Control of Z-axis MEMS Gyroscope Using Adaptive Fractional Order Dynamic Sliding Mode Approach

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ABSTRACT: This paper proposes a states feedback control method for Z-axis MEMS gyroscopes using fractional calculus and adaptive dynamic sliding mode control method. A new sliding mode control method is proposed to achieve trajectory tracking by adding a fractional order term in the conventional sliding manifold. The new proposed sliding surface contains integer order terms as well as fractional order terms and thus can provide an extra degree of freedom. Besides, in the presence of unknown system parameters, some adaptive laws containing the new designed sliding manifold are proposed to online tune controller parameters. All adaptive laws are derived in the stability framework and the stability of the control system is also guaranteed according to the Lyapunov stability theory and. Simulations results on a Z-axis vibrating gyroscope are provided to illustrate the effectiveness of the control method.

INDEX TERMS: adaptive control, dynamic sliding mode control, gyroscope, fractional order calculus

I. INTRODUCTION

MEMS gyroscopes are commonly used sensors for measuring angular velocity which are widely used in many occasions such as cell phone, navigation, quadcopter and so on. The working principle of MEMS gyroscopes is based on the inertia effect of the detecting mass caused by the Coriolis force. Due to the defect of manufacturing technology, the structure of the MEMS gyroscope is not totally symmetric, and the asymmetric will cause quadrature coupling between the driving axis and the sensing axis. Besides, there are also parameter variation and external disturbance in the gyroscope system. All the above-mentioned factors will deteriorate the performance of the MEMS gyroscope. To address these problems, great efforts have been dedicated in the investigation of MEMS gyroscope control methods. Various control methods including sliding mode control, adaptive control and other techniques are all applied in the control of MEMS gyroscopes [1-3].

Sliding mode control (SMC) is a powerful robust control scheme which has been widely used in many areas [4-6]. SMC has many attractive features such as easy implementation, insensitive to parameter variation, robust to disturbance and so on and these advantages make SMC a useful tool in the control of system dynamics. SMC has many advantages while it also has its drawbacks. It can be found from the structure of sliding mode control that there are always discontinuous robust terms in the control force. The discontinuous terms will cause chattering in control forces and the high frequency chattering will damage actuators in practical situations. How to achieve satisfactory control performance while reducing chattering phenomenon is always a hot topic since SMC was firstly introduced [4]. Dynamic sliding mode control method is one kind of high order sliding mode control method that can alleviate chattering problem by transferring discontinuous term to the derivative of the control force [7-10]. As a consequence, dynamic sliding mode control technique has attracted increasing attention and it is widely used in many occasions. Hwang et al. [7] proposed a hierarchically improved fuzzy dynamic sliding mode control method in the control of autonomous ground vehicle to achieve path tracking. In [8], Liu developed a new dynamic terminal sliding manifold and applied the new dynamic sliding control method in the control of a class of SISO systems. Utkin discussed to what extent the high order sliding mode control may serve as an
alternative to the conventional sliding mode control in [9]. In [10], Fridman proposed a continuous super twisting control which is based on higher order sliding mode observer and the control method can achieve second-order sliding mode.

Adaptive control is an effective approach to handle parameter variations and it is usually combined with many other control methods to improve performances of control systems [11-16] in the presence of unknown parameters [17-20]. Banazadeh and Taymourtash [14] proposed an adaptive control method for an insect-like flapping wing air vehicle where adaptive control technique is combined with sliding mode control for attitude and position control. A novel direct adaptive tracking control scheme is established in [16] by incorporating fuzzy systems to approximate nonlinear functions.

Fractional calculus is a generalization and extension of integer differentiation and integration to fractional orders. As a branch of mathematics, this concept has attracted increasing attention of scientists and researchers due to its importance in the investigation of system modeling and control algorithms. Fractional calculus has also been integrated with adaptive control techniques [21-23] for system performances improvement. Fei and Lu [21] proposed an adaptive fractional order sliding mode control method for a Z-axis gyroscope where a neural network is used to alleviate chattering. In [22], Ahmad and Hamidreza analyzed dynamics of chaotic fractional order systems under adaptive sliding mode control method. In [23], Ali and Hamed proposed a new fractional order dynamic sliding mode method for a class of nonlinear systems.

Thus, fractional calculus can also be used in adaptive dynamic sliding mode control (ADSMC) for MEMS gyroscopes. Gyroscope performances and parameter adaptation performances under adaptive fractional order dynamic sliding mode control (AFDSMC) shall be systematically investigated as well. At the same time, the effects on control performances and parameter adaptations performances caused by different fractional orders shall also be studied in detail. To the best of our known, control methods for MEMS gyroscopes using fractional order dynamic sliding mode control method are seldom explored in the literature.

This paper proposed a dynamic sliding mode control method using fractional calculus and simulation is conducted on a Z-axis gyroscope to validate the effectiveness of the control scheme. The main contribution of the paper can be concluded as follows:

1. One superior characteristic of the proposed control scheme is that the control method developed conventional sliding method by adding a fractional order term in the sliding manifold which can provide an extra degree of freedom, so that one can achieve better trajectory tracking performance as well as parameter adaptation performance compared to conventional integer order sliding mode method. This is the most important feature of the proposed method, as compared with conventional gyroscope control methods.

2. The proposed control method improves system tracking performance as well as robustness by dealing with system nonlinearities such as parameter variation and external disturbances. Then adaptive fractional order dynamic sliding mode control methods has been extended to the control of MEMS gyroscopes. This is a successful example using fractional calculus and dynamic sliding mode control with the MEMS gyroscope.

3. Adaptive laws are proposed in the stability framework to online tune system parameters and the stability of the entire control system is guaranteed using Lyapunov stability theorem. It shall be mentioned that all the adaptive algorithms contain fractional order terms which can provide more flexibility in the controller design as well as parameters adaptation process. Besides, as long as persistent excitation condition [11] is satisfied, all system parameters can be correctly estimated.

The rest of the paper is organized as follows: in section 2, an introduction of fractional calculus is presented and a MEMS gyroscope system model is given. In section 3, an adaptive fractional order dynamic sliding mode controller is studied and the stability of the control system is also provided. Section 4 shows the simulation results and section 5 gives the conclusions.

II. Preliminary and System Description

Fractional calculus is a generalization of integration and differentiation to fractional order fundamental operation [24,25]. Denoted by $D_t^\alpha$, the fractional order operator takes both fractional order derivative and fractional integral in a single expression defined as:

$$D_t^\alpha f(t) = \begin{cases} \frac{d^\alpha}{dt^\alpha} f(t) & \alpha > 0 \\ \frac{1}{\Gamma(\alpha)} \int_t^\infty (t-s)^{\alpha-1} f(s) ds & \alpha < 0 \end{cases}$$

where $\alpha$ and $t$ are the limits of the operator and $\alpha$ is the fractional order of the operator. There are three most commonly used definitions for general fractional order operator.

**Definition 1.** for $n-1 < \alpha < n \in Z^+$, the $\alpha-th$ order Grunwald-Letnikov(GL) fractional derivative [24,25] is expressed as

$$D_t^\alpha f(t) = \lim_{h \to 0} h^{-\alpha} \sum_{j=0}^{[t/(\alpha|h|)]} (-1)^j \binom{\alpha}{j} f(t - jh)$$

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where \([(t-a)/h]\) represents the max integer number which is less than \((t-a)/h\),
\[
\left(\begin{array}{c}
\alpha \\
n \end{array}\right) = \frac{\Gamma(\alpha + 1)}{\Gamma(k + 1)\Gamma(n-k + 1)}
\]
and \(\Gamma(\bullet)\) is the gamma function, \(\Gamma(\gamma) = \int_{0}^{\infty} e^{-t}\gamma dt\).

**Definition 2.** The \(\alpha - th\) order Riemann-Letnikov(RL) fractional derivative [24,25] is written as
\[
D^\alpha f(t) = \frac{1}{\Gamma(n-\alpha)} \int_{0}^{t} f(\tau) (t-\tau)^{n-\alpha-1} d\tau
\]
where \(n-1<\alpha<n\in Z^*\).

**Definition 3.** The \(\alpha - th\) order Caputo fractional derivative [24,25] is defined as
\[
D^\alpha f(t) = \frac{1}{\Gamma(n-\alpha)} \int_{0}^{t} f(\tau) (t-\tau)^{n-\alpha-1} d\tau
\]
where \(n-1<\alpha<n\in Z^*\).

In the following parts, we will use Caputo definition in the control method design and the operator is denoted by \(D^\alpha\) for clarity [26,27].

Generally speaking, a typical Z-axis vibratory gyroscope contains a proof mass, spring beams, electrostatic actuators and sensing mechanisms. In the gyroscope, the proof mass can only move in the X-O-Y plane and the gyroscope is rotating at a constant angular velocity around the Z-axis. Due to limitation of manufacturing technology, the gyroscope structure is not totally symmetric which will cause extra coupling between X and Y axis. The motion equation [13] of the gyroscope is derived as:
\[
\begin{align*}
\dot{m}x + d_{xx} \ddot{x} + d_{xy} \ddot{y} + k_{xx} x + k_{xy} y &= u_x + 2m\Omega_y \dot{y} \\
\dot{m}y + d_{yx} \ddot{x} + d_{yy} \ddot{y} + k_{yx} x + k_{yy} y &= u_y - 2m\Omega_x \dot{x}
\end{align*}
\]
where \(m\) is the weight of the proof mass; \(x\) and \(y\) are the coordinates of the mass; \(k_{xx}, k_{yy}, d_{xx}, d_{yy}\) are the spring coefficients and damping coefficients in the X and Y direction; \(d_{xy}, k_{xy}\) are the extra coupling terms caused by the asymmetric structure and friction imperfection; \(\Omega_x\) is the angular velocity and \(u_x, u_y\) represent control forces in the X and Y direction.

Dividing both sides of (5) by the reference mass \(q_0\), the reference length \(q_0\), and the resonance frequency \(\omega_r^2\), we can get the motion equation in non-dimensional form as
\[
\begin{align*}
\ddot{x} + d_{xx}^* \ddot{x} + d_{xy}^* \ddot{y} + \omega_r^2 x + \omega_{xy} \dot{y} &= u_x^* + 2\Omega_y \dot{y} \\
\ddot{y} + d_{yx}^* \ddot{x} + d_{yy}^* \ddot{y} + \omega_r^2 y + \omega_{yx} \dot{x} + \omega_{yy} \dot{y} &= u_y^* - 2\Omega_x \dot{x}
\end{align*}
\]
where \(x^* = \frac{x}{q_0}, y^* = \frac{y}{q_0}, \dot{x}^* = \frac{\dot{x}}{q_0\omega_0}, \dot{y}^* = \frac{\dot{y}}{q_0\omega_0}\),
\[
y^* = \frac{\dot{y}}{q_0\omega_0}, y^* = \frac{\dot{y}}{q_0\omega_0}^2, d_{xx}^* = \frac{d_{xx}}{m\omega_0^2}, d_{yy}^* = \frac{d_{yy}}{m\omega_0^2},
\]
dividing both sides of (6) by the reference length \(q_0\), the reference mass \(m\),
\[
d_{xy}^* = \frac{d_{xy}}{m\omega_0^2}, \Omega_x^* = \frac{\Omega_x}{\omega_y^2}, \Omega_y^* = \sqrt{\frac{k_{xy}}{m\omega_0^2}}, \omega_{xy} = \sqrt{\frac{k_{xy}}{m\omega_0^2}}
\]
Rewriting (6) in vector form yields
\[
\ddot{q} + D\dot{q} + Kq = u - 2\Omega \dot{q}
\]
where
\[
q = \begin{bmatrix} x^* \\ y^* \end{bmatrix}, u = \begin{bmatrix} u_x^* \\ u_y^* \end{bmatrix}, D = \begin{bmatrix} d_{xx}^* & d_{xy}^* \\ d_{yx}^* & d_{yy}^* \end{bmatrix}, K = \begin{bmatrix} \omega_r^2 & \omega_{xy} \\ \omega_{yx} & \omega_r^2 \end{bmatrix}, \Omega = \begin{bmatrix} 0 & -\Omega_x^* \\ \Omega_y^* & 0 \end{bmatrix}
\]

Taking parameters variation and external disturbances into consideration, we can further get the motion equation in vector form as
\[
\ddot{q} = u - (D + 2\Omega)\dot{q} - Kq - (\Delta D + 2\Delta \Omega)\dot{q} - \Delta Kq + d
\]
For simplicity, (8) is rewritten in the form
\[
\ddot{q} = f + u + F
\]
Where \(f = -(D + 2\Omega)\dot{q} - Kq\), \(q\) is the state vector, \(u\) is the control vector and \(F = \begin{bmatrix} F_1 \\ F_2 \end{bmatrix}\) represents lumped, matched unknown disturbances including parameters variation and external disturbance, expressed as \(F = -(\Delta D + 2\Delta \Omega)\dot{q} - \Delta Kq + d\).

**Assumption 1.** The lumped parameter uncertainties and external disturbance \(F\) is bounded such that \(|F_i| \leq F_{i_0}\), \(i = 1, 2\). The derivative of the lumped parameter uncertainties and external disturbance is bounded such that \(|\dot{F}_i| \leq \dot{F}_{i_0}\), \(i = 1, 2\), where \(F_{i_0}\), \(\dot{F}_{i_0}\) are positive constants.

In the next section, an adaptive fraction order dynamic manifold is proposed and the design of an adaptive fractional order dynamic sliding mode controller is investigated so that gyroscope trajectory can track the reference trajectory.

**III. Adaptive Fractional Order Dynamic Sliding Mode Controller**
3.1 Design of Fractional Order Dynamic Sliding Mode Control

An adaptive fractional order dynamic sliding mode controller is established with the objective to find a control law so that the gyroscope state $\dot{q}$ can track a reference trajectory $\dot{q}_d$. Assuming that all parameters in the gyroscope system are well known, the design of an ideal fractional dynamic sliding mode controller is described step by step as follows:

Define the tracking error as

$$ e = \dot{q}_d - \dot{q} $$(10)

The derivative of tracking error is

$$ \dot{e} = \dot{\dot{q}}_d - \ddot{q} $$ (11)

Define a fractional order sliding surface as

$$ S = \dot{\epsilon} + c_2 \epsilon + c_3 \dot{D}^{\alpha-1} \epsilon $$ (12)

where $c_2 = \begin{bmatrix} c_{21} & 0 \\ 0 & c_{22} \end{bmatrix}$ and $c_3 = \begin{bmatrix} c_{31} & 0 \\ 0 & c_{32} \end{bmatrix}$ are sliding surface parameters which are chosen to be diagonal matrices with the diagonal elements $c_{2i}, c_{3i}, i = 1, 2$ being positive constants. $\alpha - 1$ is the fractional order of fractional order operation.

The derivative of the sliding surface is

$$ \dot{S} = \dot{\dot{\epsilon}} + c_2 \dot{\epsilon} + c_3 \dot{D}^{\alpha-1} \dot{\epsilon} $$ (13)

$$ = (\ddot{\dot{q}}_d - \ddot{q}) + c_2 (\ddot{\dot{q}}_d - \ddot{q}) + c_3 \dot{D}^{\alpha} \epsilon $$

A fractional order dynamic sliding manifold is designed as

$$ \sigma = \dot{S} + \partial S $$ (14)

$$ \partial = \begin{bmatrix} \partial_1 & 0 \\ 0 & \partial_2 \end{bmatrix} $$

where $\partial_i, i = 1, 2$ are positive constants.

It can be seen from the definition of the dynamic sliding manifold in (14) that if $\sigma \to 0$, $\dot{S} + \partial S = 0$ is a stable system only if all the roots of $\dot{S} + \partial S = 0$ are located in the left half plane. That is to say if $\sigma \to 0$, the sliding surface designed in (12) will converge to zero with $\dot{\epsilon}$ being a positive diagonal matrix. And it can be inferred from the definition of the sliding surface in (12) that the tracking error will also converge to zero.

Differentiating both sides of (4), the derivative of the dynamic surface is

$$ \dot{\sigma} = \ddot{\dot{\epsilon}} + \partial \dot{\epsilon} $$

$$ = \ddot{\dot{\epsilon}} + c_2 \dot{\epsilon} + c_3 \dot{D}^{\alpha+1} \epsilon + \partial (\ddot{\dot{\epsilon}} + c_2 \dot{\epsilon} + c_3 \dot{D}^{\alpha} \epsilon) $$

$$ = (\ddot{\dot{q}}_d - \ddot{q}) + (c_2 + \partial) (\ddot{\dot{q}}_d - \ddot{q}) $$

$$ + c_3 \dot{D}^{\alpha+1} \epsilon + \dot{c}_3 \dot{D}^{\alpha} \epsilon + \ddot{c}_3 \dot{\epsilon} $$

$$ = (\ddot{\dot{q}}_d - \ddot{\dot{q}} - \ddot{\dot{\epsilon}}) + (c_2 + \partial) (\ddot{\dot{q}}_d - \ddot{\dot{\epsilon}}) $$

$$ + 2 \dot{c}_3 \dot{D}^{\alpha+1} \epsilon + \dot{c}_3 \dot{D}^{\alpha} \epsilon + \ddot{c}_3 \dot{\epsilon} $$

Then, a sliding mode controller can be designed in the following form

$$ u = u_{eq} + u_{sw} $$ (16)

Where

$$ u_{eq} = (c_2 + \partial) \frac{1}{\dot{c}_3} [(\ddot{\dot{q}}_d - \ddot{\dot{q}} - \ddot{\dot{\epsilon}} - \ddot{\dot{\dot{\epsilon}}}) $$

$$ + (c_2 + \partial) (\ddot{\dot{q}}_d - \ddot{\dot{\epsilon}}) $$

$$ + c_3 \dot{D}^{\alpha+1} \epsilon + \dot{c}_3 \dot{D}^{\alpha} \epsilon + \ddot{c}_3 \dot{\epsilon} ] $$

$$ u_{sw} = (c_2 + \partial) \frac{1}{\dot{c}_3} \eta \text{sgn} (\sigma) $$ (17)

$$ \eta = \begin{bmatrix} \eta_1 \\ \eta_2 \end{bmatrix} $$

$\eta_1, i = 1, 2$ chosen to be positive constants.

**Theorem 1:** If the control force designed in (16) is applied to the MEMS gyroscope system described in (9), the entire control system is stable.

**Proof:** Choose a Lyapunov candidate as

$$ V = \frac{1}{2} \sigma^T \sigma $$ (19)

Differentiating both sides of (19) and substituting (15) into $\dot{V}$ leads to

$$ \dot{V} = \sigma^T \dot{\sigma} $$

$$ = \sigma^T (\ddot{\dot{\dot{\epsilon}}} - \ddot{\dot{\dot{\epsilon}}}) $$

$$ + (c_2 + \partial) (\ddot{\dot{\dot{\epsilon}}} - \ddot{\dot{\dot{\epsilon}}}) $$

$$ + c_3 \dot{D}^{\alpha+1} \epsilon + \dot{c}_3 \dot{D}^{\alpha} \epsilon + \ddot{c}_3 \dot{\epsilon} $$

Substituting the control force in (16) into (20) yields

$$ \dot{V} = \sigma^T \dot{\sigma} $$

$$ = \sigma^T (-\eta \text{sgn} (\sigma)) $$

$$ = -\eta \sigma^T \leq 0 $$

Since the Lyapunov function is positive definite and the derivative of the Lyapunov function is negative definite, according to the Lyapunov stability theorem, it can be concluded that the entire control system is asymptotically stable.

3.2 Design of Adaptive Fractional Order Dynamic Sliding Mode Control

If the system parameters are unknown, the dynamic sliding mode controller designed in (16) cannot be implemented directly. Adaptive control technique is
adopted in this part to online tune the unknown system parameters in (16).

Control force in (16) can be modified in the new form
\[
\dot{u}_{eq} = (c_2 + \partial)(\ddot{q}_d - \dot{f} - \dot{\hat{u}}_{eq}) + (c_2 + \partial)(\ddot{q}_d - \dot{f} - \hat{u}_{eq}) + c_3D^{\alpha+1}\dot{e} + \partial c_3D^\alpha e + \partial c_\varepsilon\dot{e}
\]
(22)
where \(\hat{f} = -(\hat{D} + 2\hat{K}\hat{q} - \hat{K}\hat{q})\) is an estimate of \(f\).

And the proposed modified adaptive controller is designed as
\[
\dot{u} = \hat{u}_{eq} + u_{sw}
\]
(23)
where \(\hat{u}_{eq}\) is shown in (22) and \(u_{sw}\) remains the same as (18).

In order to update the system parameters in (23), the adaptive laws are given in the following form
\[
\dot{\theta}_1 = \eta_1\hat{\theta}_1 + \eta_2\hat{\theta}_2 + \eta_3\hat{\theta}_3 (c_2 + \partial)
\]
(24)
\[
\dot{\theta}_2 = 2(\eta_1\hat{\theta}_1 + \eta_2\hat{\theta}_2 + \eta_3\hat{\theta}_3 (c_2 + \partial))
\]
(25)
\[
\dot{\theta}_3 = \eta_1\hat{\theta}_1 + \eta_2\hat{\theta}_2 + \eta_3\hat{\theta}_3 (c_2 + \partial)
\]
(26)

The block diagram of the designed adaptive fractional order sliding mode control system is depicted in Fig.1.

![Fig.1 Block diagram of the adaptive dynamic sliding mode controller](image)

**Theorem 2:** If the designed controller in (23) and the adaptive laws in (24), (25), (26) are applied to the system in (9), the entire control system is asymptotically stable, system tracking errors will converge to zero and all the system parameters can be correctly estimated.

**Proof:** Choose a new Lyapunov candidate as
\[
V = \frac{1}{2}\sigma^T\sigma + \frac{1}{2}\text{tr}(D^T\frac{1}{\eta_1}D) + \frac{1}{2}\text{tr}(K^T\frac{1}{\eta_2}K) + \frac{1}{2}\text{tr}(\tilde{\Omega}^T\frac{1}{\eta_3}\tilde{\Omega})
\]
(27)
where \(\sigma\) is the dynamic sliding manifold, \(D, K, \tilde{\Omega}\) are the estimation errors of the known system parameters defined such that \(\ddot{D} - \hat{D} = \tilde{D}, \dot{\Omega} - \hat{\Omega} = \tilde{\Omega}, K - \hat{K} = \tilde{K}\).

Differentiating both sides of (27) and substituting (15) into \(V\) yield
\[
\dot{V} = \sigma^T\dot{\sigma} + \text{tr}(\dot{D}^T\frac{1}{\eta_1}D) + \text{tr}(\dot{K}^T\frac{1}{\eta_2}K) + \text{tr}(\dot{\tilde{\Omega}}^T\frac{1}{\eta_3}\tilde{\Omega})
\]
(28)
\[
= \sigma^T\left[(\ddot{q}_d - \dot{f} - \ddot{u}) + (c_2 + \partial)(\ddot{q}_d - \dot{f} - \dot{\hat{u}}_{eq}) + c_3D^{\alpha+1}\dot{e} + \partial c_3D^\alpha e + \partial c_\varepsilon\dot{e}\right] + \text{tr}(\dot{\sigma})
\]
where \(\text{tr}(A)\) is the trace of the matrix \(A\), and \(\text{tr}(\dot{A})\) is the trace of the derivative of the matrix \(A\) with respect to time.

Substituting the adaptive controller in (23) into (28) gives
\[
\dot{V} = \sigma^T(\ddot{q}_d - \dot{f} - \ddot{u} - \hat{F} + (c_2 + \partial)(\ddot{q}_d - \dot{f} - \dot{\hat{u}}_{eq}) + \text{tr}(\dot{\sigma})
\]
(29)
\[
= \sigma^T(\ddot{q}_d - \dot{f} - \ddot{u} - \hat{F} + (c_2 + \partial)(\ddot{q}_d - \dot{f} - \dot{u}) + \text{tr}(\dot{\sigma})
\]
\[
+ (c_2 + \partial)(\ddot{q}_d - \dot{f} - \dot{\hat{u}}_{eq} - (c_2 + \partial)f + c_3D^{\alpha+1}e + \partial c_3D^\alpha e + \partial c_\varepsilon\dot{e}) + \text{tr}(\dot{\sigma})
\]
\[
+ (c_2 + \partial)(\ddot{q}_d - \dot{f} - \dot{\hat{u}}_{eq} - (c_2 + \partial)f + c_3D^{\alpha+1}e + \partial c_3D^\alpha e + \partial c_\varepsilon\dot{e}) + \text{tr}(\dot{\sigma})
\]
\[
= \sigma^T[-\hat{F} - (c_2 + \partial)F - \dot{f} - (c_2 + \partial)f + (c_2 + \partial)\dot{\hat{u}}_{eq} - \text{tr}(\dot{\sigma})
\]
\[
+ (c_2 + \partial)(\ddot{q}_d - \dot{f} - \dot{\hat{u}}_{eq} - (c_2 + \partial)f + c_3D^{\alpha+1}e + \partial c_3D^\alpha e + \partial c_\varepsilon\dot{e}) + \text{tr}(\dot{\sigma})
\]
\[
= \sigma^T[-\hat{F} - (c_2 + \partial)F - \dot{f} - (c_2 + \partial)f + (c_2 + \partial)\dot{\hat{u}}_{eq} - \text{tr}(\dot{\sigma})
\]
\[
- \eta\text{sgn}(\sigma) + \text{tr}(\dot{\sigma})
\]
\[
= \sigma^T[-\hat{F} - (c_2 + \partial)F - \dot{f} - (c_2 + \partial)f + (c_2 + \partial)\dot{\hat{u}}_{eq} - \text{tr}(\dot{\sigma})
\]
\[
- \eta\text{sgn}(\sigma) + \text{tr}(\dot{\sigma})
\]
\[
= \sigma^T[-\hat{F} - (c_2 + \partial)F - \dot{f} - (c_2 + \partial)f + (c_2 + \partial)\dot{\hat{u}}_{eq} - \text{tr}(\dot{\sigma})
\]
\[
- \eta\text{sgn}(\sigma) + \text{tr}(\dot{\sigma})
\]
\[
= \sigma^T[-\hat{F} - (c_2 + \partial)F - \dot{f} - (c_2 + \partial)f + (c_2 + \partial)\dot{\hat{u}}_{eq} - \text{tr}(\dot{\sigma})
\]
\[
- \eta\text{sgn}(\sigma) + \text{tr}(\dot{\sigma})
\]
where $\tilde{f} = \dot{f} - f$ and $\tilde{f}$ is described as
\[\tilde{f} = -\left(\dot{D} + 2\Omega\tilde{q} - \tilde{K}q\right) - ((D - 2\Omega)\tilde{q} - Kq)\]
\[= (D - \dot{D})\tilde{q} + 2(\Omega - \dot{\Omega})\tilde{q} + (K - \dot{K})q\]
\[= (\tilde{D} - D)\tilde{q} - K\tilde{q}\]

The derivative of $\tilde{f}$ is given in the form
\[\dot{\tilde{f}} = (\tilde{D} - D)\tilde{q} - K\tilde{q}\]

Substituting (31) into (29) yields
\[\dot{V} = \sigma^T[\dot{f} - (c_2 + \partial)F] + \sigma^T \eta \text{sgn}(\sigma) + \text{tr}(\dot{\hat{K}})\]
\[+ \left([-\dot{D} - 2\Omega]q - \tilde{K}\tilde{q}\right) - \eta \text{sgn}(\sigma)\]
\[+ \text{tr}(\dot{\hat{K}})\]
\[= \sigma^T[-\dot{f} - (c_2 + \partial)F] + \sigma^T \eta \text{sgn}(\sigma) + \text{tr}(\dot{\hat{K}})\]
\[\leq \sigma^T[-\dot{f} - (c_2 + \partial)F] + \eta \text{sgn}(\sigma) + \text{tr}(\dot{\hat{K}})\]
\[= \sigma^T[-\dot{f} - (c_2 + \partial)F] + \eta \text{sgn}(\sigma) + \text{tr}(\dot{\hat{K}})\]

Substituting adaptive laws in (24), (25), (26) into (32) yields
\[\dot{V} = \sigma^T[-\dot{f} - (c_2 + \partial)F] + \eta \text{sgn}(\sigma) + \text{tr}(\dot{\hat{K}})\]

With Assumption 1, we can further get
\[\dot{V} \leq [\dot{F}_d + (c_2 + \partial)F_d - \eta] |\sigma^T|\]

If the robust sliding gain $\eta$ is selected so that $\eta \geq [\dot{F}_d + (c_2 + \partial)F_d]$, $\dot{V} \leq 0$.

$V$ is semi-negative definite implies that $\sigma, \tilde{D}, \tilde{K}, \tilde{\Omega}$ are all bounded. It can be concluded from (15) that $\dot{\sigma}$ is also bounded. The inequality $\dot{V} \leq [\dot{F}_d + (c_2 + \partial)F_d - \eta] |\sigma^T|$ implies that $|\sigma^T|$ is integrable as
\[\int_0^t |\sigma^T| dt \leq \frac{1}{F_d + (c_2 + \partial)F_d - \eta} (V(0) - V(t))\]

Since $V(0)$ is bounded, $V(t)$ is bounded and non-increasing. It can be concluded that $\int_0^t |\sigma^T| dt$ is also bounded. Since $\int_0^t |\sigma^T| dt$ is bounded and $\dot{\sigma}$ is also bounded, according to Barbalart lemma, $\sigma$ will asymptotically converge to zero.

IV. SIMULATION STUDY

This section has 3 parts, where trajectory tracking performances using ADSMC and AFDSMC are presented and parameters adaptation performances are also shown in the first part. In the second part, system responses under AFDSMC with different fractional orders are provided. Parameter adaptation performances under different fractional orders are also shown in the second part. In the third part, system dynamics with frequency deviation are studied.

It shall be emphasized that all the parameters in the design of sliding surfaces, dynamic surfaces, and adaptive laws are all the same. The only difference between ADSMC and AFDSMC is that a fractional term is added in the design of AFDSMC controller while the sliding surface of ADSMC has only integral terms. And in the second part, all simulation results are derived using AFDSMC with exactly the same parameters; the only difference is that the fractional orders are different.

As an illustrative example, we use a z-axis gyroscope dynamic model [13] for the study of adaptive fractional order dynamic sliding mode controller.

Parameters of the gyroscope dynamic model (7) are set as follows:
\[m = 1.8 \times 10^{-7} \text{kg}, \quad k_{zz} = 63.955 \text{N/m}, \quad k_{yy} = 95.92 \text{N/m}, \quad k_{xx} = 12.779 \text{N/m}, \quad d_{zz} = 1.8 \times 10^{-6} \text{N \cdot s/m}, \quad d_{yy} = 1.8 \times 10^{-6} \text{N \cdot s/m}, \quad d_{xx} = 3.6 \times 10^{-7} \text{N \cdot s/m}\]
Choose a reference length $q_0 = 1\mu m$, reference frequency $\omega_0 = 1kHz$, and the angular velocity $\Omega = 100rad/s$. The parameters can be obtained through non-dimensional transformation: $\omega_x^2 = 355.3$, $\omega_y^2 = 532.9$, $\omega_{xy} = 70.99$, $d_{xx}^* = 0.01$, $d_{yy}^* = 0.01$, $d_{xy}^* = 0.002$, $\Omega^*_x = 0.1$.

Initial conditions on matrix $D$, $K$, $\Omega$ are $D(0) = 0.95D$, $K(0) = 0.95K$, $\Omega(0) = 0.95\Omega$. The desired motion trajectory $x_d = \sin(4.11t)$ to $y_d = 0.7\sin(5.11t)$ to $x_d = \sin(2.085t)$ and $y_d = 0.7\sin(2.555t)$ at time point $t = 12\pi / 4.17$ in X axis and $t = 12\pi / 5.11$ in Y axis. External disturbances are $d = \begin{bmatrix} d_1 \\ d_2 \end{bmatrix} = \begin{bmatrix} \text{rand}(1) \\ \text{rand}(1) \end{bmatrix}$ where disturbance in each channel is uniformly distributed between 0-1.

Parameters of ADSMC and AFDSMC and Gains of adaptive laws (24)-(26) are provided in Table.1.

### Parameters in the design of ADSMC and AFDSMC

<table>
<thead>
<tr>
<th>Controller parameters</th>
<th>ADSMC</th>
<th>AFDSMC</th>
<th>AFDSMC</th>
<th>AFDSMC</th>
<th>AFDSMC</th>
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<tr>
<td>sliding surface $c_2$</td>
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<td>1 0</td>
<td>1 0</td>
<td>1 0</td>
<td>1 0</td>
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<td>0 1</td>
<td>0 1</td>
<td>0 1</td>
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<tr>
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<td>1 0</td>
<td>1 0</td>
<td>1 0</td>
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<tr>
<td>Parameter 3#</td>
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<td>0 1</td>
<td>0 1</td>
<td>0 1</td>
<td>0 1</td>
</tr>
<tr>
<td>Fractional order</td>
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<td>0.15</td>
<td>1.85</td>
<td></td>
</tr>
<tr>
<td>Dynamic sliding surface Parameter $\partial$</td>
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<td>1 0</td>
<td>1 0</td>
<td>1 0</td>
<td>1 0</td>
</tr>
<tr>
<td></td>
<td>0 1</td>
<td>0 1</td>
<td>0 1</td>
<td>0 1</td>
<td>0 1</td>
</tr>
<tr>
<td>Robust gain $\eta_1$</td>
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<td>50 50</td>
<td>50 50</td>
<td>50 50</td>
<td>50 50</td>
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<tr>
<td>Gain of (24) $\eta_2$</td>
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<tr>
<td>Gain of (25) $\eta_3$</td>
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<td>20</td>
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<td>20</td>
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<tr>
<td>Gain of (26) $\eta_4$</td>
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<td>50</td>
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</tr>
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</table>

### 4.1 Comparisons between ADSMC and AFDSMC

System trajectory tracking performances and tracking errors using ADSMC and AFDSMC are shown in Fig.2 and Fig.3. It can be found from Fig.2 that both red line (AFDSMC) and blue line (ADSMC) can track the black line (command trajectory) in a few seconds and the three lines almost overlap with each other in the end. In is shown in Fig.3 that both blue line (ADSMC) and red line (AFDSMC) will converge to zero, this means that system tracking errors using ADSMC and AFDSMC will converge to zero. It can be observed from the partial enlarged pictures in both tracking error figures that red lines (AFDSMC) response more quickly than blue lines (ADSMC) and red lines (AFDSMC) use less time than blue line (ADSMC) to converge to zero, this means that the fractional order terms in AFDSMC scheme can improve system transient performances.

Fig.4 and Fig.5 depict adaptation performances of system parameters and angular velocity. It can be clearly seen from Fig.4 that red line (AFDSMC) uses less time to become stable and the values red and blue lines converge to are exactly system parameters we set in the system. It can be seen in Fig.5 that both blue line (ADSMC) and red line (AFDSMC) are going up and down around 0.1 which is just the angular velocity set in the system. That is to say the adaptive laws using fractional order terms can correctly estimate system parameters and fractional order adaptive laws has better parameter identification performance than integral order adaptive laws.
Fig. 2 Comparison of tracking performances using ADSMC and AFDSMC

Fig. 3 Tracking errors using ADSMC and AFDSMC

Fig. 4 Adaptation of parameters of ADSMC and AFDSMC

Fig. 5 Adaptation of angular velocity of ADSMC and AFDSMC

4.2 System Dynamic and Parameter Adaptation Using AFDSMC under Different Fractional Orders

It shall be claimed that all the parameters in all the simulation cases in this part are all the same except the fractional order in the sliding surface. It can be seen from Fig. 7-Fig. 11 that there are 4 different color lines where green line represents $\alpha - 1 = 0.3$, pink line represents $\alpha - 1 = 0.85$, red line represents $\alpha - 1 = -0.15$ and blue line represents $\alpha - 1 = -0.85$. The positive fractional orders represent derivation operation and negative fractional orders represent integration operation.

System dynamics and tracking errors under AFDSMC with different fractional orders are shown in Fig. 7 and Fig. 8. It can be found from Fig. 7 that all the 4 lines can track the black line (command trajectory) in a few seconds and the 4 lines almost overlap with each other in the end. It can be seen from Fig. 8 that all the tracking errors under different fractional orders will converge to zero and it can be observed from the partial enlarged pictures in both tracking error figures that pink lines ($\alpha - 1 = 0.85$) have more rapid response than the other 3 lines while green line ($\alpha - 1 = 0.3$) responses more quickly than the other 2. This implies that adding fractional order derivative terms in the sliding surface can accelerate system response where some oscillations can be observed in the pink line ($\alpha - 1 = 0.85$). Comparing blue line ($\alpha - 1 = -1.85$) with pink line ($\alpha - 1 = 0.85$), both lines have some oscillations and both lines use more time than green ($\alpha - 1 = 0.3$) and red lines ($\alpha - 1 = -0.15$) to converge to zero. But there are some differences that the error dynamic of blue line ($\alpha - 1 = -1.85$) is very smooth while the error dynamic of pink line ($\alpha - 1 = 0.85$) is very intense. It can be inferred that big fractional derivative order will result intense system response while big fractional integration order can decelerate system responses.

Adaptation performances of system parameters and angular velocity under different fractional orders are shown in Fig. 9 and Fig. 10 where pink lines response more intensively than the other 3. Observed from the partial enlarged pictures in each figure, the adaptation progress of pink line and green line is more rapid than the other 2. This situation is also in accordance with error dynamics in Fig. 8.
Fig. 11 depicts control forces under different fractional orders showing that different fractional order terms can adjust system dynamics without causing chattering phenomenon.

Fig. 7 Tracking performances of AFDSMC under different factional orders

Fig. 8 Tracking errors of AFDSMC under different factional orders

Fig. 9 Adaptation of parameters using AFDSMC under different factional orders

Fig. 10 Adaptation of angular velocity using AFDSMC under different factional orders

Fig. 11 Control forces using AFDSMC under different factional orders

4.3 Tracking Performances Evaluation under Frequency Deviation

Control performances using AFDSMC under different fractional orders have been provided in part 4.1 and 4.2. In order to investigate performances of the control system under frequency deviation, we provide results when frequencies of command signals change from frequency A to frequency B.

Fig. 12 Tracking performances of AFDSMC under frequency deviation

Fig. 13 Tracking error of AFDSMC under frequency deviation

V. CONCLUSION

This paper proposes an adaptive fractional order dynamic sliding mode controller for MEMS gyroscopes. Fractional calculus is adopted in the design of dynamic sliding mode control and the new fractional control scheme...
is applied to a MEMS gyroscope model. Dynamic sliding mode control can help reduce chattering in control forces while fractional calculus can improve system tracking performance. Adaptive control technique is also incorporated in the controller design where the control law and all the adaptive laws are derived in the Lyapunov framework to guarantee the asymptotic stability of the closed-loop system. As long as persistent excitation condition is satisfied, all system parameters including the angular velocity can be correctly estimated. Simulation results verify the validity of the proposed control approach, demonstrating that fractional calculus can improve system tracking performances as well as parameter estimation performance.

VI. ACKNOWLEDGEMENT

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REFERENCES