Integral Sliding Mode Control: Performance, Modification, and Improvement

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Abstract—Sliding mode control (SMC) is attractive for nonlinear systems due to its invariance for both parametric and nonparametric uncertainties. However, the invariance of SMC is not guaranteed in a reaching phase. Integral SMC (ISMC) eliminates the reaching phase such that the invariance is achieved in an entire system response. To reduce chattering in ISMC, it was suggested that the switching element is smoothed by using a low-pass filter and an integral sliding variable is modified. This study discusses several crucial problems regarding the performance, modification, and improvement of ISMC. First, the modification of the integral sliding variable is revealed to be unnecessary as it degrades the performance of a sliding phase; second, ISMC is shown to be a kind of global SMC; third, it is manifested that a high-order ISMC design with super twisting involves a stability condition that may be infeasible in theory; finally, an efficient solution is suggested to attenuate chattering in ISMC without the degradation of tracking accuracy and the solution is extended to the case with uncertain control gain functions. Comprehensive simulation results have verified the arguments of this study.

Index Terms—Chattering attenuation, disturbance rejection, integral sliding mode, high-order sliding mode, nonlinear system, uncertainty estimation.

I. INTRODUCTION

Sliding mode control (SMC) is a variable structure control technique which applies a switching control law to alter the plant dynamics such that the plant states slide along a cross section termed sliding surface of its normal behavior. SMC has attracted widespread applications in industrial informatics, e.g., see [1]–[9]. An SMC process usually has two phases, namely a reaching phase and a sliding phase [10]. In the reaching phase, the plant states are forced to reach a prespecified sliding surface in finite time. Once the plant states reach the sliding surface, the closed-loop system is adaptively altered to a sliding mode such that the plant states slide toward the origin along the sliding surface. This duration is called the sliding phase. In the sliding phase, the system response remains invariant for both parametric and nonparametric uncertainties [11]. However, during the reaching phase, the invariance of SMC is not guaranteed and the system response is sensitive to perturbations [12].

Integral SMC (ISMC) eliminates the reaching phase by enforcing the sliding mode in an entire system response so that the invariance of SMC is ensured from the initial time instant [13]. The imperfect implementation of high-frequency switching in SMC results in chattering at control responses [14]. To reduce chattering in ISMC, the switching element is smoothed by using a low-pass filter based on the equivalent control method, and an integral sliding variable is modified for facilitating stability analysis in [13]. The approach of [13] was applied to position control of robot manipulators and load pressure control of die-cushion cylinder drives in [15] and [16], respectively. A more popular way to attenuate chattering in ISMC is to integrate with high-order SMC [17]–[23]. In high-order ISMC, the switching element appears in a time derivative of the sliding variable such that the actual control law is smoothed by an integral [24].

In this study, several crucial issues regarding the performance and modification of ISMC are discussed and an efficient solution is finally suggested to improve the performance of ISMC. First, the modification of the integral sliding variable is demonstrated to be unnecessary; second, ISMC and global SMC (GSMC) are shown to be closely connected; third, the stability condition of a high-order ISMC design with super twisting is proven to be theoretically infeasible; finally, a simple and feasible solution is suggested to reduce chattering in ISMC without the degradation of tracking accuracy. Simulations are provided throughout the study to validate the arguments.

Throughout this paper, $\mathbb{R}$, $\mathbb{R}^+$, and $\mathbb{R}^n$ are the spaces of real numbers, positive real numbers, and real $n$-vectors, respectively, $\|x\|$ is the Euclidean norm of $x$, $\Omega_c := \{x|\|x\| \leq c\}$ is the ball of radius $c$, and $\text{sgn}(x)$ is the standard signum function, where $c \in \mathbb{R}^+$, $x \in \mathbb{R}$, $x \in \mathbb{R}^n$, and $n$ is a natural number. For the sake of brevity, in the following sections, the arguments of a function may be omitted while the context is sufficiently explicit.

II. INTEGRAL SMC

This section presents the ISMC design in [13]. For the simplification of illustration, consider a class of perturbed uncertain
affine-in-control nonlinear systems [18]

\[
\begin{align*}
\dot{x}_i &= x_{i+1} \quad (i = 1, 2, \ldots, n-1) \\
\dot{x}_n &= f(x) + g(x)u + d(t)
\end{align*}
\]  

(1)

where \(x(t) = [x_1(t), x_2(t), \ldots, x_n(t)]^T \in \mathbb{R}^n\) is a state vector, \(f : \mathbb{R}^n \rightarrow \mathbb{R}\) is a nonlinear drifting function, \(g : \mathbb{R}^n \rightarrow \mathbb{R}\) is a control gain function, \(u(t) \in \mathbb{R}\) is a control input, and \(d(t) \in \mathbb{R}\) presents a unknown perturbation caused by nonparametric uncertainties such as unmodeled dynamics and external disturbances. In addition, \(f\) and \(g\) are separated into

\[
\begin{align*}
f(x) &= f_0(x) + \Delta f(x) \\
g(x) &= g_0(x) + \Delta g(x)
\end{align*}
\]

(2)

(3)

where \(f_0, g_0 : \mathbb{R}^n \rightarrow \mathbb{R}\) are nominal parts, and \(\Delta f, \Delta g : \mathbb{R}^n \rightarrow \mathbb{R}\) are perturbed parts caused by nonparametric uncertainties such as inaccurate parameters and parameter variations.\(^1\) Let \(x_d(t) \in \mathbb{R}^n\) be a desired output. The following standard assumptions of SMC are introduced for the subsequent discussions.

**Assumption 1:** \(\Delta f\) are locally bounded and \(g \neq 0\), i.e., there exist some constants \(\tilde{f}, \tilde{g} \in \mathbb{R}^+\) such that \(|\Delta f| \leq \tilde{f}\), \(|\Delta g| \leq \tilde{g}\) and \(|g| > \tilde{g}\) \(\forall x \in \Omega_c \subset \mathbb{R}^n\) with \(c \in \mathbb{R}^+\). Without loss of generality, it is assumed that \(g > 0\).

**Assumption 2:** \(d\) is globally bounded, such that there exists a constant \(d \in \mathbb{R}^+\) to satisfy \(|d(t)| \leq d \forall t \geq 0\).

**Assumption 3:** \(x_d^{(i)}\) exist and are bounded for \(i = 0\) to \(n\).

For clear illustration, the case of \(\Delta g \neq 0\) is initially considered, and the extension to the case of \(\Delta g = 0\) will be presented in Section VI. Let \(x_d(t) := [x_d(t), x_d(t), \ldots, x_d^{(n-1)}(t)]^T\) and define the tracking error \(e(t) := x(t) - x_d(t) = e_1(t), e_2(t), \ldots, e_n(t)\)^\(T\), where \(e_1(t) := x_1(t) - x_d(t)\). Then, one gets the open-loop tracking error dynamics

\[
\begin{align*}
\dot{e}_i &= e_{i+1} \quad (i = 1, 2, \ldots, n-1) \\
\dot{e}_n &= [f_0(x) - x_d^{(n)}] + g_0(x)u + h(x, t)
\end{align*}
\]

(4)

with \(h : \mathbb{R}^n \times \mathbb{R}^+ \rightarrow \mathbb{R}\) a lumped perturbation given by

\[
h(x, t) := \Delta f(x) + d(t).
\]

(5)

The conventional sliding variable \(s_0 \in \mathbb{R}\) is given by \(s_0(e) = (d/dt + \lambda)^{-1}e_1\) [28], in which \(\lambda \in \mathbb{R}^+\) is a design parameter specifying the performance during the sliding phase.\(^2\) It follows from the definitions of \(e_1\) and \(e\) that

\[
s_0(e) = [\Lambda^T 1]e
\]

(6)

with \(\Lambda : = [\lambda^{(n-1)}, (n-1)\lambda^{(n-2)}, \ldots, (n-1)\lambda]^T\). Taking the time derivative of \(s_0\) in (6) and using (4), one obtains

\[
s_0 = [f_0(x) + v(x, t)] + g_0(x)u + h(x, t)
\]

(7)

where \(v(x, t) := 0 [\Lambda^T 1]e - x_d^{(n)}(t)\). For the reduced-order system (7), choose the following control law:

\[
u(t) = u_0(t) + u_1(t)
\]

(8)

where \(u_0(t) \in \mathbb{R}\) is a nominal control for (7), and \(u_1(t) \in \mathbb{R}\) is a discontinuous control for the rejection of \(h\). Letting \(h = 0\) and \(u = u_0\) in (7) and choosing

\[
u_0 = -[k_c s_0 + f_0(x) + u(x, t)]/g_0(x)
\]

(9)

leads to the ideal closed-loop dynamics

\[
s_0 + k_c s_0 = 0
\]

(10)

where \(k_c \in \mathbb{R}^+\) is a control gain parameter.

The following ISMC design follows [13]. A systematic description of the method of [13] can be found in [14]. According to the equivalent control method [25, Sec. 2], an integral sliding variable is designed as follows:

\[
s(t) = s_0(e) + z(t)
\]

(11)

where \(z(t) \in \mathbb{R}\) is an integral term given by

\[
\dot{z} = -\frac{\partial s_0}{\partial e} \left[ f_0(x) - x_d^{(n)} + g_0(x)u_0 \right]
\]

(12)

with \(\bar{e}_1 := [e_2, e_3, \ldots, e_n]^T\) and \(z(0) = -s_0(e(0))\) such that \(s(0) = 0\). Applying the expressions of \(s_0\) in (6), \(u_0\) in (9) and \(v\) under (7) to (12), one obtains

\[
\dot{z} = -[\Lambda^T 1] \left[ -k_c s_0 - 0 \Lambda^T e \right] = k_c s_0.
\]

Thus, the integral sliding variable \(s\) in (11) becomes\(^3\)

\[
s(t) = s_0(e(t)) - s_0(e(0)) + k_c \int_0^t s_0(e(\tau)) d\tau.
\]

(13)

Taking the time derivative of \(s\) in (13) and using (7) yields

\[
\dot{s} = [f_0(x) + v(x, t)] + g_0(x)u + h(x, t) + k_c s_0.
\]

Applying (8) with (9) to the above result, one obtains

\[
s = g_0(x)u_1 + h(x, t).
\]

(14)

Now, choose the discontinuous control

\[
u_1 = -\alpha sgn(s), \quad \alpha \geq (\tilde{f} + \tilde{d} + \eta)/g_0
\]

(15)

in which \(\eta \in \mathbb{R}^+\) is a constant specifying the converging rate of \(s\). Substituting (15) into (14), one obtains

\[
s \dot{s} = s(h(x, t) - g_0(x)\alpha sgn(s))
\]

\[
\leq s h(x, t) - (\tilde{f} + \tilde{d} + \eta)|s|
\]

\[
\leq (|h(x, t)| - \tilde{f} - \tilde{d} - \eta)|s|.
\]

Applying (5) with Assumptions 1 and 2 to the above result, one gets a standard reaching law

\[
s \dot{s} \leq -\eta |s| \quad \forall x \in \Omega_c
\]

(16)

implying \(s(t) \rightarrow 0\) in finite time. The setting \(s(0) = 0\) removes the reaching time resulting in \(s(t) = 0 \forall t \geq 0\). Once the plant

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\(^1\)The extension to the case that \(\Delta f\) and \(\Delta g\) explicitly depend on time \(t\), i.e., \(\Delta f(x, t)\) and \(\Delta g(x, t)\), can be referred to [27, Remark 1].

\(^2\)The mapping \(s_0 : \mathbb{R}^n \rightarrow \mathbb{R}\) is also called a switching function [14].

\(^3\)A so-called total SMC of [29] is exactly the same as the ISMC of [13] where they share the same sliding variable \(s\) given by (13).
states are confined to the sliding manifold \( s = 0 \), the equivalent control \( u_{1eq} \) is obtained by setting \( s = 0 \) in (14) as follows:

\[
u_{1eq} = -h(x, t)/g_0(x).
\] (17)

Applying \( u = u_0 + u_{1eq} \) with \( u_0 \) in (9) and \( u_{1eq} \) in (17) to (7), one gets a motion equation in the sliding mode identical to the ideal dynamics (10) so that \( s_0 \to 0 \) exponentially implying \( e \to 0 \) exponentially. Therefore, the nominal control \( u_0 \) achieves the ideal dynamics (10) as if the perturbation \( h \) in (7) does not exist. Let \( h(t) \in \mathbb{R} \) be an estimate of \( h \) given by

\[
\hat{h} = -g_0(x)u_1.
\] (18)

To illustrate the performance of the above ISMC, consider a simplified model of underwater vehicles [28, Sec. 7]

\[
\begin{aligned}
\dot{x}_1 &= x_2 \\
\dot{x}_2 &= -cx_2/m + (1/m)u + d(t)
\end{aligned}
\] (19)

where \( x_1 \) (m) is the vehicle position, \( u \) (N) is the control force, \( m \) (kg) is the vehicle mass, and \( c \) is a drag coefficient. Note that \( m \) and \( c \) are not exactly known in practice. For simulations, let \( x(0) = [-1, 0]^T \), \( m = m_0, c = c_0 + 0.2 \sin(x_2(t), \text{ and } d(t) = 3 \sin(\pi t), \text{ where } m_0 = 3 \text{ and } c_0 = 1.2 \text{ are nominal values of } m \text{ and } c, \text{ respectively. The desired trajectory is comprised of an acceleration } \ddot{x}_d = 2 \text{ m/s}^2 \text{ at } t \in [0, 2] \text{ s, a velocity } \dot{x}_d = 4 \text{ m/s at } t \in [2, 4] \text{ s, and an acceleration } \ddot{x}_d = -2 \text{ m/s}^2 \text{ at } t \in [4, 6] \text{ s} [28, \text{ Sec. 7}]. \text{ To construct the ISMC in (8), set } f_0 = 1.2 x_2 |x_2|/3, g_0 = 1/3, \text{ and } k_u = 5 \text{ for } u_0 \text{ in (9), } \lambda = 5 \text{ and } s_0(e(0)) = [\lambda, 1]e(0) = -5 \text{ for } s \text{ in (13), and } \alpha = 13 \text{ for } u_1 \text{ in (15).}

Simulations are carried out in MATLAB software, where the solver is set to be fixed-step ODE 4, the step size is set to be 0.1 ms, and the other settings are kept at their defaults. Simulation trajectories are given in Fig. 1, where the state vector \( x \) exactly tracks the desired output \( x_d \) under a control input \( u \) with serious chattering after a short transient process, the estimate \( h \) does not follow the perturbation \( h \), the sliding variable \( s \) keeps very near to 0 from \( t = 0 \), but its time derivative \( \dot{s} \) exhibits chattering with low amplitudes due to imperfect switches caused by the digital simulation with a sampling time.

**Remark 1:** The system (1) under the ISMC law \( u = u_0 + u_1 \) in (8) with \( u_0 \) in (9), \( u_1 \) in (15), and \( s \) in (13) obtains exponential convergence of the tracking error \( e \) to 0, in which the reaching phase is eliminated resulting in the invariance throughout an entire system response. However, the above ISMC design includes two major limitations: 1) Chattering of the conventional SMC is inherited resulting in difficulty for practical applications; and 2) the estimation of the perturbation \( h \) by \( \hat{h} \) in (18) is not realizable due to the intrinsic discontinuity of \( u_1 \).

**Remark 2:** Like most existing SMC results, we assume that the state vector \( x \) in the system (1) is fully measurable. If only \( x_1 \) is measurable, \( x_2 \) to \( x_n \) can be exactly estimated by sliding mode observers (SMOs). For instance, the following SMO can be applied to the system (19) [10, Sec. 1.6]:

\[
\begin{aligned}
\dot{\hat{x}}_1 &= -\beta \text{sgn}(z_1) \\
\dot{\hat{x}}_2 &= \varphi(\hat{x}_2, u) - \hat{x}_2 - \beta \text{sgn}(z_1)
\end{aligned}
\] (20)

in which \( \varphi(\hat{x}_2, u) := -c_0 \hat{x}_2 |\hat{x}_2|/m + (1/m_0)u \) is the nominal part of the system (19), \( \hat{x}_1 \in \mathbb{R} \) and \( \hat{x}_2 \in \mathbb{R} \) denote estimates of \( x_1 \) and \( x_2 \), respectively, \( z_1 := \hat{x}_1 - x_1 \) is an estimation error, \( \beta \in \mathbb{R}^+ \) is a switching gain, and \( \tau \in \mathbb{R}^+ \) is a filtering constant. Control results under the SMO (20) with \( \beta = 4.5 \) and \( \tau = 0.01 \) are depicted in Fig. 2. It is shown that the SMO (20) provides a favorable estimation of \( x_2 \) and the tracking accuracy of the SMO-based ISMC is just slightly degraded compared with that of the original ISMC. Please refer to [10] for more about SMOs as they are out of the scope of this discussion.

### III. INTEGRAL SLIDING SURFACE MODIFICATION

To reduce chattering in the ISMC design of Section II, it was suggested in [14, Sec. 7] that the discontinuous control \( u_1 \) given by (15) is low-pass filtered by a first-order linear filter

\[
\mu \dot{u}_{1av}(t) = -u_{1av}(t) + u_1(t)
\] (21)

with \( u_{1av}(0) = 0 \), where \( u_{1av}(t) \in \mathbb{R} \) is an averaging counterpart of \( u_1(t) \), and \( \mu \in (0, 1) \) is a filtering constant that should be small enough to avoid disturbing the slow component of the switching action in \( u_1 \) equal to the equivalent control \( u_{1eq} \) given
Combining (17) with \( u_{1eq} = u_{1av} \), one gets
\[
 u_{1av} = -h(x, t)/g_0(x) \tag{25}
\]
such that the estimate of \( h \) is given by
\[
 \hat{h} = -g_0(x)u_{1av}. \tag{26}
\]

However, due to the intrinsic discontinuity of \( u_1 \) in (15), the ideal result \( \dot{s} = 0 \ \forall t \geq 0 \) is impossible to be obtained from (14) in practice. Without \( \dot{s} = 0 \), (17) is not obtainable such that (25) is also not obtainable, which implies the exact estimation of \( h \) by \( \hat{h} \) in (26) is not realizable. Even if one has \( \dot{s} = 0 \ \forall t \geq 0 \), setting \( \dot{s} = 0 \) in (24) yields the sliding mode dynamics
\[
 s_0 + k_c s_0 + g_0(x)(u_1 - u_{1av}) = 0 \tag{27}
\]
which implies \( s_0 \) converges to a neighborhood of 0 subject to \( g_0, \mu, \) and \( k_c \). Therefore, the tracking error \( e \) only converges to a compact set \( \Omega_c \) subject to \( g_0, \mu, k_c, \) and \( \lambda \), where the size of \( \Omega_c \) is not necessarily small since high-frequency switches in \( u_1 \) may result in a large discrepancy between \( u_1 \) and \( u_{1av} \) even if the filtering constant \( \mu \) in (21) is small.

To demonstrate the modified ISMC in this section, consider the same illustrative example with the same settings as that in Section II. The construction of the control law follows the steps in Section II except the filtering constant \( \mu \) in (21) is set to be 0.1. Simulation trajectories are shown in Fig. 3, where the state vector \( x \) follows its desired signal \( x_d \) under a smooth control input \( u \). However, although the sliding variable \( s \) still keeps near to 0 from \( t = 0 \), its time derivative \( \dot{s} \) demonstrates serious chattering resulting in a large phase lag in the estimation of \( h \) even the setting of \( \mu = 0.1 \) is small. These results are consistent with the results of [14, Fig. 8.18]. Further decreasing \( \mu \) is not suggested as it may introduce serious chattering at \( u \).

Remark 3: For the system (1) driven by the modified ISMC law \( u = u_0 + u_{1av} \) given by (22) with \( u_0 \) in (9), \( u_1 \) in (15), \( u_{1av} \) in (21), and \( s \) in (24), the performance of the reaching phase is identical to (16) by the original ISMC in Section II. Yet, the exact estimation of \( h \) in (5) is not realizable, and the performance of the sliding phase is degraded resulting in a compromise among chattering, tracking accuracy, and robustness, which are not consistent with the claim “The concept of integral sliding mode can also be extended to construct a new type of perturbation estimator that solves the chattering problem without loss of robustness and control accuracy.” in [14, Sec. 7].

IV. GLOBAL SMC

GSMC originated in [12] also aims to eliminate the reaching phase such that the sliding mode invariably exists and the invariance is guaranteed in an entire system response. In the GSMC, the integral term \( z \) of the sliding variable \( s \) in (11) is designed to satisfy the following conditions [30]: 1) \( z(0) = -s_0(e(0)) \); 2) \( z(t) \to 0 \) as \( t \to \infty \); 3) \( z(t) \) exists and is bounded. In the ISMC, if \( z \) is chosen as follows:
\[
 \dot{z} = -k_c z, \quad z(0) = -s_0(e(0)) \tag{28}
\]
then the sliding variable \( s \) becomes
\[
 s(t) = s_0(e(t)) - s_0(e(0))e^{-k_c t} \tag{29}
\]
which satisfies all the aforementioned conditions such that the resultant SMC belongs to a kind of GSIC.

Taking the time derivative of \( s \) in (29) and using (7) yields

\[
\dot{s} = [f_0(x) + v(x, t)] + g_0(x)u + h(x, t) - k_c z.
\]  

(30)

The control law is chosen as \( u = u_0 + u_1 \) in (8) with the nominal control \( u_0 \) in (9) and the discontinuous control \( u_1 \) in (15). Applying \( u_0 \) in (9) to (30), one obtains

\[
\dot{s} = -k_c s + g_0(x)u_1 + h(x, t).
\]  

(31)

Applying \( u_1 \) in (15) to (31), one obtains

\[
\dot{s} = -k_c s^2 + s\left(h(x, t) - g_0(x)\left|\text{sgn}(s) + \dot{\vartheta}\right|\right).
\]

(32)

Noting (5) with Assumptions 1 and 2, the above result leads to an exponential reaching law [31]

\[
\dot{s} \leq -k_c s^2 - \eta|s| \quad \forall x \in \Omega_c,
\]

(32)

which is different from the standard reaching law (16). Combining (32) with \( s(0) = 0 \), one obtains \( s(t) = 0 \) \( \forall t \geq 0 \). Setting \( s = 0 \) in (29), one obtains

\[
s_0(e(t)) = s_0(e(0))e^{-k_c t}
\]

(33)

which is exactly the solution of the ideal dynamics (10). Therefore, the performance of the GSIC at the sliding mode is the same as that of the ISMC in Section II.

**Remark 4:** The system (1) driven by the GSIC law \( u = u_0 + u_1 \) in (8) with \( u_0 \) in (9), \( u_1 \) in (15) and \( s \) in (29) ensures exponential convergence of the tracking error \( e \) to 0 during the reaching phase, which is the same as that by the original ISMC in Section II. Due to the exponential reaching law (32), the GSIC provides extra robustness against unanticipated perturbations compared with the ISMC in Section II [31].

### V. HIGH-ORDER INTEGRAL SMC

A more popular way of chattering attenuation in ISMC is to integrate with high-order ISMC [10, Sec. 4]. Existing high-order ISMC methods require the information of \( \dot{x}_e \) except the method of [23] with super twisting, where the perturbation \( h \) does not depend on the state vector \( x \) in [23]. It is worth noting that \( \dot{x}_e \) is usually not accessible for measurement so that the requirement on \( \dot{x}_e \) is not desirable in practice. The Lyapunov stability of the high-order SMC with super twisting was established in [33]. In this section, it is demonstrated that the high-order ISMC with super twisting [23] leads to a stability condition that is usually infeasible in theory if \( h \) depends on \( x \).

Recall the reduced-order system (7) driven by the control law \( u = u_0 + u_1 \) in (8) with \( u_0 \) in (9). The following additional assumptions are required to proceed the control design.

**Assumption 4:** \( \Delta f \) and \( \Delta g \) are continuously differentiable.

**Assumption 5:** \( \dot{d} \) is globally bounded, i.e., there is a constant \( \bar{d}_d \in \mathbb{R}^+ \) such that \( \bar{d}_d \leq \bar{d}_d \forall t \geq 0 \).

The discontinuous control \( u_1 \) in (8) of Section II is replaced by a super-twisting algorithm [23]

\[
\begin{align*}
  u_1 &= (-k_c s^2 + s(h(x, t) - g_0(x)))/g_0(x) \\
  \dot{\vartheta} &= -\alpha \text{sgn}(s)
\end{align*}
\]  

(34)

with \( \vartheta(0) = 0, k_c = 1.5\sqrt{\bar{d}_d} \) and \( \alpha = 1.1\bar{d}_d \), where \( s \) is given by (13), \( \vartheta \in \mathbb{R} \) is an auxiliary variable, and \( \bar{d}_d \in \mathbb{R}^+ \) is a upper bound of \( h \) that satisfies \( |h(x, t)| \leq \bar{d}_d, \forall x \in \Omega_c \). Applying (34) to (14) yields the closed-loop dynamics

\[
\begin{align*}
  s &= -k_c s^2 + \vartheta + h(x, t) \\
  \dot{\vartheta} &= -\alpha \text{sgn}(s)
\end{align*}
\]  

(35)

It follows from [23] that (35) guarantees \( s(t) = 0 \) \( \forall t \geq 0 \) and \( \vartheta(t) = -h(x(t), t) \) after a finite time. Thus, the tracking error \( e \) exponentially converges to 0 and the estimate \( \hat{h} := -\vartheta \) exactly follows the perturbation \( h \) after a finite time.

The intention of introducing Assumptions 4 and 5 is to ensure the existence of the upper bound \( \bar{h}_d \). Yet, \( \bar{h}_d \) still may not exist under the additional two assumptions. To clarify this claim, the time derivative of \( h \) given by (5) is derived as follows:

\[
\dot{h}(x, t) = \frac{\partial \Delta f}{\partial x} \left( f(x) + g(x)u + d(t) \right) + \dot{d}(t)
\]

(33)
with $\mathbf{x}_1 := [x_2, x_3, \ldots, x_n]^T$. Thus, one obtains

$$h(x, t) = f_a(x) + f_b(x)[f(x) + g(x)u + d(t)] + d(t).$$

where $f_a(x) := \sum_{i=1}^{n-1} \frac{\partial f}{\partial x_i} x_{i+1}$ and $f_b(x) := \frac{\partial f}{\partial x_n}$. Applying $u = u_0 + u_1$ in (8) to the above result leads to

$$\dot{h}(x, t) = f_a(x) + f_b(x)(f(x) + g(x)u_0 + d(t)) + d(t) + f_b(x)g(x)u_1 \leq \tilde{h}_d.$$

(36)

Intuitively, it follows from (36) that $\tilde{h}_d$ is hard to be determined as $u_1$ depends on $\dot{h}_d$ via $k_s$ and $\alpha$ in (34) so that a large $\tilde{h}_d$ also leads to a large amplitude at the left side of “$\leq$” in (36). The detailed analysis is given as follows.

Noting Assumptions 1 and 4, let $c_a := \max_{x \in \Omega_{r_0}} \{|f_a(x)|\} \in \mathbb{R}^+$, $c_b := \max_{x \in \Omega_{r_0}} \{|f_b(x)|\} \in \mathbb{R}^+$ and $c_{u_0} := \max_{x \in \Omega_{r_0}} \{|g_0(x)u_0|\} \in \mathbb{R}^+$. Applying the above definitions to (36) and noting Assumptions 1, 2, and 5 and $g(x) = g_0(x)$, one gets

$$h(x, t) \leq c_a + c_b(\bar{f} + d + c_{u_0}) + d_d + f_b(x)g_0(x)u_1.$$

Applying $k_s = 1.5\sqrt{\tilde{h}_d}$ and $\alpha = 1.1\tilde{h}_d$ to (34), one obtains

$$g_0(x)u_1 = -1.5\sqrt{\tilde{h}_d}|s|\text{sgn}(s) - 1.1\tilde{h}_d \int_0^t \text{sgn}(s(\tau))d\tau.$$

Combining with the above two results, one obtains

$$\dot{h}(x, t) \leq c_a + c_b(\bar{f} + d + c_{u_0}) + d_d + 1.5c_b\sqrt{\tilde{h}_d}|s| - 1.1\tilde{h}_d f_b(x) \int_0^t \text{sgn}(s(\tau))d\tau$$

where the upper bound at the right size of “$\leq$” is not conservative as it is the frequent case that upper bounds are positive.

The above inequality implies the selection of $\tilde{h}_d$ has to satisfy

$$\tilde{h}_d \left(1 + 1.1f_b(x) \int_0^t \text{sgn}(s(\tau))d\tau\right) \geq c_a + c_b(\bar{f} + d + c_{u_0}) + d_d + 1.5c_b\sqrt{\tilde{h}_d}|s|.$$

As all items at the right side of the above inequality are positive, a necessary condition for the existence of $\tilde{h}_d$ is

$$f_b(x) \int_0^t \text{sgn}(s(\tau))d\tau \geq -1/1.1$$

(37)

which is very stringent since it requires the left of “$\leq$” always being equal to or larger than $-1/1.1$. At an extreme case that (37) is satisfied, let $c_f := \min_{x \in \Omega_{r_0}} \{1 + 1.1f_b(x) \int_0^t \text{sgn}(s(\tau))d\tau\} \in \mathbb{R}^+$, $c_f := \max_{x \in \Omega_{r_0}} (1.5c_b\sqrt{|s|}) \in \mathbb{R}^+$ and $c_h := c_a + c_b(\bar{f} + d + c_{u_0}) + d_d$. Then, one obtains

$$\left(\sqrt{\tilde{h}_d}\right)^2 - (c_g/c_f)\sqrt{\tilde{h}_d} \geq c_h/c_f.$$ 

Completing the square at the left side of the above inequality, one gets its equivalent expression

$$\left(\sqrt{\tilde{h}_d} - c_g/(2c_f)\right)^2 \geq c_h/c_f + c_g^2/(2c_f)^2$$

which is also equivalent to

$$\tilde{h}_d \geq \left[\sqrt{c_h/c_f + c_g^2/(2c_f)^2} + c_g/(2c_f)\right]^2$$

(38)

where the right side of “$\geq$” is usually extremely larger than $(\bar{f} + d)$, the upper bound of $h(x, t)$ for the ISMC in Section II.

To illustrate the high-order ISMC with super twisting in this section, consider the same illustrative example with the same settings as that in Section II. The construction of the control law follows the steps in Section II except the upper bound $\tilde{h}_d$ in (34) is set as 20. Simulation trajectories are depicted in Fig. 4. Despite the limitations discussed in [26], the super-twisting high-order ISMC with the larger switching gain $\alpha$ achieves high tracking and estimation accuracy under a smooth control input $u$, which is not consistent with the above analysis. The chattering at $\dot{s}$ is similar to that of the original ISMC in Section II.

Remark 5: For the system (1) under the super-twisting high-order ISMC law $u = u_0 + u_1$ in (8) with $u_0$ in (9), $u_1$ in (34), and $s$ in (13), exponential convergence of the tracking error $e$ and an exact estimation of the perturbation $h$ are shown under the stability condition (37) with (38) that is usually infeasible in theory. However, the superior tracking and estimation performances shown in Fig. 4 differ from the theoretical analysis, which implies the selection of the gains $k_s$ and $\alpha$ in (34) may
be improper and further studies based on advanced high-order SMC techniques such as [32] is still desirable.

VI. SUGGESTED CHATTERING REDUCTION SCHEME

To resolve the chattering problem, we suggest to utilize the filtering solution in Section III without the integral sliding variable modification in (23), i.e., we use the control law

\[ u(t) = u_0(t) + u_{1av}(t) \]  

(39)

where the nominal control \( u_0 \) is given by (9), and the averaged control \( u_{1av} \) is generated by (21) with the discontinuous control \( u_1 \) in (15) as its input. The sliding variable \( s \) is given by (29), corresponding to the suggested GSMC design. This suggestion is well supported by the equivalent control principle [14, Sec. 7.4]; the equivalence of \( u_1 \) is equal to \( u_{1av} \) as long as the filtering bandwidth of (18) covers the frequency spectrum of the perturbation \( h \). Substituting (39) with (9) to (30), one obtains

\[ \dot{s} = -k_c s + g(x)u_{1av} + h(x, t) \]  

(40)

which differs from (31). Noting the continuity of \( u_{1av} \) and \( s(t) = 0 \; \forall t \geq 0 \), the ideal result \( \dot{s} = 0 \; \forall t \geq 0 \) is possible from (40) in practice. Setting \( \dot{s} = s = 0 \) in (40) yields [14, Sec. 7]

\[ u_{1eq} = u_{1av} = -h(x, t)/g_0(x). \]  

(41)

Hence, if the filtering constant \( \mu \) in (21) is chosen to be small enough such that the filtering bandwidth of (21) covers the frequency spectrum of \( h \), \( \hat{h} \) in (26) provides an exact estimation of \( h \) according to (41). In Section II, it is unreasonable to let \( u_1 = u_{1eq} \) as \( u_1 \) includes high-frequency contents that may not present in \( u_{1eq} \). Instead, it is more reasonable to have \( u_{1av} = u_{1eq} \) as high frequency contents are filtered out in \( u_{1av} \). Consequently, it is reasonable to expect \( u_{1av} \) performs similarly to \( u_1 \) with respect to tracking accuracy under reduced chattering.

In the above discussions, we assume \( \Delta g(x) = 0 \) for clear presentation. Now, the suggested GSMC design is extended to the case with \( \Delta g(x) \neq 0 \). Let \( \Delta g(x) \neq 0 \) in (1) such that the reduced-order system (7) becomes

\[ \dot{s}_0 = [f_0(x) + v(x, t)] + \Delta f(x) + d(t) \]  

\[ + [g_0(x) + \Delta g(x)]u. \]  

(42)

Applying the control law (8) with (9) to (42), one obtains

\[ \dot{s}_0 = -k_c s_0 + g(x)(\hat{h}(x, t) + u_1) \]  

(43)

where the perturbation \( h \) is redefined by

\[ h(x, t) := [\Delta f(x) + \Delta g(x)u_0 + d(t)]/g(x). \]  

(44)

Taking the time derivative of \( s \) in (29) along (43) yields

\[ \ddot{s} = -k_c s + g(x)(\dot{h}(x, t) + u_1). \]  

(45)

Choose the discontinuous control

\[ u_1 = -\text{sgn}(s), \; \alpha \geq (f + \bar{g}u_0 + \bar{d} + \eta)/g \]  

(46)

with \( \bar{u}_0 := \max_{x \in \Omega_1} \{u_0\} \). Applying (46) to (45) yields

\[ \dddot{s} = -k_c s_1^2 + g(x)(\dot{h}(x, t) - \text{sgn}(s)) \]  

\[ \leq -k_c s_1^2 + g(x)((\dot{h}(x, t) - (f + \bar{g}u_0 + \bar{d} + \eta)/g)|s|. \]

Noting (44) and Assumptions 1 and 2, one obtains the exponential reaching law (32). Following the same derivations as the case with \( \Delta g(x) = 0 \), one gets \( e \to 0 \) exponentially. To tackle the chattering problem, the GSMC law is chosen as (39), where \( u_0 \) is given by (9), \( u_{1av} \) is generated by (21) with \( u_1 \) in (46) as its input, and \( s \) is given by (29). Then, one obtains

\[ \dot{s} = -k_c s + g(x)(\dot{h}(x, t) + u_{1av}) \]  

(47)

so that an exact estimation of \( h \) in (44) is obtained by

\[ \hat{h} = -u_{1av} \]  

(48)

as in the case of \( \Delta g(x) = 0 \). Note that to handle the uncertainty \( \Delta g(x) \) in the control gain function \( g(x) \), the switching gain \( \alpha \) in (46) is designed to be much larger than that of the case with an exactly known \( g(x) \) in (15).

To illustrate the suggested GSMC under the case of \( \Delta g(x) = 0 \), consider the same illustrative example with the same settings and controller construction as that in Section II. Simulation trajectories by the suggested GSMC are depicted in Fig. 5, where an exact estimation of \( h \) in (5) by \( \hat{h} \) in (26) is achieved without
any phase lag, an exact tracking of $x_d$ by $x$ is obtained under a smooth control input $u$ without the tradeoff between chattering reduction and tracking accuracy, and the chattering at $\dot{s}$ is reduced to be much lighter than that by the super-twisting high-order ISMC in Fig. 4. A tracking error comparison of different ISMC designs is presented in Fig. 6, where the original ISMC in Section II exhibits chattering due to the intrinsic discontinuity of the control input $u$, the modified ISMC in Section III gets the worst tracking accuracy due to the imperfect estimation of $h$ [see Fig. 3(a)], and the suggested GSMC achieves the highest tracking accuracy without chattering.

To validate the superiority of the suggested GSMC, consider the same illustrative example with the same settings as that in Section II except the amplitude of the external disturbance $d$ in (1) is increased from 3 to 5 after $t = 2$ s. The suggested ISMC is obtained by applying $s$ given by (13) to replace $s$ given by (29) in the suggested GSMC. A tracking error comparison between the suggested ISMC and GSMC designs is presented in Fig. 7, where the exact estimation of $h$ in (5) by $\hat{h}$ in (26) loses after $t = 2$ s for both the controllers due to the unanticipated large $d$, and the suggested GSMC exhibits better robustness against $d$ reflected by higher tracking accuracy after $t = 2$ s.

To illustrate the suggested GSMC under the case of $\Delta q(x) \neq 0$, consider the same illustrative example with the same settings as Section II except $m = m_0 + 1.5 \sin(|x_2|t)$ implying $\Delta f(x) = (1.2 + 0.2 \sin(|x_2|t)x_2|x_2|/(3 + 1.5 \sin(|x_2|t))$ and $\Delta g(x) = 1/(3 + 1.5 \sin(|x_2|t)) - 1/3$. The construction of the control law is the same as that of the case with $\Delta g(x) = 0$ except the switching gain $\alpha$ is increased from 13 to 19. With the definitions of $h$ in (44), $u_0$ in (9), and $v$ under (7), one gets that the piecewise continuity of $\dot{\bar{x}}_g$ results in two piecewise points of $h$ at $t = 2$ and 4 s. Simulation trajectories of this case depicted in Fig. 8 are very similar to those in Fig. 5 except light oscillations exist in $\hat{h}$ and $\dot{s}$ near $t = 2$ and 4 s due to the piecewise points of $h$ at $t = 2$ and 4 s.

Remark 6: The suggested ISMC/GSMC design significantly alleviates the compromise among chattering, tracking accuracy, and robustness in SMC. Compared with the high-order ISMC in Section V, the distinctive feature of the suggested design is that the perturbation $h$ in (5)/(44) is exactly estimated by a much simpler control scheme. One major deficiency of the suggested design is that the setting of the filtering constant $\mu$ in (21) is subject to the frequency spectrum of $h$ such that chattering may not be well attenuated when $\mu$ needs to be very small. Note that the transient occurs if $h(0) \neq 0$ as $u_{1av}(0) = 0$ in (21).

Fig. 6. Tracking error comparison of various ISMC designs.

Fig. 7. Tracking error comparison of suggested control designs.

Fig. 8. Simulation trajectories by the suggested GSMC with $\Delta g(x) \neq 0$. (a) Control and estimation performances. (b) Evaluations of $s$ and $\dot{s}$. 

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VII. CONCLUSION

This study has discussed several crucial problems in ISMC, where conclusions are drawn as follows:

1) The modification of the integral sliding variable degrades the performance of the sliding mode.
2) ISMC belongs to a kind of GSNC.
3) The super-twisting high-order ISMC usually results in a stability condition that is infeasible in theory.

The low-pass filtering solution without the modification of the integral sliding variable is suggested for chattering attenuation in ISMC. Comprehensive simulations have validated the arguments of study and have demonstrated a superior performance of the suggested ISMC design in terms of chattering attenuation, trajectory tracking, and disturbance estimation. The application of the suggested ISMC to handle perturbations in composite learning control [34] would be interesting for further study.

REFERENCES


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