

Static sliding mode control of systems with arbitrary relative degree by using artificial delay

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Abstract—Static output-feedback stabilization of systems with relative degree n with matched disturbances is considered. Assuming that the system is controllable, a static output-feedback sliding mode controller (SMC) is designed, where the output derivatives up to the order $(n - 1)$ are approximated by using the current and the delayed values of the output. Numerical examples illustrate the efficiency of the method.

Index Terms—Sliding mode control, static feedback, time-delay systems.

I. INTRODUCTION

State-Of-Art. Sliding mode control (SMC) has attractive features for the theoretically exact compensation of the matched uncertainties and disturbances, and finite-time convergence of the system's trajectory to the sliding surface [1]. The static output-feedback SMC paradigm for systems with relative degree one was introduced in [2], where only the measured output (and not its derivatives) was used as SMC surface. For systems with state delays, state-feedback SMC and static output-feedback SMC were suggested [3] and [4] respectively by using the descriptor approach [6].

Note that the presence of input delays destroy the convergence to the sliding motions, or even lead to the instability of the closed-loop system [7]. Practical stabilization of systems with input delays by static output-feedback SMC was suggested in [5] by using a singular perturbation approach.

In [8] using of artificial delay for static output-feedback SMC of systems with relative degree one was introduced. LMI-based conditions were proposed for the stability analysis of the resulting delayed closed-loop system.

A new approach to stabilization of the wide class of systems introducing artificial delay and estimation of the upper bound of such delay was proposed in [9],[10],[11] using the Taylor expansion with the integral remainder and appropriate Lyapunov-Krasovskii functionals.

In this paper we propose a static output-feedback SMC design for systems with relative degree n with matched

perturbations. For the design of sliding surface for such systems, the complete information about the system state is usually required. For the output-based sliding mode controllers, the estimation of the states has been based on Luenberger observers [12], additive filters [13], and robust exact differentiators/observers with finite-time convergence [14],[15],[16],[22].

Paper Novelty. In this paper we consider stabilization of systems with relative degree n and matched disturbances. The objective is to design a static output-feedback sliding mode controller that for the estimation of the states employs the delayed values of the output. With this aim:

- the delayed sliding surface is proposed using estimation of the system states based on the artificial delay;
- finite-time attractivity of the proposed delayed sliding surface is proved (Theorem 2);
- Lyapunov-Krasovskii functional is employed to achieve the convergence of system states to the neighborhood of zero (Theorem 1);
- the design parameters are chosen for a good trade-off between the approximation accuracy and the reduction of controller gain.

Notations. For a real symmetric matrix X , $X \leq 0$ (respectively, $X < 0$) means that the matrix X is negative semidefinite (respectively, negative definite). $B(\cdot, \cdot)$ is Euler's beta function. Define a symmetric matrix as $\text{He}(M) = M + M^T$, and the symmetric elements of a symmetric matrix is represented by \star . The notations $\|\cdot\|$ and $|\cdot|$ stand for the Euclidean norm and 1-norm of a vector, respectively. By using $O(h)$, a matrix/scalar function of $h \in \mathbb{R}_+$ is defined to satisfy $\lim_{h \rightarrow 0^+} \frac{1}{h} O(h) = m$, where $m > 0$.

II. PROBLEM FORMULATION AND PRELIMINARIES

Consider the following n -th order relay system:

$$z^{(n)}(t) = \sum_{i=1}^n A_i z^{(i-1)}(t) + B_1 [u(t) + d(z, \dots, z^{(n-1)}, t)], \quad (1)$$

where $n \geq 2$, $z(t) = z^{(0)}(t) \in \mathbb{R}^k$ is the measurement, $z^{(i)}(t)$ is the i th derivative of $z(t)$, $u(t) \in \mathbb{R}^m$ is the control input, and $d(z, z^{(1)}, \dots, z^{(n)}, t)$ is the matched perturbation. Model (1) represents the linearized nonlinear system of relative degree n with matched state-dependent uncertainties, which represents the local behavior of any Lipschitz system linearized in the vicinity of the equilibrium.

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For the convenience of representation, we define

$$x(t) = \text{col}\{z(t), z^{(1)}(t), \dots, z^{(n-1)}(t)\} \\ \triangleq \text{col}\{x_1(t), x_2(t), \dots, x_n(t)\}.$$

System (1) is equivalent to

$$\dot{x}(t) = Ax(t) + B(u(t) + d(x, t)) \quad (2)$$

which arrives at

$$x_{n+1}(t) \triangleq \dot{x}_n(t) = \bar{A}x(t) + B_1[u(t) + d(x, t)] \quad (3)$$

where

$$A = \begin{bmatrix} 0 & I_k & 0 & \dots & 0 \\ 0 & 0 & I_k & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & I_k \\ A_1 & A_2 & A_3 & \dots & A_n \end{bmatrix}, B = \begin{bmatrix} 0 \\ 0 \\ \dots \\ 0 \\ B_1 \end{bmatrix} \\ \bar{A} = [A_1 \quad A_2 \quad A_3 \quad \dots \quad A_n].$$

Inspired from [1], the following linear sliding motion is always used for system (1):

$$s^*(t) = \bar{C}x(t) \quad (4)$$

and the design matrix is given as

$$\bar{C} = [C_1 \quad C_2 \quad \dots \quad C_n]$$

with $C_l \in \mathbb{R}^{m \times k}$, $l = 1, 2, \dots, n-1$.

Since only $z(t)$ is accessible to the controller, we approximate the derivatives by a few past measurements:

$$\hat{x}(t, h) \approx N^{-1}(h)x(t) \quad (5)$$

which becomes

$$x(t) \approx N(h)\hat{x}(t, h) \quad (6)$$

where $N(h) = (MF(h))^{-1}$, $\hat{x}(t, h) = M\bar{x}(t, h)$, $\bar{x}(t, h) = F(h)x(t)$, and

$$\hat{x}(t, h) = \text{col}\{x_1(t), x_1(t-h), \dots, x_1(t-(n-1)h)\} \\ F(h) = \text{diag}\{I_k, -hI_k, \dots, (-h)^{n-1}I_k\}$$

$$M = \begin{bmatrix} I_k & 0 & 0 & \dots & 0 \\ I_k & I_k & \frac{1}{2!}I_k & \dots & \frac{1}{(n-1)!}I_k \\ \dots & \dots & \dots & \dots & \dots \\ I_k & jI_k & \frac{j^2}{2!}I_k & \dots & \frac{j^{n-1}}{(n-1)!}I_k \\ \dots & \dots & \dots & \dots & \dots \\ I_k & (n-1)I_k & \frac{(n-1)^2}{2!}I_k & \dots & \frac{(n-1)^{n-1}}{(n-1)!}I_k \end{bmatrix}.$$

Matrix M is a Vandermonde-type matrix. Equation (5) indicates that the states of system (5) can be estimated by using the past measurements at the time instant j ($j = 0, 1, \dots, n-1$).

To this end, the following delay-dependent sliding variable is adopted:

$$s(t) = \bar{C}N(h)\hat{x}(t, h)$$

where h is the artificial time delay.

Our objective is to design the artificial time-delay estimator (5) in sliding mode control, which avoid introducing additional dynamics for estimation and enhancing the robustness to measurement noises. Then, we present the problem formulation of this work:

right) a delay-dependent sliding surface $s(t) = 0$ and an SMC law $u(t)$ will be designed such that the sliding motion $s(t)$ is quadratically stable in finite time.

To this end, the following lemmas are necessary.

Lemma 1. (Jensen's Inequality [10]) Define $G = \int_a^b f(s)x(s)ds$, where $a < b$, $f: [a, b] \rightarrow [0, \infty)$, $x(s) \in \mathbb{R}^n$, and the integration concerned is well defined. Then, for any $n \times n$ matrix $R > 0$, the following inequality holds:

$$G^T R G \leq \int_a^b f(\theta)d\theta \int_a^b f(s)x^T(s)R x(s)ds. \quad (7)$$

Lemma 2. [20, 23] Given a positive scalar $\bar{\epsilon}$, and symmetric matrices M_1 , M_2 and M_3 with the same dimensions, the inequality holds for any $\epsilon \in (0, \bar{\epsilon}]$: $M_1 + \epsilon M_2 + \epsilon^2 M_3 \leq 0$, if and only if $M_1 \leq 0$, $M_1 + \bar{\epsilon} M_2 \leq 0$, $M_1 + \bar{\epsilon} M_2 + \bar{\epsilon}^2 M_3 \leq 0$.

A. Taylor's formula with the integral remainder

For n -times continuously differentiable function $x_1(t)$ with absolutely continuous $x_n(t)$ over the time interval $[t-jh, t]$, the Taylor expansion is written as

$$x_1(t-jh) = x_1(t) + j(-h)x_2(t) + \frac{j^2(-h)^2}{2!}x_3(t) \\ + \dots + \frac{j^{n-1}(-h)^{n-1}}{(n-1)!}x_n(t) + \delta_j(t, h) \quad (8) \\ = M(j)F(h)x(t) + \delta_j(t, h)$$

where $j = 0, 1, \dots, n-1$, and

$$M(j) = \begin{bmatrix} I_k & jI_k & \frac{j^2}{2!}I_k & \dots & \frac{j^{n-1}}{(n-1)!}I_k \end{bmatrix}.$$

Wherein, the remainder $\delta_j(t, h)$ has two equivalent forms:

$$\delta_j(t, h) = \frac{(-1)^n}{(n-1)!} \int_{t-jh}^t (s-t+jh)^{n-1} x_{n+1}(s) ds \quad (9)$$

$$\delta_j(t, h) = \frac{(-1)^{n-1}}{(n-2)!} \int_{t-jh}^t (s-t+jh)^{n-2} \mu(s, t) ds \quad (10)$$

with $\mu(s, t) = x_n(s) - x_n(t)$.

Remark 1. The form of the integral terms of Lyapunov functionals is constructed based on (9) (as in [10]). Here, the representation of (10) is used for estimating the bound of the remainder $\delta_j(t, h)$.

Then, we have

$$\hat{x}(t, h) = N^{-1}(h)x(t) + \Delta_1(t, h) \quad (11)$$

where $\Delta_1(t, h) = \text{col}\{\delta_0(t, h), \delta_1(t, h), \dots, \delta_{n-1}(t, h)\}$.

It is easy to verify that, if $\lim_{h \rightarrow 0} |\delta_i(t, h)| = 0$, $\lim_{h \rightarrow 0} \frac{1}{h^{n+1}} |\delta_i(t, h)| = 0$, then we have

$$\Delta_1(t) = O(h^n). \quad (12)$$

The approximation error is given by $\Delta_1(t)$, which is used to verify the approximation accuracy of the artificial time-delay method. Hence, we obtain $s(t) = s^*(t) + \bar{C}N(h)\Delta_1(t, h)$ $e_s(t) = s(t) - s^*(t) = \bar{C}N(h)\Delta_1(t, h) = O(h^n)$.

Remark 2. The sliding motion $s(t) = 0$ can imitate the behavior of $s^*(t) = 0$ with the accuracy $O(h^n)$ such that

the delayed output-feedback controller contains more system information than the pure static one.

It is reasonable to assume that

$$|x_{n+1}(t)| \leq \beta \quad (13)$$

where β is the upper bound of $x_{n+1}(t)$, and β is a scalar defined by the domain in which model (1) is valid. The following assumptions will be used:

- A1) The perturbation term $d(x, t)$ is bounded, i.e. $|d(x, t)| \leq d^*$, where d^* is a positive scalar.
A2) The dynamic $x_n(t)$ is β -Lipschizian on the small time interval $[t - jh - \delta, t + \delta]$:

$$|\mu(s, t)| \leq \beta|t - s|, \text{ for all } s \in [t - jh, t] \quad (14)$$

where δ is a small positive scalar.

Remark 3. All system states $x_i(t)$ ($i = 1, 2, \dots, n$) are still continuous, i.e. $x_i(t) = x_i(t_0) + \int_{t_0}^t x_{i+1}(s)ds$, $\lim_{t \rightarrow t_0^+} x_i(t) = \lim_{t \rightarrow t_0^+} x_i(t)$, which reveals the validity of Assumption A2). For example, a mass-spring-damper system contains the states of position and velocity. The acceleration can change the directions by using forces with $\text{sgn}(\cdot)$, but the velocity and position are still continuous.

III. MAIN RESULTS

A. Analysis of sliding motions

From (10), we have

$$\dot{\Delta}_1(t, h) = \Delta_2(t, h) + Y(h)x_{n+1}(t) \quad (15)$$

where $Y(h) = \frac{(-1)^{n-2}}{(n-1)!} \text{col}\{0, h^{n-1}I_k, \dots, [(n-1)h]^{n-1}I_k\}$. The form of $\Delta_2(t, h)$ is written as

$$\Delta_2(t, h) = \text{col}\{\rho_0(t), \rho_1(t, h), \dots, \rho_{n-1}(t, h)\}$$

where $\rho_j(t, h)$ takes the following two forms:

$$\rho_j(t, h) = \frac{(-1)^{n+1}}{(n-2)!} \int_{t-jh}^t (s-t+jh)^{n-2} x_{n+1}(s)ds \quad (16)$$

$$\rho_j(t, h) = \begin{cases} \frac{(-1)^n}{(n-3)!} \int_{t-jh}^t (s-t+jh)^{n-3} \mu(s, t)ds, n \geq 3 \\ x_n(t) - x_n(t-jh), n = 2. \end{cases} \quad (17)$$

From (2) and (15), the derivative of $s(t)$ is given as

$$\dot{s}(t) = \bar{C}\dot{x}(t) + \bar{C}H(h)x_{n+1}(t) + \bar{C}N(h)\Delta_2(t, h) \quad (18)$$

where $H(h) = N(h)Y(h)$. By virtue of (1) and (2), we can rewrite (18) as

$$\dot{s}(t) = J_1(h)x(t) + J_2(h)(u(t) + d(x, t)) + \bar{C}N(h)\Delta_2(t, h)$$

where $J_1(h) = \bar{C}A + \bar{C}H(h)\bar{A}$, $J_2(h) = \bar{C}B + \bar{C}H(h)B_1$.

The equivalent control law is formulated as

$$u_{eq}^*(t) = -d(x, t) - J_2^{-1}(h)(J_1(h)x(t) + \bar{C}N(h)\Delta_2(t, h)) \quad (19)$$

which requires the full state information of $x(t)$. By using the estimation (6) in (19), we arrive at

$$u_{eq}(t) = -d(x, t) - J_2^{-1}(h)[J_1(h)N(h)\hat{x}(t, h) + \bar{C}N(h)\Delta_2(t, h)] \quad (20)$$

which is equivalent to

$$u_{eq}(t) = -d(x, t) - J_2^{-1}(h)[J_1(h)x(t) + J_1(h)N(h)\Delta_1(t) + \bar{C}N(h)\Delta_2(t, h)]. \quad (21)$$

Remark 4. In (20), the equivalent control law is implemented without using the derivatives of the measurement output.

B. Stabilization of the closed-loop system

Based on the descriptor model transformation in [3, 5], substituting (21) into system (2) yields

$$E\dot{\eta}(t) = \mathcal{A}_c\eta(t) + \mathcal{A}_{d1}\Delta_1(t) + \mathcal{A}_{d2}\Delta_2(t) \quad (22)$$

where $\eta(t) = \text{col}\{x(t), x_{n+1}\}$, and

$$E = \begin{bmatrix} I_{\bar{n}} & 0 \\ 0 & 0 \end{bmatrix}, \quad \mathcal{A} = \begin{bmatrix} \Gamma_1 & 0 \\ \bar{A} & -I_k \end{bmatrix},$$

$$\mathcal{B} = \begin{bmatrix} 0 \\ B_1 \end{bmatrix}, \quad \Gamma_1 = [0 \quad I_{\bar{n}-k}], \quad \bar{n} = nk,$$

$$\mathcal{A}_c = \mathcal{A} + \mathcal{B}\Gamma_2(h), \quad \Gamma_2(h) = [-J_2^{-1}(h)J_1(h) \quad 0]$$

$$\mathcal{A}_{d1} = -\mathcal{B}J_2^{-1}(h)J_1(h)N(h), \quad \mathcal{A}_{d2} = -\mathcal{B}J_2^{-1}(h)\bar{C}N(h).$$

Define $A_m = [\bar{A} - B_1J_2^{-1}(h)J_1(h) \quad 0]$, and

$$A_{n1} = -B_1J_2^{-1}(h)J_1(h)N(h), \quad A_{n2} = -B_1J_2^{-1}(h)\bar{C}N(h).$$

With $u_{eq}(t)$, equation (1) can be further represented as

$$x_{n+1}(t) = A_m\eta(t) + A_{n1}\Delta_1(t) + A_{n2}\Delta_2(t). \quad (23)$$

Then, we will derive the delay-dependent LMI conditions for the stabilization of the closed-loop system (22).

Theorem 1. For the given tuning scalar $\epsilon^* > 0$ and the prescribed matrix \bar{C} , the descriptor system (22) is asymptotically stable, if there exist symmetric matrices $P_1 \in \mathbb{R}^{(\bar{n}-k) \times (\bar{n}-k)}$, $P_3 \in \mathbb{R}^{k \times k}$, $X \in \mathbb{R}^{k \times k}$, and $W \in \mathbb{R}^{k \times k}$, and the matrix $P_2 \in \mathbb{R}^{k \times (\bar{n}-k)}$ such that the following inequalities hold:

$$\begin{bmatrix} \text{He}(P^T \mathcal{A}_c) & P^T \mathcal{A}_{d1} & P^T \mathcal{A}_{d2} & 0 \\ * & -\mathcal{W} & 0 & 0 \\ * & * & -\mathcal{X} & 0 \\ * & * & * & -\bar{Q} \end{bmatrix} \leq 0 \quad (24)$$

$$\begin{bmatrix} \text{He}(P^T \mathcal{A}_c) & P^T \mathcal{A}_{d1} & P^T \mathcal{A}_{d2} & \sqrt{\epsilon^*} A_m^T \bar{X} \\ * & -\mathcal{W} & 0 & \sqrt{\epsilon^*} A_{n1}^T \bar{X} \\ * & * & -\mathcal{X} & \sqrt{\epsilon^*} A_{n2}^T \bar{X} \\ * & * & * & -\bar{X} \end{bmatrix} \leq 0 \quad (25)$$

$$\begin{bmatrix} \text{He}(P^T \mathcal{A}_c) & P^T \mathcal{A}_{d1} & P^T \mathcal{A}_{d2} & \sqrt{\epsilon^*} A_m^T \bar{Q}(h^*) \\ * & -\mathcal{W} & 0 & \sqrt{\epsilon^*} A_{n1}^T \bar{Q}(h^*) \\ * & * & -\mathcal{X} & \sqrt{\epsilon^*} A_{n2}^T \bar{Q}(h^*) \\ * & * & * & -\bar{Q}(h^*) \end{bmatrix} \leq 0 \quad (26)$$

where $h^* = {}^{2(n-1)}\sqrt{\epsilon^*}$, and

$$\bar{X} = \sum_{j=0}^{n-1} \bar{X}_{j+1}, \quad \bar{X}_{j+1} = j^{2n-1} X_{j+1}$$

$$\bar{Q} = \sum_{j=0}^{n-1} (j)^{2(n-1)} X_{j+1}, \quad \bar{Q}(h^*) = \sum_{j=0}^{n-1} Q_{j+1}(h^*)$$

$$Q_{j+1}(h^*) = (j)^{2(n-1)} [X_{j+1} + (jh^*)^2 W_{j+1}].$$

Proof. The following Lyapunov functional is used:

$$V(\eta, y, t) = V_1(\eta, t) + V_2(x_{n+1}) + V_3(x_{n+1})$$

where

$$V_1(\eta, t) = \eta^T(t)EP\eta(t)$$

$$V_2(x_{n+1}) = \sum_{j=0}^{n-1} (jh)^n \int_{t-jh}^t (s-t+jh)^n \nu_{j+1}(s) ds$$

$$V_3(x_{n+1}) = \sum_{j=0}^{n-1} (jh)^{n-1} \int_{t-jh}^t (s-t+jh)^{n-1} \varphi_{j+1}(s) ds$$

$$\nu_{j+1}(s) = x_{n+1}^T(s)W_{j+1}x_{n+1}(s), \quad W_{j+1} = W_{j+1}^T > 0$$

$$\varphi_{j+1}(s) = x_{n+1}^T(s)X_{j+1}x_{n+1}, \quad X_{j+1} = X_{j+1}^T > 0.$$

Matrix P is specified as

$$P = \begin{bmatrix} P_1 & 0 \\ P_2 & P_3 \end{bmatrix}, \quad P_1 = P_1^T > 0.$$

Differentiating $V_1(\eta, t)$ with respect to time t yields

$$\dot{V}_1(\eta, t) = \eta^T(t)\text{He}(P^T \mathcal{A}_c)\eta(t) + 2\eta^T(t)P^T \mathcal{A}_{d1}\Delta_1(t, h) + 2\eta^T(t)P^T \mathcal{A}_{d2}\Delta_2(t, h). \quad (27)$$

To deal with the term with $\Delta_1(t)$ in (27), the functional $V_2(x_{n+1})$ is taken into account, with its derivative given as

$$\begin{aligned} \dot{V}_2(x_{n+1}) &= \sum_{j=0}^{n-1} (jh)^{2n} x_{n+1}^T(t)W_{j+1}x_{n+1}(t) \\ &\quad - n \sum_{j=0}^{n-1} (jh)^n \int_{t-jh}^t (s-t+jh)^{n-1} \nu_{j+1}(s) ds. \end{aligned} \quad (28)$$

It follows from Lemma 1 that

$$\begin{aligned} &-n \sum_{j=0}^{n-1} (jh)^n \int_{t-jh}^t (s-t+jh)^{n-1} \nu_{j+1}(s) ds \\ &\leq -\Delta_1^T(t, h)\mathcal{W}\Delta_1(t, h), \end{aligned} \quad (29)$$

where $\mathcal{W} = (n!)^2 \text{diag}\{W_0, W_1, \dots, W_{n-1}\}$.

From (28) and (29), we have

$$\begin{aligned} \dot{V}_2(x_{n+1}) &\leq \zeta^T(t)\mathcal{V}^T \left\{ \sum_{j=0}^{n-1} (jh)^{2n} W_{j+1} \right\} \mathcal{V}\zeta(t) \\ &\quad - \Delta_1^T(t, h)\mathcal{W}\Delta_1(t, h), \end{aligned} \quad (30)$$

where

$$\mathcal{V} = \begin{bmatrix} A_m & A_{n1} & A_{n2} \end{bmatrix}, \quad \zeta(t) = \text{col}\{\eta(t), \Delta_1(t), \Delta_2(t)\}.$$

Moreover, the time derivative of the delay-dependent functional $V_3(x_{n+1})$ along the solution of (22) is written as

$$\begin{aligned} \dot{V}_3(x_{n+1}) &= \sum_{j=0}^{n-1} (jh)^{2(n-1)} x_{n+1}^T(t)X_{j+1}x_{n+1}(t) \\ &\quad - (n-1) \sum_{j=0}^{n-1} (jh)^{n-1} \int_{t-jh}^t (s-t+jh)^{n-2} \varphi_{j+1}(s) ds. \end{aligned}$$

By using the representation (16) and applying Lemma 1, we find

$$\begin{aligned} &-(n-1) \sum_{j=0}^{n-1} (jh)^{n-1} \int_{t-jh}^t (s-t+jh)^{n-2} \varphi_{j+1}(s) ds \\ &\leq -\Delta_2^T(t, h)\mathcal{X}\Delta_2(t, h) \end{aligned} \quad (31)$$

where

$$\mathcal{X} = (n-1)!^2 \text{diag}\{X_0, X_1, \dots, X_{n-1}\}.$$

Taking (31) into consideration, it is easy to obtain that

$$\begin{aligned} \dot{V}_3(x_{n+1}) &\leq \zeta^T(t)\mathcal{V}^T \sum_{j=0}^{n-1} [(jh)^{2(n-1)} X_{j+1}] \mathcal{V}\zeta(t) \\ &\quad - \Delta_2^T(t, h)\mathcal{X}\Delta_2(t, h). \end{aligned} \quad (32)$$

Adding (27), (30) to (32), we have

$$\begin{aligned} \dot{V}(t) &\leq \zeta^T(t)\Xi\zeta(t) \\ &\quad + \zeta^T(t)\mathcal{V}^T \sum_{j=0}^{n-1} [(jh)^{2(n-1)} X_{j+1} + (jh)^{2n} W_{j+1}] \mathcal{V}\zeta(t) \end{aligned}$$

where

$$\Xi = \begin{bmatrix} \text{He}(P^T \mathcal{A}_c) & P^T \mathcal{A}_{d1} & P^T \mathcal{A}_{d2} \\ \star & -\mathcal{W} & 0 \\ \star & \star & -\mathcal{X} \end{bmatrix}.$$

After some manipulation using the Schur Complement Lemma, the inequality $\dot{V}(t) \leq 0$ is equivalently represented as

$$\Gamma(\epsilon, h) \leq 0 \quad (33)$$

where

$$\begin{aligned} \epsilon &= h^{2(n-1)}, \quad \bar{Q}(h) = \sum_{j=0}^{n-1} (j)^{2(n-1)} [X_{j+1} + (jh)^2 W_{j+1}] \\ \Gamma(\epsilon, h) &= \begin{bmatrix} \text{He}(P^T \mathcal{A}_c) & P^T \mathcal{A}_{d1} & P^T \mathcal{A}_{d2} & \epsilon A_m^T Q(h) \\ \star & -\mathcal{W} & 0 & \epsilon A_{n1}^T Q(h) \\ \star & \star & -\mathcal{X} & \epsilon A_{n2}^T Q(h) \\ \star & \star & \star & -\epsilon Q(h) \end{bmatrix}. \end{aligned}$$

Performing the congruent transformation given as

$$\mathcal{T}_s = \text{diag}\{I, I, I, \frac{1}{\sqrt{\epsilon}}I\}$$

to inequality (33) yields

$$\Gamma^*(\epsilon, h) \leq 0 \quad (34)$$

where

$$\Gamma^*(\epsilon, h) = \begin{bmatrix} \text{He}(P^T \mathcal{A}_c) & P^T \mathcal{A}_{d1} & P^T \mathcal{A}_{d2} & \sqrt{\epsilon} A_m^T Q(h) \\ \star & -\mathcal{W} & 0 & \sqrt{\epsilon} A_{n1}^T Q(h) \\ \star & \star & -\mathcal{X} & \sqrt{\epsilon} A_{n2}^T Q(h) \\ \star & \star & \star & -Q(h) \end{bmatrix}.$$

It follows from Lemma 2 that the sufficient conditions for achieving (34) are given as

$$\exists h^* > 0, \text{ s.t. } \Gamma(0, 0) \leq 0, \quad \Gamma(\epsilon^*, 0) \leq 0, \quad \Gamma(\epsilon^*, h^*) \leq 0,$$

for all $h \in (0, h^*]$, with $\epsilon^* = (h^*)^{2(n-1)}$, which can be transformed into LMIs (24)–(26). This completes the proof. \square

Remark 5. Theorem 1 can be used for investigating the upper bound of the delay h . We can set any h satisfying $0 < h \leq h^*$ in the formulation of $s(t)$ and $u(t)$.

Remark 6. Consider $n(t)$ as the measurement noises with a frequency f_n . The robustness to the high-frequency $n(t)$ can be concluded from the averaging theory, which requires

$$f_n \ll 1/h, \quad 0 < h \leq h^* \quad (35)$$

For low-frequency noise $n(t)$, the artificial time-delay estimator will not work as a filter. Thus, using the results in Theorem 1, it is possible to choose the suitable time delay for the finite-time convergence to the vicinity of origin.

C. Attractivity of the sliding surface

In this section, the attractivity of the delay-dependent sliding surface $s(t) = 0$ will be analyzed. The physical SMC law is

$$u(t) = -kJ_2^{-1}(h)\text{sgn}(s(t)) - J_2^{-1}(h)J_1(h)N(h)\hat{x}(t, h) \quad (36)$$

where k is a positive scalar. Moreover, an equivalent form of (36) can be represented as

$$u(t) = -kJ_2^{-1}(h)\text{sgn}(s(t)) - J_2^{-1}(h)J_1(h)x(t) - J_2^{-1}(h)J_1(h)N(h)\Delta_1(t, h). \quad (37)$$

Next, we will estimate the upper bound of $|\Delta_1(t, h)|$ and $|\Delta_2(t, h)|$. With Assumption A2), it is easy to verify that

$$\begin{aligned} |\Delta_1(t, h)| &= \sum_{j=0}^{n-1} |\delta_j(t, h)| \\ &= \frac{1}{(n-2)!} \sum_{j=0}^{n-1} \int_{t-jh}^t |(s-t+jh)^{n-2} \cdot \mu(s, t)| ds \\ &\leq \frac{1}{(n-2)!} \sum_{j=0}^{n-1} \int_{t-jh}^t |(s-t+jh)^{n-2}| \cdot |\mu(s, t)| ds \\ &\leq b_{r1}(\beta) \end{aligned}$$

where

$$b_{r1}(\beta) = \frac{\beta}{(n-2)!} \sum_{j=0}^{n-1} \int_{t-jh}^t |(s-t+jh)^{n-2}| \cdot |t-s| ds.$$

Similarly, we obtain that $|\Delta_2(t, h)| \leq b_{r2}(\beta)$, where

$$\begin{aligned} b_{r2}(\beta) &= \begin{cases} g(\beta), & n \geq 3 \\ |x_n(t) - x_n(t-jh)|, & n = 2, \end{cases} \\ g(\beta) &= \frac{\beta}{(n-3)!} \sum_{j=0}^{n-1} \int_{t-jh}^t |(s-t+jh)^{n-3}| \cdot |t-s| ds. \end{aligned}$$

Using (37) in the representation of $\dot{s}(t)$ yields

$$\begin{aligned} \dot{s}(t) &= -k\text{sgn}(s(t)) - J_1N(h)\Delta_1(t, h) \\ &\quad + J_2d(x, t) + \bar{C}N(h)\Delta_2(t, h). \end{aligned} \quad (38)$$

The switching term $-k\text{sgn}(s(t))$ is used to compensate for $d(x, t)$, $\Delta_1(t, h)$ and $\Delta_2(t, h)$ to force the state trajectories to be attractive to the sliding surface $s(t) = 0$.

The following theorem investigates the reachability of the sliding motion in finite time.

Theorem 2. *Under Assumptions A1)–A2), the control input (36) makes the sliding surface stable and globally attractive in finite time, if the following condition holds:*

$$k \geq \gamma_1\beta + \|J_2\|d^* + k_0 \quad (39)$$

$$\gamma_2\beta \geq k\|B_1J_1^{-1}\| + \|B_1\|d^* + k_1 \quad (40)$$

where k_0 and k_1 are positive scalars, and

$$\begin{aligned} \gamma_1 &= \|J_1N(h)\|b_1^* + \|\bar{C}N(h)\|b_2^*, \\ \gamma_2 &= 1 - \|A_{n1}\|b_1^* - \|A_{n2}\|b_2^*, \\ b_1^* &= \sum_{j=0}^{n-1} \frac{1}{(n-2)!} (jh)^n B(2, n-1), \\ b_2^* &= \begin{cases} \sum_{j=0}^{n-1} \frac{1}{(n-3)!} (jh)^{n-1} B(2, n-2), & n \geq 3 \\ h, & n = 2 \end{cases} \end{aligned} \quad (41)$$

and the reaching time is given as $t_r = \sqrt{2V(0)}/k_0$.

Proof. Consider the Lyapunov function as $V(t) = 0.5s^T(t)s(t)$. By differentiating $V(t)$ with respect to time t , we have

$$\begin{aligned} \dot{V}(t) &= s(t)\dot{s}(t) \\ &= s(t)[-k\text{sgn}(s(t)) - J_1(h)N(h)\Delta_1(t, h) \\ &\quad + J_2(h)d(x, t) + \bar{C}N(h)\Delta_2(t, h)], \quad N(h) \quad (42) \\ &\leq -k|s(t)| + \|s(t)\| \cdot \|J_2(h)\| \cdot \|d(x, t)\| \\ &\quad + \|s(t)\| \cdot \|J_1(h)N(h)\| \cdot \|\Delta_1(t, h)\| \\ &\quad + \|s(t)\| \cdot \|\bar{C}N(h)\| \cdot \|\Delta_2(t, h)\|. \end{aligned}$$

Note that $\|s(t)\| \leq |s(t)|$, $\|\Delta_1(t, h)\| \leq |\Delta_1(t, h)|$, and $\|\Delta_2(t, h)\| \leq |\Delta_2(t, h)|$. With Assumption A1) and A2), inequality (42) becomes

$$\begin{aligned} \dot{V}(t) &\leq -|s(t)| \cdot [k - \|J_1(h)N(h)\| \cdot b_{r1}(\beta) \\ &\quad - \|J_2(h)\|d^* - \|\bar{C}N(h)\| \cdot b_{r2}(\beta)] \end{aligned}$$

which indicates that (39). Finally, we obtain that $\dot{V}(t) \leq -k_0|s| < 0$, which guarantees the convergence of system (1) towards the surface $s(t) = 0$, and

$$\begin{aligned} b_{r1}(\beta) &= \frac{\beta}{(n-2)!} \sum_{j=0}^{n-1} \int_{t-jh}^t |(s-t+jh)^{n-2}| \cdot |t-s| ds \\ &= \beta \sum_{j=0}^{n-1} (jh)^n \int_0^1 |s^{n-1}(1-s)| ds = b_1^*\beta, \\ b_{r2}(\beta) &= \frac{\beta}{(n-3)!} \sum_{j=0}^{n-1} \int_{t-jh}^t |(s-t+jh)^{n-3}| \cdot |t-s| ds \\ &= \sum_{j=0}^{n-1} (jh)^{n-1} \int_0^1 |s^{n-2}(1-s)| ds \cdot \beta = b_2^*\beta, \quad n \geq 3, \\ b_{r2}(\beta) &= h\beta = b_2^*\beta, \quad n = 2. \end{aligned}$$

Here, b_1^* and b_2^* are the linear coefficients.

Hence, the state trajectory is capable to reach the sliding surface $s(t) = 0$ in finite time, and the reaching time given below can be adjusted by changing k_0 :

$$t_r = |s(0)|/k_0 = \sqrt{2V(0)}/k_0.$$

Moreover, with control law in (36), we obtain

$$\begin{aligned} x_{n+1}(t) &= A_m\eta(t) + A_{n1}\Delta_1(t, h) + A_{n2}\Delta_2(t, h) \\ &\quad - kB_1J_2^{-1}(h)\text{sgn}(s(t)) + B_1d(x, t) \end{aligned}$$

which implies that

$$\|x_{n+1}(t)\| \leq \|A_m \eta(t)\| + \beta \|A_{n1}\| b_1^* + \beta \|A_{n2}\| b_2^* + k \|B_1 J_2^{-1}(h)\| + \|B_1\| d^*.$$

Considering that $\|x_{n+1}(t)\| < |x_{n+1}(t)| < \beta$, we arrive at

$$k_1 + \beta \|A_{n1}\| b_1^* + \beta \|A_{n2}\| b_2^* + k \|B_1 J_2^{-1}(h)\| + \|B_1\| d^* < \beta. \quad (43)$$

Note that in (40), the item k_1 is added to (43) in order to compensate the bounded item $\|A_m \eta(t)\|$, i.e. $\|A_m \eta(t)\| \leq |(\bar{A} - B_1 J_2^{-1}(h) J_1(h))x| < k_1$. This completes the proof. \square

Remark 7. The upper bounds of estimation errors are $b_1^* \beta = O(h^n)$ and $b_2^* \beta = O(h^n)$, which are small scalars for small enough delay h . Thus, $\gamma_1 \beta \approx 0$, $1 \geq \gamma_2 > 0$. First, we approximately choose k satisfying $k \geq \|J_2\| d^* + k_0$. Then, the value of β is determined via $\beta \geq k \|B_1 J_1^{-1}\| / \gamma_2 + \|B_1\| d^* / \gamma_2 + k_1 / \gamma_2$.

Design Steps. The model (1) is derived based on the linearization around the equilibrium. Then, the matrices A_i and B_1 are obtained, where the domain for model (2) is defined as $|x(t)| \leq \psi$. Here ψ is a known scalar. The upper bound d^* is known. With Assumptions A1)–A2), we will follow

- 1) Select k_1 that satisfies

$$|(\bar{A} - B_1 J_2^{-1} J_1)x| \leq |(\bar{A} - B_1 J_2^{-1} J_1)| \cdot \psi \leq k_1.$$

- 2) Choose \bar{C} , and formulate the ideal sliding surface $s^*(t)$ and the delay-dependent sliding surface $s(t)$.

- 3) Find $h^* = \max h$, if the solution to (24)–(26) exists:

$$\max h, \text{ s.t. LMIs (24)–(26) hold;}$$

- 4) Choose h , k , k_0 and β to satisfy the constraints (39)–(40), and $0 < h \leq h^*$;

- 5) Design the output SMC controller in the form of (36) by using the values of \bar{C} , k and h .

IV. SIMULATION EXAMPLE

In this section, a simulation example of a Magnetic Levitation System (MLS) is used. The following nonlinear model in [21] is considered:

$$\begin{cases} \dot{x}_1(t) = x_2(t) \\ \dot{x}_2(t) = -\frac{k}{M} x_2(t) + \frac{aL_0}{2M} \frac{x_3^2(t)}{(a+x_1(t))^2} - g \\ \dot{x}_3(t) = \frac{1}{L(x_1(t))} (-R_0 x_3(t) - aL_0 \frac{x_2(t)x_3(t)}{(a+x_1(t))^2} + v(t)) \end{cases} \quad (44)$$

where $x_1(t)$, x_2 and $x_3(t)$ are, respectively, the plate's position in [m], velocity in [m/s] and coil current in [A], and $v(t)$ is the control input (voltage applied to coil). Moreover, M is the mass of the plate, g is the gravity acceleration, k is a viscous friction coefficient, R_0 is the electric resistance, and $L(x_1(t)) = L_1 + \frac{aL_0}{a+x_1(t)}$ is the coil inductance, and a , L_0 and L_1 are positive constants. In the simulation setup, these parameters are given as $M = 0.1203\text{kg}$, $g = 9.815\text{m/s}^2$, $k = 0.01\text{N} \cdot \text{m/s}$, $L_1 = 0.1\text{H}$, $L_0 = 0.245\text{H}$, $a = 0.0088\text{m}$, and $R_0 = 1.75\Omega$. By diffeomorphism, the following variables are defined:

$$\begin{cases} \sigma_1(t) = x_1(t), \quad \sigma_2(t) = x_2(t), \\ \sigma_3(t) = -\frac{k}{M} x_2(t) + \frac{aL_0}{2M} \frac{x_3^2(t)}{(a+x_1(t))^2} - g. \end{cases}$$

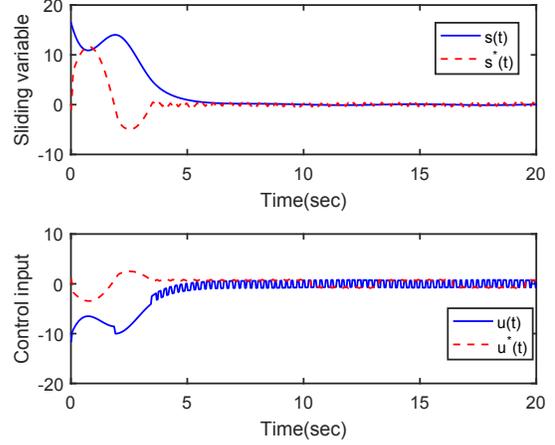


Fig. 1. Comparison of the proposed method with the full-state-information sliding mode controller ($n(t) = 0$).

Then, system (44) is equivalently represented as

$$\dot{\sigma}_1(t) = \sigma_2(t), \quad \dot{\sigma}_2(t) = \sigma_3(t), \quad \dot{\sigma}_3(t) = f(\sigma) + g(\sigma)v(t) \quad (45)$$

where $\sigma(t) = \text{col}\{\sigma_1(t), \sigma_2(t), \sigma_3(t)\}$, and

$$\begin{aligned} f(\sigma) &= \frac{k^2}{M^2} \sigma_1(t) + \frac{kg}{M} - 2 \left(\frac{R}{L(\sigma_1(t))} + \frac{k}{2M} + \frac{\sigma_2(t)}{a + \sigma_1(t)} \right. \\ &\quad \left. - \frac{aL_0 \sigma_2(t)}{L(\sigma_1(t))(a + \sigma_1(t))} \right) \cdot (\sigma_3(t) + \frac{k}{M} \sigma_2(t) + g), \\ g(\sigma) &= \frac{2aL_0(\sigma_3(t) + \frac{k}{M} \sigma_2(t) + g)}{M^{\frac{1}{2}} L(\sigma_1(t))(a + \sigma_1(t))}. \end{aligned}$$

Applying $v(t) = -g^{-1}(\sigma(t))[f(\sigma(t)) + u(t)]$ to system (46) yields the following perturbed triple integrator:

$$\dot{\sigma}_1(t) = \sigma_2(t), \quad \dot{\sigma}_2(t) = \sigma_3(t), \quad \dot{\sigma}_3(t) = u(t) + d(x, t) \quad (46)$$

where $v(t)$ is the virtual control input to be designed, and $d(x, t)$ is a perturbation caused by external signals or parameter or model uncertainties. The considered domain is defined as $|x(t)| \leq |x_1(t)| + |x_2(t)| + |x_3(t)| \leq \psi$, with $\psi = 20$. Then, $d(x, t) = 0.5 \cos(t) + 0.5 \sin(t)$, and $d^* = 1$. Only the noisy measurement of $\sigma_1(t)$ is available for control purpose, which is denoted as $f(t) = \sigma_1(t) + n(t)$, where $n(t)$ is the noises satisfying $|n(t)| \leq \varepsilon$, and ε is a positive scalar.

A. Design of output sliding mode controller via artificial time-delay estimation

We characterize the initial values as $x_1(0) = 3$, $x_2(0) = -3$, $x_3(0) = 3$. The design steps are listed:

- 1) Based on the interested domain, we select $k_1 = 13.2$ to satisfy that $|(\bar{A} - B_1 J_2^{-1} J_1)x| \leq |0.51| \cdot \psi \leq k_1$.

- 2) Choose $\bar{C} = [c_1 \quad c_2 \quad c_3]$ with $c_1 = 1.3$, $c_2 = 2.45$, $c_3 = 1$. Then, sliding variables $s^*(t)$ and $s(t)$ are $s^*(t) = 1.3x_1(t) + 2.45x_2(t) + x_3(t)$, $s(t) = \bar{C}_d \hat{x}(t, h)$,

where $\hat{x}(t, h) = \text{col}\{x_1(t), x_1(t-h), x_1(t-2h)\}$, $c_{d1} = c_1 + 1.5c_2/h + c_3/h^2$, and

$$\bar{C}_d = [c_{d1} \quad -2c_2/h - 2c_3/h^2 \quad 0.5c_2/h + c_3/h^2].$$

- 3) By solving LMIs of Theorem 1, the feasible solutions are obtained as $P_3 = 0.976$, $X = 0.568$, $W = 3.124$, and

$$P_1 = \begin{bmatrix} 0.876 & 0.622 \\ * & 1.456 \end{bmatrix}, \quad P_2 = \begin{bmatrix} -0.234 & 1.723 \end{bmatrix}.$$

Meanwhile, we find that $h^* = 0.87$.

- 4) Choose $k_0 = 0.5$, $k = 2$ and $\beta = 9$ such that the constraints (39) and (40) simultaneously hold.
 5) From (36), the output sliding mode controller via artificial time-delay estimation is written as

$$u(t) = -2\text{sgn}(s(t)) + g_1x_1(t) + \sum_{i=2}^3 g_i x_1(t - (i-1)h)$$

and $g_1 = (-10 - 10/h)$, $g_2 = 10/h + 10/h^2$, $g_3 = -5/h^2$, and the value of h should satisfy condition (35).

Inspired from [1], the control law for the case of the differentiator or sliding mode observer is written as

$$u^*(t) = -\frac{k_c}{2.7879}\text{sgn}(s^*(t)) - C_c \hat{x}(t)$$

where $k_c = 6$, $C_c = \begin{bmatrix} 0.7609 & 1.8913 & 1 \end{bmatrix}$, and $\hat{x}(t) = \text{col}\{f(t), \hat{x}_2(t), \hat{x}_3(t)\}$. States $x_2(t)$ and $x_3(t)$ are estimated via a robust exact differentiator/sliding mode observer. Fig.1 reveals the trajectories of $s^*(t)$, $s(t)$, $u(t)$ and $u^*(t)$, which shows that the artificial time-delay sliding surface can approximate the linear sliding surface with the acceptable precision.

B. Comparison of the artificial time-delay estimation with the robust differentiator/sliding mode observer

We use a robust exact differentiator (RED) [15]:

$$\begin{cases} \dot{\hat{x}}_1(t) = \hat{x}_2(t) - 3L_d^{\frac{1}{3}} e_d^{\frac{2}{3}}(t)\text{sgn}(e_d(t)) \\ \dot{\hat{x}}_2(t) = \hat{x}_3(t) - 1.5L_d^{\frac{1}{2}} e_d^{\frac{1}{2}}(t)\text{sgn}(e_d(t)) \\ \dot{\hat{x}}_3(t) = -1.1L_d \text{sgn}(e_d(t)) \end{cases}$$

where $e_d(t) = \hat{x}_1(t) - f(t)$, and $L_d \geq d^* + \sup|u|$ ([15]). Moreover, a higher order sliding mode observer (HOSMO) is used:

$$\begin{cases} \dot{\hat{z}}_1(t) = \hat{z}_2(t) + 3L_o^{\frac{1}{3}} e_o^{\frac{2}{3}}(t)\text{sgn}(e_o(t)) \\ \dot{\hat{z}}_2(t) = \hat{z}_3(t) + 1.5L_o^{\frac{1}{2}} e_o^{\frac{1}{2}}(t)\text{sgn}(e_o(t)) \\ \dot{\hat{z}}_3(t) = 1.1L_o \text{sgn}(e_o(t)) + u^*(t) \end{cases}$$

where $e_o(t) = \hat{z}_1(t) - f(t)$, and $L_o \geq d^*$ (see [22]). Here, we choose $L_d = 250$ ([15]), and $L_o = 50$, because HOSMO takes into account the known part of MLS dynamics ([22]). When $n(t) = 0$, Fig. 2 reveals that RED and achieve the better tracking precision of system dynamics than the artificial time-delay estimator without measurement noises. Apparently, RED/HOSMO have faster response rate than the proposed one, because RED and HOSMO adopt the dynamic output-feedback structure, while the artificial time-delay estimator is of static structure.

Next, sinuous measurement noises will be imposed in $x_1(t)$ for the the comparative simulations. They are assumed to be in the form of $n(t) = A_n \sin(f_n t)$, where A_n is the amplitude of noises and f_n is the noise frequency. Our method is very flexible, because h will be chosen with respect to

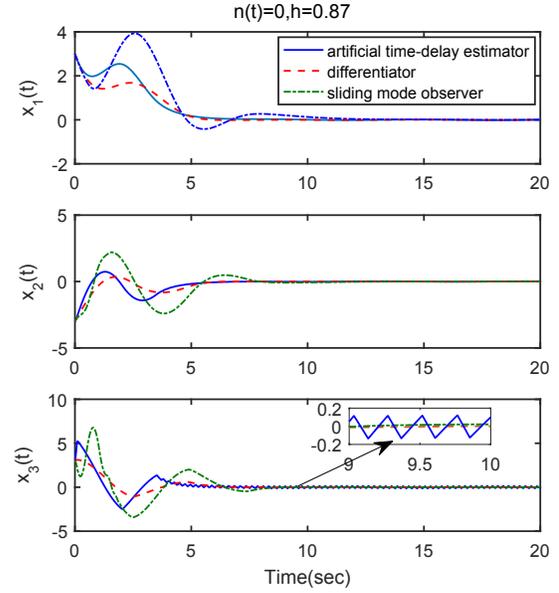


Fig. 2. Comparison of the proposed method with the RED and HOSMO ($n(t) = 0$).

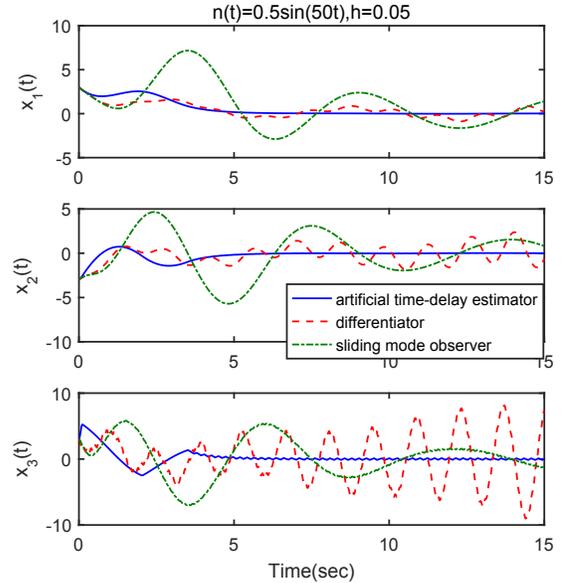


Fig. 3. Comparison of the proposed method with the RED and HOSMO ($n(t) = 0.5 \sin(50t)$).

f_n . Here, noises with different frequencies are, respectively, imposed on the measurement $x_1(t)$: $n(t) = 0.5 \cos(50t)$ and $n(t) = 0.1 \cos(1000t)$. For the selection of h , the trade-off will be made between the approximation accuracy and filtration against noises. According to condition (35), we choose $h = 0.05$ and $h = 0.01$, respectively, for $n(t) = 0.5 \cos(50t)$ and $n(t) = 0.5 \cos(1000t)$. Figs. 3–4 illustrate that the artificial time-delay estimator has certain filtering quantities against the measurement noises and can achieve a better robustness than the RED/HOSMO. Moreover, the proposed artificial time-delay estimator is of static output-feedback structure, which

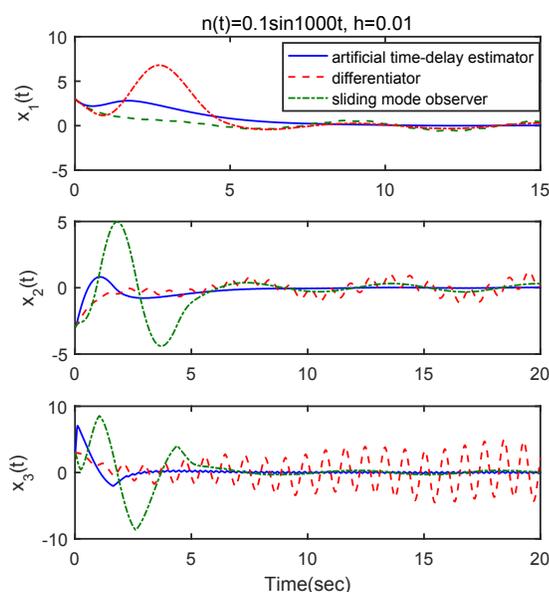


Fig. 4. Comparison of the proposed method with the RED and HOSMO ($n(t) = 0.1 \sin(1000t)$).

does not require introducing additional dynamics for state estimation.

V. CONCLUSION

In this paper, the static output-feedback sliding mode controller for systems with relative degree n has been introduced, where the output derivatives up to the order $n - 1$ are approximated by using the current and the delayed values of the output. First we design a delayed sliding surface, and then prove its finite-time attractivity. Lyapunov-Krasovskii functional is suggested to achieve the practical stabilization. The design parameters are chosen for a good trade-off between the approximation accuracy in the presence of measurement noises and the reduction of controller gain. A simulation example is given to show the merits of the proposed design method.

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