leq C e^{2\kappa t_{*}} \sum_{n=1}^{\infty} e_{n}^{2}(0) \\ = C e^{2\kappa t_{*}} \|e(\cdot, 0)\|_{L^{2}}^{2} \leq C e^{2\kappa t_{*}} \frac{4}{\pi^{2}} \|e_{x}(\cdot, 0)\|_{L^{2}}^{2}.$$
(42)

The latter and (41) imply

$$\sum_{n=1}^{N} e_n^2(t) \le \lambda_{\min}^{-1}(P) V_{\tau}(t) \le M_1 e^{-2\delta t} \|e_x(\cdot, 0)\|_{L^2}^2$$
(43)

with some $M_1 > 0$. Finally, we have

$$\begin{aligned} \|\hat{z}(\cdot,t) - z(\cdot,t)\|_{L^{2}}^{2} &= \|e(\cdot,t)\|_{L^{2}}^{2} \\ &= \sum_{n=1}^{N} e_{n}^{2}(t) + \sum_{n=N+1}^{\infty} e_{n}^{2}(t) \overset{(43),(31)}{\leq} M^{2} e^{-2\delta t} \|e_{x}(\cdot,0)\|_{L^{2}}^{2} \end{aligned}$$
(44)

with some M > 0. Thus, (27) is true.

Remark 3: We have to use the \mathcal{H}^1 -norm in the right-hand side of (27), since the L^2 -norm does not take into account the point values that we use as measurements (4). Namely, we cannot bound ζ without using the space derivative as in (33).

Corollary 1: The observer (17) with $L = (l_1, \ldots, l_N)^T$ satisfying (24) converges to (3) with the decay rate δ in the sense of (27) if the delay bound τ_M is small enough.

Proof: Take P from (24), $P_2 = P$, $P_3 = \varepsilon I > 0$, $R = \mu^{-1}I > 0$, G = S = 0, and $\tau_M = 0$. Then

$$\Phi \stackrel{(26)}{=} \left[\frac{M_1 | M_2}{M_2^T | M_3} \right]$$

with

$$M_1 = \begin{bmatrix} A^T P + PA + 2\delta P - \mu^{-1}I \ \varepsilon A^T \\ * \ -2\varepsilon I \end{bmatrix},$$
$$M_2 = \begin{bmatrix} \mu^{-1}I - PLC \ 0 \\ -\varepsilon LC \ 0 \end{bmatrix}, \quad M_3 = \begin{bmatrix} -2\mu^{-1}I \ \mu^{-1}I \\ * \ -\mu^{-1}I \end{bmatrix}.$$

Clearly,

$$M_3 < 0$$
 and $M_3^{-1} = -\mu \begin{bmatrix} I & I \\ I & 2I \end{bmatrix}$.

By Schur's complement lemma, $\Phi < 0$ is equivalent to

$$M_{1} - M_{2}M_{3}^{-1}M_{2}^{T} = \begin{bmatrix} P(A - LC) + (A - LC)^{T}P + 2\delta P \ \varepsilon (A - LC)^{T} \\ \varepsilon (A - LC) & -2\varepsilon I \end{bmatrix} \\ + \mu \begin{bmatrix} PLC \\ \varepsilon LC \end{bmatrix} \begin{bmatrix} PLC \\ \varepsilon LC \end{bmatrix}^{T} < 0.$$
(45)

In view of (24), the later holds for small $\varepsilon > 0$ and $\mu > 0$. Thus, $\Phi < 0$ is feasible for $\tau_M = 0$. By continuity, it remains so for a small $\tau_M > 0$. Then Theorem 1 implies (27).

The well-posedness of (8), (29) with $\tau(t) \equiv 0$ can be proved using [20, Theorem 6.3.1]. Then Theorem 1 and Corollary 1 imply the following result.

Corollary 2: For $\tau(t) \equiv 0$, the observer (17) with $L = (l_1, \ldots, l_N)^T$ satisfying (24) exponentially converges to (3) with the decay rate δ in the sense of (27).

Remark 4: The LMIs of Theorem 1 allow to find appropriate injection gain $L = (l_1, \ldots, l_N)^T$. Following [21, Section 5.2], one can take $P_3 = \varepsilon P_2$, where ε is a tuning parameter, and use $Y = P_2^T L$ as a new decision variable. After solving the resulting LMIs, the injection gain can be found as $L = (P_2^T)^{-1}Y$.

III. SAMPLED-DATA BOUNDARY MEASUREMENTS

In this section, we construct an observer for the system (3) under the sampled in time boundary measurements

$$y(t) = z(0, t_k), \quad t \in [t_k, t_{k+1}), \quad k \in \mathbb{N},$$
 (46)

where $0 = t_1 < t_2 < t_3 < \cdots$ are sampling instants satisfying

$$0 < t_{k+1} - t_k \le h, \quad \lim_{k \to \infty} t_k = \infty.$$
(47)

Remark 5: The output (46) can be presented as (4) with

$$\tau(t) = t - t_k, \quad t \in [t_k, t_{k+1}), \quad k \in \mathbb{N}$$
(48)

such that $0 \le \tau(t) \le \tau_M = h$ and (5) is satisfied with $t_* = 0$. The condition $0 < \tau_m \le \tau(t)$ was imposed only to establish the well-posedness of (29) (see Lemma 3) and we will show that it is not required for the measurements (46). Therefore, the results of Theorem 1 can be applied. However, we will perform a more subtle analysis using the ideas of [22], which take into account the saw-tooth shape of $\tau(t)$ and lead to simpler convergence conditions. Similarly to (17), the boundary observer is constructed as

$$\hat{z}(x,t) = \sum_{n=1}^{N} \hat{z}_n(t)\phi_n(x),
\frac{d}{dt}\hat{z}_n(t) = -\lambda_n \hat{z}_n(t) - l_n[\hat{z}(0,t_k) - y(t)],
t \in [t_k, t_{k+1}), \quad k \in \mathbb{N},
\hat{z}_n(0) = 0, \quad n = 1, \dots, N.$$
(49)

Theorem 2: Consider the system (3) with the measurements (46) subject to (47) and the boundary observer (49) with λ_n , ϕ_n from (12), N satisfying (16) with an arbitrary decay rate $\delta > 0$, and $L = (l_1, \ldots, l_N)^T \in \mathbb{R}^N$. Let there exist matrices $P_2, P_3 \in \mathbb{R}^{N \times N}$ and positive-definite matrices $P, W \in \mathbb{R}^{N \times N}$ such that² $\Upsilon < 0$, where $\Upsilon = {\Upsilon_{ij}}$ is the symmetric matrix composed from

$$\begin{split} &\Upsilon_{11} = (A - LC)^T P_2 + P_2^T (A - LC) + 2\delta P, \\ &\Upsilon_{12} = P - P_2^T + (A - LC)^T P_3, \quad \Upsilon_{13} = -P_2^T LC, \\ &\Upsilon_{22} = -P_3 - P_3^T + h^2 e^{2\delta h} W, \quad \Upsilon_{23} = -P_3^T LC, \\ &\Upsilon_{33} = -\frac{\pi^2}{4} W \end{split}$$
(50)

with A and C from (23). Then there exists M > 0 such that (27) holds for any initial function z_0 from (8).

Proof: Similarly to (29), the estimation error $e(x,t) = \hat{z}(x,t) - z(x,t)$ satisfies

$$e_t(x,t) = e_{xx}(x,t) + ae(x,t) - l(x)e(0,t_k), t \in [t_k, t_{k+1}), k \in \mathbb{N}, e_x(0,t) = e(1,t) = 0, e(\cdot,0) = -z_0,$$
(51)

where $l(x) = \sum_{n=1}^{N} l_n \phi_n(x)$. Similarly to Lemma 3, the wellposedness of (8), (51) is established considering $f(x,t) = -l(x)e(0,t_k)$ as constant inhomogeneities on every step $[t_k, t_{k+1}), k \in \mathbb{N}$. Presenting *e* as (19), we obtain (cf. (22))

$$\dot{\bar{e}}(t) = (A - LC)\bar{e}(t) - LCv(t) + L\zeta(t_k), \ t \in [t_k, t_{k+1}), \ (52)$$

where $v(t) = \bar{e}(t_k) - \bar{e}(t)$ for $t \in [t_k, t_{k+1})$ and the other notations are from (23). Consider the functional $V_h = V_0 + V_W$ with $V_0 = \bar{e}^T(t)P\bar{e}(t)$ and

$$V_W = h^2 e^{2\delta h} \int_{t_k}^t e^{-2\delta(t-s)} \dot{e}^T(s) W \dot{\bar{e}}(s) ds$$

$$-\frac{\pi^2}{4} \int_{t_k}^t e^{-2\delta(t-s)} v^T(s) W v(s) ds, \quad t \in [t_k, t_{k+1}).$$
(53)

Note that $V_W \ge 0$ due to the exponential Wirtinger inequality [23, Lemma 1]. Moreover, V_h does not increase in the jumps at t_k and is continuous elsewhere. We have

$$V_{0} + 2\delta V_{0} = 2\bar{e}^{T}P\bar{e} + 2\delta\bar{e}^{T}P\bar{e},$$

$$\dot{V}_{W} + 2\delta V_{W} = h^{2}e^{2\delta h}\bar{e}^{T}(t)W\bar{e}(t) - \frac{\pi^{2}}{4}v^{T}(t)Wv(t),$$

$$0 = 2[\bar{e}^{T}P_{2}^{T} + \bar{e}^{T}P_{3}^{T}] \times$$

$$[-\bar{e} + (A - LC)\bar{e}(t) - LCv(t) + L\zeta(t_{k})], \quad t \in [t_{k}, t_{k+1}).$$

(54)

Summing up, we obtain

$$\dot{V}_h + 2\delta V_h - \gamma \zeta^2(t_k) = \xi^T \Xi \xi, \tag{55}$$

²MATLAB codes for solving the LMIs are available at https://github.com/AntonSelivanov/TAC18a

where $\xi = \operatorname{col}\{\bar{e}, \dot{\bar{e}}, v, \zeta(t_k)\}$ and

$$\Xi = \begin{bmatrix} & P_2^T L \\ \Upsilon & P_3^T L \\ & 0_{N \times 1} \\ \bar{L}^T \bar{P}_2 \bar{L}^T \bar{P}_3 \bar{0}_{1 \times N} & -\gamma \end{bmatrix}.$$
(56)

The rest of the proof is similar to that of Theorem 1.

Corollary 3: The observer (49) with $L = (l_1, \ldots, l_N)^T$ satisfying (24) converges to (3) with the decay rate δ in the sense of (27) if the sampling period h is small enough.

Proof: Take *P* from (24), $P_2 = P$, $P_3 = \varepsilon I > 0$, $W = \mu^{-1}I > 0$, and h = 0. Calculating the Schur complement, we find that $\Upsilon < 0$ is equivalent to (45), which, in view of (24), holds for small $\varepsilon > 0$ and $\mu > 0$. Thus, $\Upsilon < 0$ is feasible for h = 0 and, by continuity, remains so for a small $\tau_M > 0$. Then Theorem 2 implies (27).

Remark 6: The LMIs of Theorem 2 can be transformed to solve the design problem in a manner similar to Remark 4.

Remark 7: If the sampling is uniform, i.e., $t_k = kh$, the system (52) can be studied using the discretization [21, Section 7.1.1]. Combining it with the modal decomposition technique, one will obtain necessary and sufficient conditions for (3), (46), (49) to satisfy (27). The advantage of the Lyapunov-Krasovskii approach developed here is that it leads to simple conditions under variable sampling (47).

IV. EXAMPLE

Consider the system (3) with a = 25 and sampled in time boundary measurements (46) subject to (47). We consider an unstable plant since otherwise $\hat{z}(x,t) = 0$ is an exponentially converging estimate. Let $\delta = 1$ be the desired rate of convergence of the observation error. Since (16) holds with N = 2, the observer (49) with appropriate injection gains l_1 , l_2 provides exponentially converging state estimate for a small enough sampling period h. To find l_1 , l_2 , and h, we take small h and increase it while the design LMIs with $\varepsilon = 0.5$ (see Remarks 4 and 6) remain feasible. This gives

$$h = 0.048, \quad L = \begin{bmatrix} l_1 \\ l_2 \end{bmatrix} \approx \begin{bmatrix} 23.2 \\ -1.1 \end{bmatrix}.$$
 (57)

The analytical bound for the uniform sampling is $h \approx 0.081$, which we found using the method described in Remark 7. Note that we used the Lyapunov functional with the Wirtinger-based term (53) that leads to simple LMIs on the account of some conservatism. Less conservative conditions may be derived using other types of Lyapunov functionals (see, e.g., [24]).

The results of numerical simulations for the initial function

$$z_0(x) = \sin(2\pi x), \quad x \in [0, 1]$$
 (58)

are given in Figs. 1 and 2. For comparison, Fig. 2 also shows the error under the continuous measurements y(t) = z(0, t).

The observer (49) coincides with (17) for $\tau(t)$ defined in (48). Thus, it can be studied using Theorem 1 and Remark 4. In the considered example, these conditions lead to a smaller sampling period h = 0.031 with approximately the same injection gains l_1 , l_2 .

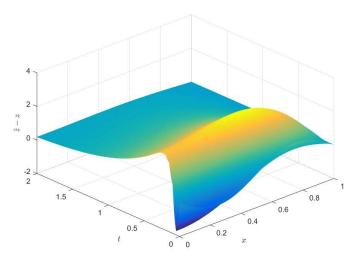


Fig. 1. Estimation error $\hat{z}(x,t) - z(x,t)$ of the observer (49) under the sampled-data measurements (46)

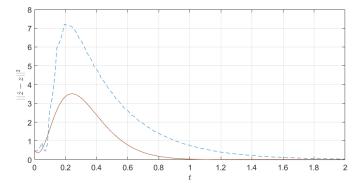


Fig. 2. Evolution of $\|\hat{z}(\cdot,t) - z(\cdot,t)\|_{L^2}^2$ for sampled-data (dashed blue line) and continuous-time (solid red line) measurements

V. CONCLUSION

We have designed finite-dimensional observers for a 1D reaction-diffusion system under delayed and sampled in time boundary measurements. We showed how to choose the observer injection gains and proved that it provides exponentially converging estimate if the time-delay or sampling period are small enough. The obtained LMIs allow to find admissible bounds on the delay and sampling period. The proposed observers can be used to design network-based controllers for parabolic systems. This may be a subject of the future research.

APPENDIX A Proof of Lemma 3

The proof is based on [16, Theorem 7.7] and the step method. Since $t - \tau(t) \leq 0$ for $t \in [0, \tau_m]$,

$$f(x,t) = -l(x)e(0,t-\tau(t)), \quad t \in [0,\tau_m]$$
(59)

can be viewed as inhomogeneity $f: [0, \tau_m] \to L^2(0, 1)$ and

$$\int_{0}^{\tau_{m}} \|f(s)\|_{L^{2}}^{2} ds \stackrel{(29c)}{\leq} \int_{0}^{\tau_{m}} \|l(\cdot)z_{0}(0)\|_{L^{2}}^{2} ds = \tau_{m}z_{0}^{2}(0)\|l\|_{L^{2}}^{2} < \infty.$$
(60)

Therefore, $f \in L^2((0, \tau_m); L^2(0, 1))$ and [16, Theorem 7.7] guarantees the existence of a unique strong solution $e \in C([0, \tau_m]; \mathcal{H}^1)$.

Since $t - \tau(t) \le \tau_m$ for $t \in [\tau_m, 2\tau_m]$,

$$f(x,t) = -l(x)e(0,t-\tau(t)), \quad t \in [\tau_m, 2\tau_m]$$
(61)

can be viewed as inhomogeneity $f: [\tau_m, 2\tau_m] \to L^2(0, 1)$. Since $e(\cdot, t)$ is continuous on $[0, \tau_m]$ in \mathcal{H}^1 , e(0, t) is also continuous on $[0, \tau_m]$:

$$e(0,t_1) - e(0,t_2)| = \left| \int_0^1 \left(e_x(y,t_1) - e_x(y,t_2) \right) \, dy \right|$$

$$\leq \| e_x(\cdot,t_1) - e_x(\cdot,t_2) \|_{L^2}.$$
(62)

Thus, there exists $M_e \in \mathbb{R}$ such that $\sup_{t \leq \tau_m} |e(0,t)| \leq M_e$. Clearly,

$$\int_{\tau_m}^{2\tau_m} \|f(s)\|_{L^2}^2 \, ds \le \tau_m M_e^2 \|l\|_{L^2}^2 < \infty. \tag{63}$$

Therefore, $f \in L^2((\tau_m, 2\tau_m); L^2(0, 1))$ and [16, Theorem 7.7] guarantees the existence of a unique strong solution $e \in C([\tau_m, 2\tau_m]; \mathcal{H}^1)$. Repeating the same reasoning consequently on every interval $[j\tau_m, (j+1)\tau_m]$ with $j = 2, 3, \ldots$, we obtain the existence of a unique strong solution on $[0, \infty)$.

REFERENCES

- E. Fridman and A. Blighovsky, "Robust sampled-data control of a class of semilinear parabolic systems," *Automatica*, vol. 48, no. 5, pp. 826– 836, 2012.
- [2] N. Bar Am and E. Fridman, "Network-based H_∞ filtering of parabolic systems," *Automatica*, vol. 50, no. 12, pp. 3139–3146, 2014.
- [3] A. Selivanov and E. Fridman, "Delayed point control of a reactiondiffusion PDE under discrete-time point measurements," *Automatica*, vol. 96, pp. 224–233, oct 2018.
- [4] A. Smyshlyaev and M. Krstic, "Closed-form boundary state feedbacks for a class of 1-D partial integro-differential equations," *IEEE Transactions on Automatic Control*, vol. 49, no. 12, pp. 2185–2202, 2004.
- [5] M. Krstic and A. Smyshlyaev, Boundary Control of PDEs: A Course on Backstepping Designs. SIAM, 2008.
- [6] D. L. Russell, "Controllability and stabilizability theory for linear partial differential equations: recent progress and open questions," *SIAM Review*, vol. 20, no. 4, pp. 639–739, 1978.
- [7] R. Triggiani, "Boundary Feedback Stabilizability of Parabolic Equations," *Applied Mathematics and Optimization*, vol. 6, no. 1, pp. 201– 220, 1980.
- [8] I. Lasiecka and R. Triggiani, "Stabilization and Structural Assignment of Dirichlet Boundary Feedback Parabolic Equations," *SIAM Journal on Control and Optimization*, vol. 21, no. 5, pp. 766–803, 1983.
- [9] J.-M. Coron and E. Trélat, "Global steady-state controllability of onedimensional semilinear heat equations," SIAM Journal on Control and Optimization, vol. 43, no. 2, pp. 549–569, 2004.
- [10] I. Karafyllis and M. Krstic, "Sampled-data boundary feedback control of 1-D parabolic PDEs," *Automatica*, vol. 87, pp. 226–237, 2018.
- [11] C. Prieur and E. Trélat, "Feedback stabilization of a 1D linear reactiondiffusion equation with delay boundary control," *IEEE Transactions on Automatic Control*, vol. PP, no. 1, p. 1, sep 2018.
- [12] H. Logemann, R. Rebarber, and G. Weiss, "Conditions for Robustness and Nonrobustness of the Stability of Feedback Systems with Respect to Small Delays in the Feedback Loop," *SIAM Journal on Control and Optimization*, vol. 34, no. 2, pp. 572–600, mar 1996.
- [13] M. B. Cheng, V. Radisavljevic, C. C. Chang, C.-F. Lin, and W.-C. Su, "A sampled-data singularly perturbed boundary control for a heat conduction system with noncollocated observation," *IEEE Transactions on Automatic Control*, vol. 54, no. 6, pp. 1305–1310, 2009.
- [14] G. Hardy, J. Littlewood, and G. Polya, *Inequalities*. Cambridge University Press, 1952.
- [15] K. Liu and E. Fridman, "Delay-dependent methods and the first delay interval," Systems & Control Letters, vol. 64, pp. 57–63, 2014.
- [16] J. C. Robinson, Infinite-dimensional dynamical systems: an introduction to dissipative parabolic PDEs and the theory of global attractors. Cambridge University Press, 2001.

- [17] K. Gu, V. L. Kharitonov, and J. Chen, Stability of Time-Delay Systems. Boston: Birkhäuser, 2003.
- [18] P. Park, J. W. Ko, and C. Jeong, "Reciprocally convex approach to stability of systems with time-varying delays," *Automatica*, vol. 47, no. 1, pp. 235–238, 2011.
- [19] E. Fridman, "New Lyapunov-Krasovskii functionals for stability of linear retarded and neutral type systems," *Systems & Control Letters*, vol. 43, pp. 309–319, 2001.
- [20] A. Pazy, Semigroups of Linear Operators and Applications to Partial Differential Equations. New York: Springer, 1983.
- [21] E. Fridman, Introduction to Time-Delay Systems: Analysis and Control. Birkhäuser Basel, 2014.
- [22] K. Liu and E. Fridman, "Wirtinger's inequality and Lyapunov-based sampled-data stabilization," *Automatica*, vol. 48, no. 1, pp. 102–108, 2012.
- [23] A. Selivanov and E. Fridman, "Observer-based input-to-state stabilization of networked control systems with large uncertain delays," *Automatica*, vol. 74, pp. 63–70, 2016.
- [24] E. Fridman, "A refined input delay approach to sampled-data control," *Automatica*, vol. 46, no. 2, pp. 421–427, 2010.