Sampled-data Control for a Class of Linear Hyperbolic System via the Lyapunov-Razumikhin Technique*

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***Abstract*— This work investigates the stability for a class of linear hyperbolic systems with distributed sampled-data controllers. First, we convert the original system into an equivalent system in which the sampling induced error is modeled as a reset integrator. Then by means of an appropriate Lyapunov function coupled with the Razumikhin technique, sufficient conditions are given for the Rε - stability of the system. Finally, our results are validated by a numerical example.**

I. INTRODUCTION

The application of digital computer in control system has become a general trend, which makes sampled-data control an active field of research in the past decades [1], [2], [21]. Stability and control design for finite-dimentional systems have been considered in many research works: see e.g. [16], [22], [29], [30]. Compared with the research method of finite dimensional system, the analysis and control of infinite dimensional systems is more challenging. Few results exist for sampled-data infinite dimensional system [25]–[27].

In general, sampled-data systems can be analyzed using discrete-time, time-delay and Input-Output methods (see [16] and references therein.) For the sake of partial differential equations (PDEs), using discrete-time finite-dimensional approximate models, [36] proposed a methodology for the design of sampled-data controller with practical stability guarantees. In references [14], [18], [34], [35], time-delay approach has been used for the analysis of parabolic PDEs. Hold boundary feedback control in one-dimensional linear hyperbolic systems were considered in [20]. In [10], [13], event-triggered sampled-data control with controller on the boundaries was developed. The boundary feedback control of a 2×2 hyperbolic system was implemented by backstepping method in [6], [12].

It can be seen from the literature review that the analysis of sampled-data controller for hyperbolic PDEs is a wide-open area of research, and there are still many topics worth studying. The present paper aims at studying the distributed sampled-control for a class of hyperbolic PDEs. The idea is to generalize the Input-Output approach ([15], [19], [31]) for finite dimensional systems, to the case of hyperbolic PDEs. An interconnected equivalent system consisting of a continuous-time PDE and a reset-integral operator is derived from the original system. In our previous work [11], the stability of linear hyperbolic systems with sampled-data controller has been ensured for sufficiently small sampling period. In the present paper, new stability condition which is based on Lyapunov-Razumikhin stability condition (e.g. [23], [24]) is proposed and the result of constructive local stability is obtained with the provided LMI method.

The paper is structured as follows: Section II presents the systems and the problem under study. In Section III, we propose the equivalent remodelling of system, followed by the concrete stability analysis process. A numerical example is given to illustrate the feasibility of our method in Section IV. The paper is ended with conclusions and perspectives.

Notations: $\mathbb{N}$ is the set of nonnegative integers from 0 to infinity, $\mathbb{R}_+$ is the set of nonnegative reals, $\mathbb{R}^n$ is used to denote the set of n-dimensional Euclidean space with the norm $|\cdot|$. $L^2(0,L)$ stands for the Hilbert space of square integrable scalar functions on $(0,L)$ with the corresponding norm $\|\cdot\|_{L^2(0,L)}$, defined by $\|\tau\|_{L^2(0,L)}=\sqrt{\int_0^L|\tau(x)|^2dx}$. The associated norm to Sobolev space $H^1(0,L)$ is defined as $\|\tau\|_{H^1(0,L)}=\sqrt{\int_0^L\left(|\tau(x)|^2+|\tau_x(x)|^2\right)dx}$. The set of functions $\varsigma:[0,L]\to\mathbb{R}^n$ such that $\int_0^L|\varsigma(x)|^2dx<\infty$ is denoted by $L^2([0,L];\mathbb{R}^n)$ and given a functional $V:H^1([0,L];\mathbb{R}^n)\to\mathbb{R}_+$ such that $\mathfrak{L}_{V\leq C}=\left\{y\in H^1([0,L];\mathbb{R}^n):V(y)\leq C\right\}$. The notation $W\leq 0$ denotes that W is symmetric and negative semidefinite. The symmetric elements are denoted by $*$ in the symmetric matrix. The identity matrix is denoted by I and $\lambda_{\min}(\Theta)$ and $\lambda_{\max}(\Theta)$ are the minimum and maximum eigenvalues of the matrix Θ. $\mathcal{C}^0$ is the continuous function and $\mathcal{C}^1$ is the continuously differentiable function. $\lceil\cdot\rceil$ is the ceiling function.

II. SYSTEM DESCRIPTION AND PROBLEM FORMULATION

A. System Description

We consider the following sampled-data controlled hyperbolic system (1)

$$\begin{cases}\partial_t z(t,x)+\Lambda\partial_x z(t,x)+\Gamma z(t,x)+u(t,x)=0, & \text{(1a)}\\ u(t,x)=F z(t_k,x),\forall t\in[t_k,t_{k+1}),k\in\mathbb{N}, & \text{(1b)}\\ z(t,0)=z(t,L)=0,\forall t\geq 0, & \text{(1c)}\\ z(0,x)=z_0(x),\forall x\in[0,L], & \text{(1d)}\end{cases}$$

where $z:[0,+\infty)\times[0,L]\to\mathbb{R}^n$, $\Lambda=diag\{\lambda_1,\lambda_2,...,\lambda_n\}$ with $\lambda_1,\lambda_2,...,\lambda_n>0$, Γ and F are real $n\times n$ constants

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matrices. The sampling instants are defined as a sequence $\{t_k\}_{k\in\mathbb{N}}$ where

$$t_0 = 0, t_{k+1} - t_k \in [\underline{h}, \bar{h}], \tag{2}$$

and $\bar{h} \geq \underline{h} > 0$.

To address the issue under consideration, we need the compatibility condition given below:

Condition 1. The initial condition $z_0(x)$ for $\forall x \in [0, L]$, satisfies

$$z_0(0) = z_0(L) = 0, \partial_x z_0(0) = \partial_x z_0(L) = 0. \tag{3}$$

Remark 1. Now we explain the concept of the solution in the current work. We rewrite the system (1)-(2) as a first order system

$$\begin{cases} \frac{dz(t)}{dt} = \Upsilon z(t) + f(z(t_k)), t \in [t_k, t_{k+1}), k \in \mathbb{N}, \\ z(0) = z_0, \end{cases}$$

where $f(z(t_k)) = -F z(t_k)$, and Υ is an operator defined by

$$\Upsilon z = -\Lambda \partial_x z(t,x) - \Gamma z(t,x), \tag{4}$$

with domain

$$\mathcal{D}(\Upsilon) = \left\{ z \in H^1(0, L; \mathbb{R}^n) \;\middle|\; \begin{array}{c} z(0) = z(L) = 0, \\ z_x(0) = z_x(L) = 0. \end{array} \right\} \tag{5}$$

A stable C_0 semigroup is produced by the operator Υ (see the proof of theorem A.1. in [3]). Moreover, the fact is that $f_k : H^1(0, L) \to H^1(0, L)$ is continuously differentiable for $t \in [t_k, t_{k+1})$. If $z_0 \in \mathcal{D}(\Upsilon)$, then in the light of Theorem 6.1.5 of [32], there is a classical solution for each $t \in [t_k, t_{k+1}), k \in \mathbb{N}$. Consequently, a solution can be constructed by selecting the last value of the previous sampling interval as the initial condition for the next sampling interval so that it is continuous at each sampling instant.

B. Problem Formulation

In the present work, we adopt the Rε-stability for the system (1)-(2), which is defined as follows.

Definition 1. Rε-stability [33]
Consider positive scalars R and ε, such that $0 < \varepsilon < R$, and a candidate Lyapunov function $V : H^1([0, L]; \mathbb{R}^n) \to \mathbb{R}_+$. If for all solutions of system (1) with $z_0(x) \in \mathfrak{L}_{V<R}$, the trajectory of the state $z(t, x)$ converges to $\mathfrak{L}_{V\leq\varepsilon}$ as t goes to infinity, then, system (1) is called Rε-stable from $\mathfrak{L}_{V<R}$ to $\mathfrak{L}_{V\leq\varepsilon}$.

Our main goal is to ensure that the closed-loop system (1)-(2) is Rε-stable due to Input-Output method.

III. MAIN RESULT

This section consists of two parts. Firstly, the sampled-data system is equivalently expressed as a continuous hyperbolic PDE with sampling induced error as disturbances in the input. Secondly, we provide constructive Rε-stability criteria based on the provided model.

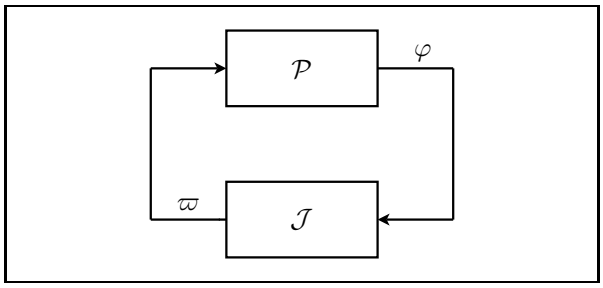

Fig. 1. Alternative representation of the closed-loop system.

A. System Remodelling

System (1) can be rewritten equivalently as

$$\begin{cases} \partial_t z(t,x) + \Lambda \partial_x z(t,x) + (\Gamma + F) z(t,x) \\ \qquad + F\varpi(t,x) = 0, \forall t \in [t_k, t_{k+1}), k \in \mathbb{N}, & \text{(6a)} \\ z(t,0) = z(t,L) = 0, \forall t \geq 0, & \text{(6b)} \\ z(0,x) = z_0(x), \forall x \in [0, L]. & \text{(6c)} \end{cases}$$

with the sampling induced error

$$\varpi(t,x) = z(t_k, x) - z(t,x). \tag{7}$$

Define the function φ as

$$\varphi(t,x) = \frac{\partial z(t,x)}{\partial t}, \forall t \geq 0, x \in [0, L]. \tag{8}$$

Note that

$$\begin{aligned} \varpi(t,x) = -\int_{t_k}^{t} \frac{\partial z(\theta,x)}{\partial \theta} d\theta = -\int_{t_k}^{t} \varphi(\theta,x)\, d\theta, \\ \forall t \in [t_k, t_{k+1}), k \in \mathbb{N}, x \in [0, L]. \end{aligned} \tag{9}$$

Therefore, the closed-loop system can be seen as the interconnection of two systems $\mathcal{P}$ and $\mathcal{J}$ shown in Figure 1, where the operator $\mathcal{P} : L^2(0, L) \to L^2(0, L)$ is defined by

$$\mathcal{P} : \begin{cases} \partial_t z(t,x) = -\Lambda \partial_x z(t,x) - (\Gamma + F) z(t,x) \\ \qquad - F\varpi(t,x), \\ z(t,0) = z(t,L) = 0, \forall t \geq 0, \\ z(0,x) = z_0(x), \forall x \in [0, L], \\ \varphi(t,x) = -\Lambda \partial_x z(t,x) - (F + \Gamma) z(t,x) \\ \qquad - F\varpi(t,x) = \partial_t z(t,x), \end{cases} \tag{10}$$

and the operator $\mathcal{J} : L^2(0, L) \to L^2(0, L)$ is defined by

$$\mathcal{J} : \begin{cases} \varpi(t,x) = (\mathcal{J}z)(t,x) = -\int_{t_k}^{t} \varphi(\theta,x)\, d\theta, \\ \forall t \in [t_k, t_{k+1}), k \in \mathbb{N}, x \in [0, L]. \end{cases} \tag{11}$$

Remark 2. The operator $\mathcal{P}$ is a nominal continuous-time control-loop, $\mathcal{J}$ is an integral operator. For the simplicity of the closed-loop system structure, we choose this form of $\mathcal{J}$ so that we need one output φ instead of two $z(t, x)$ and $z(t_k, x)$, which is why we wrote (7) as (9).

B. Stability Analysis

In the following, we present our primary results. Theorem 1 gives the conditions under which Rε-stability can be obtained.

Theorem 1. Consider systems (10)-(11) with (2) and a candidate Lyapunov function $V : H^1([0,L];\mathbb{R}^n) \to \mathbb{R}_+$ which is differentiable w.r.t. its argument and such that there exists $0 < \iota_1 < \iota_2$

$$\iota_1 \|\varrho\|^2_{H^1([0,L];\mathbb{R}^n)} \leq V(\varrho) \leq \iota_2 \|\varrho\|^2_{H^1([0,L];\mathbb{R}^n)}. \quad (12)$$

Suppose that along the trajectories of the system (10)-(11), the corresponding solution $z(t,\cdot)$ satisfies $\dot{V}(z) + 2\delta V(z) \leq 0$, for some $\delta > 0$, whenever

1) $R > V(z(t,\cdot)) \geq \max(\varepsilon, V(z(t_k,\cdot))/\alpha)$, with some $\alpha > 1$,
2) $z(t_k,\cdot) \in \mathfrak{L}_{V<R}$.

Then the system is Rε-stable from $\mathfrak{L}_{V<R}$ to $\mathfrak{L}_{V\leq\varepsilon}$.

Proof. During a sampling interval $[t_k, t_{k+1})$ with initial state $z(t_k,\cdot)$:

- If $V(z(t_k,\cdot)) \leq \varepsilon$, $V(z(t,\cdot))$ will remain in ε during $[t_k, t_{k+1})$. (Otherwise, we will have $\dot{V}(z) > 0 > -2\delta V(z)$ at some point when $V(z(t,\cdot)) \geq \varepsilon \geq V(z(t_k,\cdot))/\alpha$, which would contradict the proposition in Theorem 1.)
- If $R > V(z(t_k,\cdot)) \geq \varepsilon$:
  (a) We have $V(z(t,\cdot)) \leq V(z(t_k,\cdot))$ during $[t_k, t_{k+1})$. (Otherwise, we will have $\dot{V}(z) > 0 > -2\delta V(z)$ at some point when $V(z(t,\cdot)) \geq V(z(t_k,\cdot)) \geq V(z(t_k,\cdot))/\alpha$, which would contradict the proposition in Theorem 1.)
  (b)We can further show that

$$V(z(t,\cdot)) \leq \max(\varepsilon, \frac{V(z(t_k,\cdot))}{\alpha}, e^{-2\delta(t-t_k)}V(z(t_k,\cdot))) \quad (13)$$

  during $[t_k, t_{k+1})$. In the following, we will discuss two possibilities in case (b):
  (b1) If there exists $t' \in [t_k, t_{k+1})$ such that $V(z(t',\cdot)) = \max(\varepsilon, V(z(t_k,\cdot))/\alpha)$. If $t \in [t_k, t')$, $\dot{V}(z) + 2\delta V(z) \leq 0$ holds, and we have $V(z(t,\cdot)) \leq e^{-2\delta(t-t_k)}V(z(t_k,\cdot)), \forall t \in [t_k, t')$ (Otherwise, we will have $\dot{V}(z) > 0 > -2\delta V(z)$ at some point when $V(z(t,\cdot)) \geq V(z(t_k,\cdot))/\alpha$, which would contradict the proposition in Theorem 1.) If $t \in [t', t_{k+1})$, $V(z(t,\cdot))$ cannot go back above $\max(\varepsilon, V(z(t_k,\cdot))/\alpha)$ (Otherwise, we will have $\dot{V}(z) > 0 > -2\delta V(z)$ at some point when $V(z(t,\cdot)) \geq V(z(t_k,\cdot))/\alpha$, which would contradict the proposition in Theorem 1.), then we have $V(z(t,\cdot)) \leq \max(\varepsilon, V(z(t_k,\cdot))/\alpha), \forall t \in [t', t_{k+1})$. So, over the whole sampling interval $t \in [t_k, t_{k+1})$, we can get inequality (13).
  (b2) When $V(z(t,\cdot)) > \max\{\varepsilon, V(z(t_k,\cdot))/\alpha\}, \forall t \in [t_k, t_{k+1})$, since $\dot{V}(z) + 2\delta V(z) \leq 0$, we have $V(z(t,\cdot)) \leq e^{-2\delta(t-t_k)}V(z(t_k,\cdot))$. Then it is not hard to get (13).

Now, let us consider $z(t_k,\cdot) \in \mathfrak{L}_{V<R}$, $t \in [t_k, t_{k+1}), k \in \mathbb{N}$. We have

$$\begin{aligned} &V(z(t,\cdot)) \\ &\leq \max\left(\varepsilon, V(z(t_k,\cdot))/\alpha, e^{-2\delta(t-t_k)}V(z(t_k,\cdot))\right) \\ &= \max\left(\varepsilon, \xi V(z(t_k,\cdot))\right), \end{aligned} \quad (14)$$

with $\xi = \max\left(1/\alpha, e^{-2\delta(t-t_k)}\right) \leq 1$, then we can derive

$$V(z(t_k,\cdot)) \leq \max\left(\varepsilon, \zeta V(z(t_{k-1},\cdot))\right), \quad (15)$$

with $\zeta = \max\left(1/\alpha, e^{-2\delta\underline{h}}\right) < 1$, $\forall k \in \mathbb{N}\setminus\{0\}$, where $\underline{h}$ is the lower bound of the sampling interval.
By recursion, the following inequality holds if $z(t_0,\cdot) \in \mathfrak{L}_{V<R}$, we have

$$\begin{aligned} V(z(t_k,\cdot)) &\leq \max\left(\varepsilon, \zeta max\left(\varepsilon, \zeta V(z(t_{k-2},\cdot))\right)\right) \\ &\leq \max\left(\varepsilon, \zeta\varepsilon, \zeta^2 V(z(t_{k-2},\cdot))\right) \\ &\leq \max\left(\varepsilon, \zeta^2 V(z(t_{k-2},\cdot))\right) \\ &\leq \cdots \\ &\leq \max\left(\varepsilon, \zeta^k V(z(t_0,\cdot))\right), \forall k \in \mathbb{N}. \end{aligned} \quad (16)$$

Then combining (14) and (16), we get that

$$\begin{aligned} V(z(t,\cdot)) &\leq \max\left(\varepsilon, \xi\max\left(\varepsilon, \zeta^k V(z(t_0,\cdot))\right)\right) \\ &= \max\left(\varepsilon, \xi\varepsilon, \xi\zeta^k V(z(t_0,\cdot))\right) \\ &\leq \max\left(\varepsilon, \zeta^k V(z(t_0,\cdot))\right) \\ &= \varepsilon \end{aligned} \quad (17)$$

when k is large enough. Therefore, there $\exists\ \bar{k} = \lceil \log_\zeta {}^{\varepsilon}/_{V(z(t_0,\cdot))} \rceil$, such that $z(t,\cdot) \in \mathfrak{L}_{V\leq\varepsilon}, \forall\ t \geq t_{\bar{k}}$, which leads the proof of Rε-stability. ■

Remark 3. Theorem 2 is based on the generalization of the Razumikhin technique for hyperbolic system. In the following theorem, we will show how it can be used in constructed manner.

Theorem 2. Consider systems (10)-(11) with (2) and an initial condition satisfying (3):

1) Let $\underline{\lambda} = \min_{i\in\{1,\ldots,n\}} \lambda_i$. Assume that there exist $\mu, \gamma, \kappa > 0$, $\alpha > 1$ and symmetric positive matrices $\Theta_1 \in \mathbb{R}^{n\times n}$, $\Theta_2 \in \mathbb{R}^{n\times n}$ satisfying

$$W(0) + \kappa N(0) \preceq 0, \quad W(L) + \kappa N(L) \preceq 0, \quad (18)$$

with $W(x)$ defined for all $x \in [0, L]$ as

$$W(x) = \begin{bmatrix} G_1 & G_3 & 0 & 0 \\ * & -\gamma I & 0 & 0 \\ * & * & G_2 & G_4 \\ * & * & * & -\gamma I \end{bmatrix}, \quad (19)$$

where

$$\begin{aligned} G_1 &= -e^{-2\mu x}\left[(\digamma + \Gamma)^T\Theta_1 + \Theta_1(\digamma + \Gamma)\right], \\ G_2 &= -e^{-2\mu x}\left[\Gamma^T\Theta_2 + \Theta_2\Gamma + \beta\Theta_2\right], \\ G_3 &= -e^{-2\mu x}\Theta_1\digamma, \\ G_4 &= -e^{-2\mu x}\Theta_2\digamma. \end{aligned} \quad (20)$$

and

$$N(x) = e^{-2\mu x}\begin{bmatrix} G_5 & -\Theta_1 & 0 & 0 \\ * & -\Theta_1 & 0 & 0 \\ * & * & \alpha\Theta_2 & 0 \\ * & * & * & -\Theta_2 \end{bmatrix}, \quad (21)$$

where

$$G_5 = (\alpha - 1)\Theta_1. \quad (22)$$

2) If $\exists \varepsilon \in \mathbb{R}_+$, $R \in \mathbb{R}_+$ s.t. $0 < \varepsilon < R$ and

$$\begin{aligned} &\gamma 3\bar{h}\left(|\Lambda|^2\Omega_1 + \left(|\Upsilon|^2 + |\digamma|^2\right)\Omega_2\right) + \gamma\Omega_1 \\ &\leq (2\sigma - \beta)\varepsilon - 2\delta R, \end{aligned} \quad (23)$$

with

$$\Omega_1 = \frac{R}{\lambda_{\min}(\Theta_2)\, e^{-2\mu L}}, \Omega_2 = \frac{R}{\lambda_{\min}(\Theta_1)\, e^{-2\mu L}},$$

where $\sigma = \mu\underline{\lambda}$, $0 < \beta < 2\sigma$, $\delta > 0$.
Then the considered system (1) is Rε-stable from $\mathfrak{L}_{V<R}$ to $\mathfrak{L}_{V\leq\varepsilon}$ for any sampling sequence satisfying (2), with the Lyapunov function defined by

$$V(z) = V_1(z) + V_2(z), \quad (24)$$

with

$$V_1(z) = \int_0^L z^T e^{-2\mu x}\Theta_1 z dx, \quad (25)$$

$$V_2(z) = \int_0^L z_x^T e^{-2\mu x}\Theta_2 z_x dx. \quad (26)$$

Proof. Consider the Lyapunov function (24)-(26). It can be bounded as follows:

$$\begin{aligned} \Phi \left\|z(t,\cdot)\right\|^2_{H^1([0,L];\mathbb{R}^n)} &\leq V(z(t,\cdot)) \\ &\leq \Psi \left\|z(t,\cdot)\right\|^2_{H^1([0,L];\mathbb{R}^n)}, \end{aligned} \quad (27)$$

where

$$\begin{aligned} \Phi &= \min(\lambda_{\min}(\Theta_1), \lambda_{\min}(\Theta_2))e^{-2\mu L}, \\ \Psi &= \max(\lambda_{\max}(\Theta_1), \lambda_{\max}(\Theta_2)). \end{aligned}$$

Step 1: In this step we study the continuity of the function V defined in (24).

1) Since $z(t,x)$ is continuous with respect to t for all $t \in [t_k,\ t_{k+1}), k \in \mathbb{N}$, and continuous at sampling instants by construction (see Remark 1) , then V_1 is continuous for all $t \geq 0$.
2) From system (1), we can get

$$z_x(t,x) = \Lambda^{-1}\left(-z_t(t,x) - \Gamma z(t,x) - \digamma z(t_k,x)\right), \quad (28)$$

for all $t \in [t_k, t_{k+1}),\ k \in \mathbb{N}$. Since all the terms on the right of the equation (28) are continuous in t on (t_k, t_{k+1}), $\forall k \in \mathbb{N}$, then $z_x(t,x)$ and thus V_2 are also continuous in t for all $(t_k, t_{k+1}), k \in \mathbb{N}$.
Now we consider the time interval $[t_k, t_{k+1})$, for some $k \in \mathbb{N}$ and an initial condition $z_k(x)$. The solution of (1) is defined as $z(t,x)$ on the time interval $[t_k, t_{k+1})$, and is such that z and z_x are both $\mathcal{C}^0$ in $t \in [t_k, t_{k+1})$. Next, we prolong the solution to $\mathcal{C}^1$ in t on $[t_k, t_{k+1}]$. We denote $y(t,x)$ the solution on $[t_k, t_{k+1}]$ with initial condition $z_k(x)$. $y(t,x)$ and $y_x(t,x)$ are $\mathcal{C}^0$ in t on $[t_k, t_{k+1}]$. We got the following properties on $[t_k, t_{k+1})$

$$\begin{cases} z(t,x) = y(t,x), \\ z_x(t,x) = y_x(t,x). \end{cases} \quad (29)$$

Then the left limit can be calculated as

$$\lim_{t\to t_{k+1}^-} z_x(t,x) = \lim_{t\to t_{k+1}^-} y_x(t,x) = y_x(t_{k+1},x). \quad (30)$$

For the next time interval $[t_{k+1}, t_{k+2})$, we set the initial condition $z_{k+1}(x) = y(t_{k+1},x)$. Then the solution $z(t,x)$ of system (1) on $[t_{k+1}, t_{k+2})$ satisfies $z_x(t,x)$ is $\mathcal{C}^0$ in t on $[t_{k+1}, t_{k+2})$. Therefore, we have the right limit property

$$\lim_{t\to t_{k+1}^+} z_x(t,x) = z_x(t_{k+1},x) = y_x(t_{k+1},x). \quad (31)$$

According to (30) and (31), we can see that by construction, $z_x(t,x)$ is continuous in t at time instant t_{k+1}. Similarly, we can show that the function $z_x(t,x)$ is continuous at all sampling instants, which shows both the continuity of $z_x(t,x)$ with respect to time for all $t \geq 0$ and the continuity of V_2.

Remark 4. V_1 is used to bound z, and V_2 is used to deal with the term z_x that appears in the derivative of V_1.

Step 2: In this step we study the time derivative of $V(z)$ defined in (24). We first compute the time derivative of $V_1(z)$ along the solutions to (10)-(11),$\forall t \in [t_k, t_{k+1}), k \in \mathbb{N}$,

$$\begin{aligned} \dot{V}_1(z) &= \int_0^L \left(\partial_t z^T e^{-2\mu x}\Theta_1 z + z^T e^{-2\mu x}\Theta_1 \partial_t z\right)dx \\ &= \int_0^L \Big((-\Lambda\partial_x z - (\digamma + \Gamma)z - \digamma\varpi)^T e^{-2\mu x}\Theta_1 z \\ &\quad + z^T e^{-2\mu x}\Theta_1\left(-\Lambda\partial_x z - (\digamma + \Gamma)z - \digamma\varpi\right)\Big)dx \\ &= \int_0^L -\partial_x\left[z^T\Lambda e^{-2\mu x}\Theta_1 z\right]dx \\ &\quad + \int_0^L \Big(-z^T(\digamma + \Gamma)^T e^{-2\mu x}\Theta_1 z \\ &\quad - z^T e^{-2\mu x}\Theta_1(\digamma + \Gamma)z - z^T e^{-2\mu x}\Theta_1\digamma\varpi \\ &\quad - 2\mu z^T\Lambda e^{-2\mu x}\Theta_1 z - \varpi^T\digamma^T e^{-2\mu x}\Theta_1 z\Big)dx \\ &= -\left[z^T\Lambda e^{-2\mu x}\Theta_1 z\right]_0^L \\ &\quad + \int_0^L \Big(-z^T\Big((\digamma + \Gamma)^T e^{-2\mu x}\Theta_1 \\ &\quad + e^{-2\mu x}\Theta_1(\digamma + \Gamma)\Big)z \\ &\quad - \varpi^T\digamma^T e^{-2\mu x}\Theta_1 z - z^T e^{-2\mu x}\Theta_1\digamma\varpi\Big)dx \\ &\quad - 2\mu\int_0^L z^T\Lambda e^{-2\mu x}\Theta_1 z dx. \end{aligned} \quad (32)$$

In order to get the time derivative of z_x in V_2, we refer to the original system (1). Since $z : [0,+\infty)\times[0,L] \to \mathbb{R}^n$ has

consecutive partial derivatives in $[0,+\infty)\times[0,L]$, according to Schwartz's theorem ([17]) we can obtain

$$\begin{aligned}\partial_{xt}z\left(t,x\right)&=\partial_{tx}z\left(t,x\right)\\&=-\Lambda\partial_{xx}z\left(t,x\right)-\Gamma\partial_{x}z\left(t,x\right)-\digamma\partial_{x}z\left(t_{k},x\right),\\&\forall t\in\left(t_{k},t_{k+1}\right),k\in\mathbb{N}.\end{aligned}\tag{33}$$

For the next calculation of the time derivative of V_2, we use Lemma 1 in the appendix. According to (33) and Lemma 1, we have

$$\begin{cases}\partial_{x}z(t,0)=\partial_{x}z(t,L)=0,\forall t\in[t_{k},t_{k+1}),k\in\mathbb{N},&\text{(34a)}\\z_{0}(0)=z_{0}(L)=0,\quad\partial_{x}z_{0}(0)=\partial_{x}z_{0}(L)=0.&\text{(34b)}\end{cases}$$

Similarly to the computation of $\dot{V}_1$, the time derivative of $V_2(z)$ along the solutions to (33)-(34), $\forall t\in\left[t_{k},t_{k+1}\right),k\in\mathbb{N}$ is shown as follows

$$\begin{aligned}\dot{V}_{2}(z)=&-\left[\partial_{x}z^{T}\Lambda e^{-2\mu x}\Theta_{2}\partial_{x}z\right]_{0}^{L}\\&+\int_{0}^{L}\left(-\partial_{x}z^{T}\left(\Gamma^{T}e^{-2\mu x}\Theta_{2}+e^{-2\mu x}\Theta_{2}\Gamma\right)\partial_{x}z\right.\\&-\partial_{x}z^{T}\left(t_{k},\cdot\right)\digamma^{T}e^{-2\mu x}\Theta_{2}\partial_{x}z\\&\left.-\partial_{x}z^{T}e^{-2\mu x}\Theta_{2}\digamma\partial_{x}z^{T}\left(t_{k},\cdot\right)\right)dx\\&-2\mu\int_{0}^{L}\partial_{x}z^{T}\Lambda e^{-2\mu x}\Theta_{2}\partial_{x}zdx.\end{aligned}\tag{35}$$

Adding $\gamma\|\varpi\left(s,\cdot\right)\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}-\gamma\|\varpi\left(s,\cdot\right)\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}$ to (32) and

$$\begin{aligned}&\gamma\|\partial_{x}z\left(t_{k},\cdot\right)\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}-\gamma\|\partial_{x}z\left(t_{k},\cdot\right)\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2},\\&\beta\int_{0}^{L}z_{x}^{T}e^{-2\mu x}\Theta_{2}z_{x}dx-\beta\int_{0}^{L}z_{x}^{T}e^{-2\mu x}\Theta_{2}z_{x}dx\end{aligned}$$

to (35) for some $\gamma>0,\beta>0$, and using boundary condition (6b) and (34a) we have

$$\begin{aligned}\dot{V}(z)=&\dot{V}_{1}(z)+\dot{V}_{2}(z)\\\leq&\int_{0}^{L}\left(-z^{T}\mathcal{A}_{1}z-\varpi^{T}\digamma^{T}e^{-2\mu x}\Theta_{1}z\right.\\&\left.-z^{T}e^{-2\mu x}\Theta_{1}\digamma\varpi-\gamma\varpi^{T}\varpi\right)dx\\&-2\sigma V_{1}(z)+\gamma\left\|\varpi\left(s,\cdot\right)\right\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}\\&+\int_{0}^{L}\left(-\partial_{x}z^{T}\mathcal{A}_{2}\partial_{x}z\right.\\&-\partial_{x}z^{T}\left(t_{k},\cdot\right)^{T}e^{-2\mu x}\Theta_{2}\partial_{x}z\\&-\partial_{x}z^{T}e^{-2\mu x}\Theta_{2}\partial_{x}z^{T}\left(t_{k},\cdot\right)\\&\left.-\gamma\partial_{x}z^{T}\left(t_{k},\cdot\right)\partial_{x}z\left(t_{k},\cdot\right)\right)dx\\&-2\sigma V_{2}(z_{x})+\gamma\left\|\partial_{x}z\left(t_{k},\cdot\right)\right\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}\\&+\beta\int_{0}^{L}z_{x}^{T}e^{-2\mu x}\Theta_{2}z_{x}dx\\=&-2\sigma V_{1}(z)-2\sigma V_{2}(z)+\int_{0}^{L}\eta^{T}W(x)\eta dx\\&+\gamma\left\|\varpi\left(s,\cdot\right)\right\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}\\&+\gamma\left\|\partial_{x}z\left(t_{k},\cdot\right)\right\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}\\&+\beta\int_{0}^{L}z_{x}^{T}e^{-2\mu x}\Theta_{2}z_{x}dx\\=&-2\sigma V_{1}(z)-(2\sigma-\beta)V_{2}(z)\\&+\int_{0}^{L}\eta^{T}W(x)\eta dx+\gamma\left\|\varpi\left(s,\cdot\right)\right\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}\\&+\gamma\left\|\partial_{x}z\left(t_{k},\cdot\right)\right\|_{L^{2}\left(\left[0,L\right];\,\mathbb{R}^{n}\right)}^{2}.\end{aligned}\tag{36}$$

with

$$\begin{aligned}&\sigma=\mu\underline{\lambda},\\&\eta=[z^{T}\ \varpi^{T}\ \left(\partial_{x}z\right)^{T}\ \left(\partial_{x}z(t_{k},\cdot)\right)^{T}]^{T},\\&\mathcal{A}_{1}=-G_{1},\ \mathcal{A}_{2}=-G_{2}.\ (defined\ in\ (20)\ Theorem\ 2)\end{aligned}\tag{37}$$

and $W(x)$ defined in (19).

Step 3: In this step, we show that $\dot{V}(z)+2\delta V(z)\leq 0$, whenever

$$\begin{cases}R>V(z(t,\cdot))\geq\max(\varepsilon,V(z(t_{k},\cdot))/\alpha),&\text{(38a)}\\z(\theta,\cdot)\in\mathfrak{L}_{V<R},\forall\theta\in[t_{k},t),k\in\mathbb{N}.&\text{(38b)}\end{cases}$$

Let us assume that conditions (38) hold. Since condition (18) holds, by convexity, we have $W(x)+\kappa N(x)\leq 0$, for $x\in[0,L]$. Therefore, we get

$$\int_{0}^{L}\eta^{T}(W(x)+\kappa N(x))\eta dx\leq 0,\tag{39}$$

with $W(x)$ and $N(x)$ given in (19) and (21).

Now, consider $t\in[t_k,t_{k+1})$ and a trajectory z satisfying (38). Since condition (38a) is satisfied, we have $V(z(t,\cdot))\geq V(z(t_k,\cdot))/\alpha$ with some $\alpha>1$, which can be rewritten as

$$\int_{0}^{L}\eta^{T}N(x)\eta dx\geq 0.\tag{40}$$

In view of (39), (40) and $\kappa>0$, by S-procedure, it implies that

$$\int_{0}^{L}\eta^{T}W(x)\eta dx\leq 0.\tag{41}$$

According to condition (38b), we have

$$\begin{aligned}&\int_{0}^{L}z^{T}e^{-2\mu x}\Theta_{1}zdx<R,\forall z(\theta,\cdot)\in\mathfrak{L}_{V<R},\\&\int_{0}^{L}\partial_{x}z^{T}e^{-2\mu x}\Theta_{2}\partial_{x}zdx<R,\forall\partial_{x}z(\theta,\cdot)\in\mathfrak{L}_{V<R}.\end{aligned}$$

The following inequalities are further derived

$$\left\|z\left(\theta,\cdot\right)\right\|_{L^{2}\left(\left[0,L\right];\mathbb{R}^{n}\right)}^{2}<\frac{R}{\lambda_{\min}\left(\Theta_{1}\right)e^{-2\mu L}},\forall\theta\in[t_{k},t],\tag{42}$$

$$\left\|\partial_{x}z\left(\theta,\cdot\right)\right\|_{L^{2}\left(\left[0,L\right];\mathbb{R}^{n}\right)}^{2}<\frac{R}{\lambda_{\min}\left(\Theta_{2}\right)e^{-2\mu L}},\forall\theta\in[t_{k},t].\tag{43}$$

Using (9), (42) and (43) for $t \in [t_k, t_{k+1}), k \in \mathbb{N}$, the upper bound of $\|\varpi(s,\cdot)\|^2_{L^2([0,L];\mathbb{R}^n)}$ can be calculated

$$
\begin{aligned}
&\|\varpi(s,\cdot)\|^2_{L^2([0,L];\mathbb{R}^n)} \\
&= \int_0^L |\varpi(s,x)|^2 dx = \int_0^L \left| \int_{t_k}^t \frac{\partial z(\theta,x)}{\partial \theta} d\theta \right|^2 dx \\
&= \int_0^L \left| \int_{t_k}^t (\Lambda \partial_x z(\theta,x) + \Gamma z(\theta,x) + \digamma z(t_k,x)) d\theta \right|^2 dx \\
&\le 3\int_0^L \int_{t_k}^t \Big(|\Lambda|^2 |\partial_x z(\theta,x)|^2 + |\Gamma|^2 |z(\theta,x)|^2 \\
&\quad + |\digamma|^2 |z(t_k,x)|^2 \Big) d\theta dx \\
&\le 3\int_{t_k}^t \Big(|\Lambda|^2 \|\partial_x z(\theta,x)\|^2_{L^2([0,L];\mathbb{R}^n)} \\
&\quad + |\Gamma|^2 \|z(\theta,x)\|^2_{L^2([0,L];\mathbb{R}^n)} \\
&\quad + |\digamma|^2 \|z(t_k,x)\|^2_{L^2([0,L];\mathbb{R}^n)} \Big) d\theta \\
&\le 3\bar{h} \left(|\Lambda|^2 \frac{R}{\lambda_{\min}(\Theta_2) e^{-2\mu L}} \right. \\
&\quad \left. + \left(|\Gamma|^2 + |\digamma|^2 \right) \frac{R}{\lambda_{\min}(\Theta_1) e^{-2\mu L}} \right) \\
&= \omega. \qquad (44)
\end{aligned}
$$

In addition, since condition (38a) is satisfied, we have

$$
\begin{aligned}
&-2\sigma V_1(z) - (2\sigma - \beta) V_2(z) \\
&\le -(2\sigma - \beta)(V_1(z) + V_2(z)) \\
&< -(2\sigma - \beta)\varepsilon. \qquad (45)
\end{aligned}
$$

Therefore, substituting (41), (44) and (45) into (36), we have for all $t \in [t_k, t_{k+1}), k \in \mathbb{N}$,

$$
\begin{aligned}
\dot{V}(z) < &-(2\sigma - \beta)\varepsilon + \gamma\omega \\
&+ \gamma \frac{R}{\lambda_{\min}(\Theta_2) e^{-2\mu L}}. \qquad (46)
\end{aligned}
$$

Since (23) holds, we deduce from (46),

$$
\dot{V}(z) < -2\delta R \le -2\delta V(z). \qquad (47)
$$

Therefore, we have shown that $\dot{V}(z) + 2\delta V(z) \le 0$, whenever conditions (38) are satisfied.

Step 4: In this step, we show that if $z(t_k,\cdot) \in \mathfrak{L}_{V<R}$, then $z(t,\cdot) \in \mathfrak{L}_{V<R}, \forall t \in [t_k, t_{k+1})$. Consider z such that $z(t_k,\cdot) \in \mathfrak{L}_{V<R}$, assume that $\exists\ t^\circ \in (t_k, t_{k+1})$ s.t. $V(z(t^\circ,\cdot)) \ge R$. Let us then call T° the minimum of such t°, then $\forall t \in [t_k, T^\circ)$, $V(z(t,\cdot)) < R$. Therefore conditions (38) are going to be satisfied for any $t \in [t_k, T^\circ)$. From step 3, we know that V is going to decrease during that time interval, either continuously, or until V reaches below $max(\varepsilon, V(z(t_k,\cdot))/\alpha)$ and when it reaches that region, it never gets back out. Therefore, we have $V(z(T^\circ,\cdot)) < V(z(t_k,\cdot)) < R$, which contradicts the assumption that there exists $t^\circ \in (t_k, t_{k+1})$ such that $V(z(t^\circ,\cdot)) \ge R$.

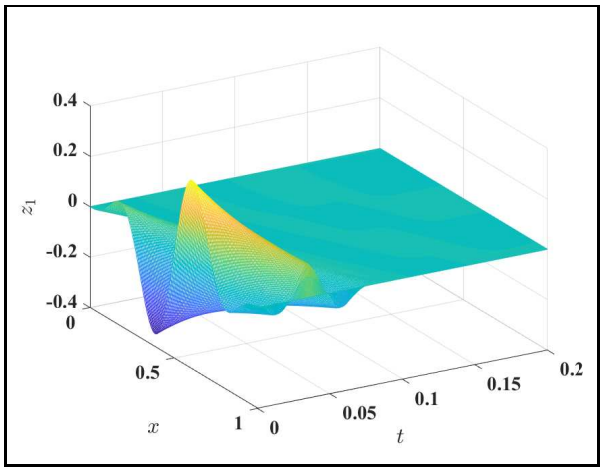

Fig. 2. Response of state z_1.

Summary: From step 3 and step 4, it is clear that $\dot{V}(z) + 2\delta V(z) \le 0$ wherever

$$
\begin{cases}
R > V(z(t,\cdot)) \ge \max(\varepsilon, V(z(t_k,\cdot))/\alpha), & \text{(48a)} \\
z(t_k,\cdot) \in \mathfrak{L}_{V<R}, & \text{(48b)}
\end{cases}
$$

and therefore, the conditions of Theorem 1 are satisfied, which concludes the proof of Rε-stability. ■

Remark 5. It is worth pointing out that we use several parameters and now we summarize each parameter in detail. For Rε-stability, R is the domain of attraction for a given Lyapunov function, ε specifies the positive invariant level set of V. They satisfy $0 < \varepsilon < R$. In this paper, we can fix R then compute ε or vice versa. α is a parameter introduced in the Lyapunov-Razumikhin method to define level set in which the time derivative of $V(z(t,\cdot))$ should be negative between two sampling interval, we choose it greater than 1. μ is related to the decay rate of V_1, V_2, and δ is related to the decay rate of V. γ and κ are found by line search to realize the conditions given in Theorem 1. Due to (23), we adjust γ, β to be the smallest possible and μ to be the largest possible. In the numerical section, we sort out their relationship: $\mu > 0,\ \delta > 0,\ \underline{\lambda} = \min_{i \in \{1,\dots,n\}} \lambda_i > 0,\ \sigma = \mu\underline{\lambda},\ 0 < \beta < 2\sigma$.

IV. NUMERICAL SIMULATION

In this section, we present a numerical example to illustrate the method we proposed in Section 3. Consider system (1) where

$$
\Lambda = \begin{bmatrix} 10 & 0 \\ 0 & 12 \end{bmatrix}, \quad \Gamma = \begin{bmatrix} 20 & 15 \\ 20 & 25 \end{bmatrix}, \quad \digamma = \begin{bmatrix} 2 & 0 \\ 5 & 4 \end{bmatrix},
$$

$$
L = 1,\ \bar{h} = 0.1,\ z_0(x) = \begin{bmatrix} 0.2(1-\cos 2\pi x)\sin 4\pi x \\ 0.15(1-\cos 4\pi x)\sin 2\pi x \end{bmatrix}.
$$

According to Remark 5, the parameters are selected as:

$$
\beta = 0.01, \quad \mu = 0.09, \quad \kappa = 1.8, \quad \delta = 0.001,
$$
$$
\gamma = 0.001, \quad \alpha = 1.001,
$$

which satisfy the conditions (18) and (23).

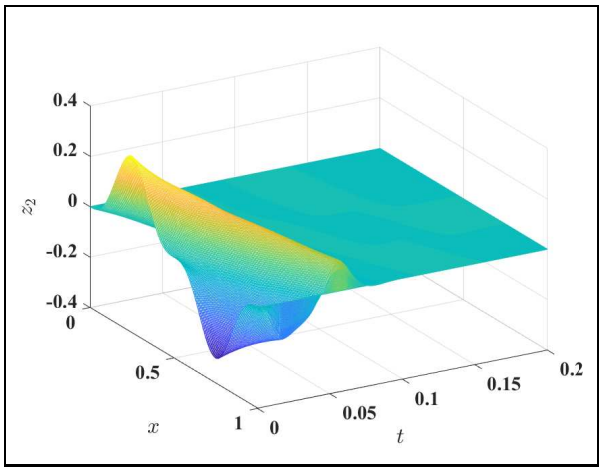

Fig. 3. Response of state z_2.

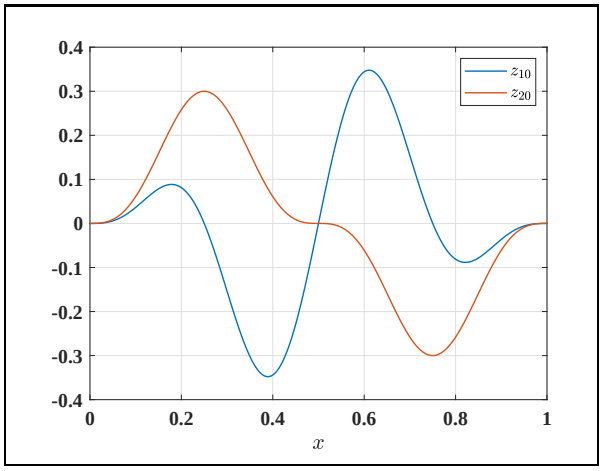

Fig. 4. Initial data of z_1 and z_2.

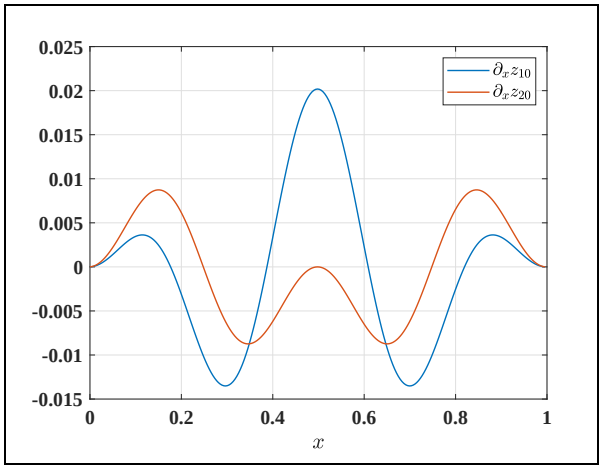

Fig. 5. Initial data of $\partial_x z_1$ and $\partial_x z_2$.

We fix $R = 20$, and choose

$$\Theta_1 = \begin{bmatrix} 2.8342 & -0.2219 \\ * & 1.3796 \end{bmatrix},$$

$$\Theta_2 = \begin{bmatrix} 3.4987 & -0.0635 \\ * & 2.0097 \end{bmatrix},$$

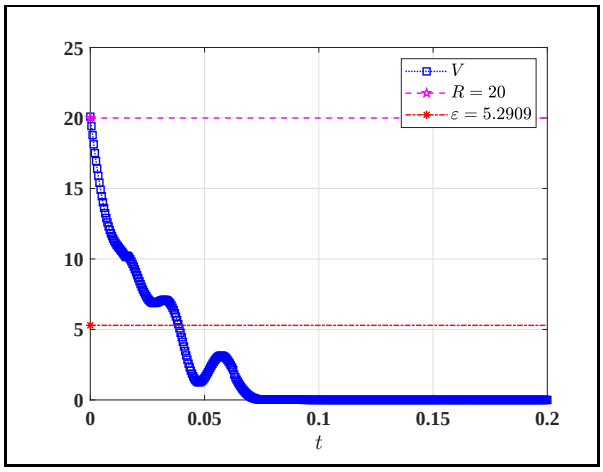

Fig. 6. Time-evolution of function V.

to observe the evolution of states.

The simulation results are introduced in Figs. 2-6. Figs. 2-3 present that both state trajectories converge to near the origin with the controller, respectively. The initial conditions satisfying the compatibility condition (3) are illustrated in Figs. 4-5. As can be seen from Fig. 6, the time-evolution of Lyapunov function $V(z(t,\cdot))$ decreases when $R > V(z(t,\cdot)) \geq \max(\varepsilon, V(z(t_k,\cdot))/\alpha), \alpha > 1$.

V. CONCLUSIONS

The main work of this paper is to use a sampling controller for distributed control of linear hyperbolic balance laws. The closed-loop system is reformulated based on Input-Output approach. New stability condition has been obtained by means of the Lyapunov-Razumikhin method. In the future, we will consider the global stability with controller discretized both in time and in space.

APPENDIX

Lemma 1. Consider the system (1)-(2) with initial condition z_0 satisfying Condition 1. Then $\forall t \in [t_k, t_{k+1}), k \in \mathbb{N}$, $\partial_x z(t,0) = \partial_x z(t,L) = 0$.

Proof: We recall system (1)

$$\begin{cases} \partial_t z(t,x) + \Lambda \partial_x z(t,x) + \Gamma z(t,x) + u(t,x) = 0, & \text{(49a)} \\ u(t,x) = F z(t_k,x), \forall t \in [t_k, t_{k+1}), k \in \mathbb{N}, & \text{(49b)} \\ z(t,0) = z(t,L) = 0, \forall t \geq 0, & \text{(49c)} \\ z(0,x) = z_0(x), \forall x \in [0,L], & \text{(49d)} \end{cases}$$

The time derivative of the boundary condition leads to

$$\partial_t z(t,0) = \partial_t z(t,L) = 0, \forall t \in [t_k, t_{k+1}), k \in \mathbb{N}. \quad (50)$$

Combining (49a) with (50), we obtain

$$\begin{cases} 0 = \partial_t z(t,0) = -\Lambda \partial_x z(t,0) - \Gamma z(t,0) - F z(t_k,0), \\ 0 = \partial_t z(t,L) = -\Lambda \partial_x z(t,L) - \Gamma z(t,L) - F z(t_k,L). \end{cases}$$

Since $z(t,0) = z(t,L) = 0, \forall t \geq 0$, we have

$$\partial_x z(t,0) = \partial_x z(t,L) = 0, \forall t \in [t_k, t_{k+1}), k \in \mathbb{N}. \quad (51)$$

■

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