TDE based model-free control for rigid robotic manipulators under nonlinear friction

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\begin{abstract}
This paper establishes a model-free finite-time tracking control of nonlinear robotic manipulator systems. The proposed controller incorporates both Time Delay Estimation (TDE) and an enhanced Terminal Sliding Mode Control (TSMC). The improved TSMC scheme is devised using Fractional-Order TSMC (FOTSMC) and Proportional-Integral-Derivative (PID) control to obtain robust tracking and high control performance. The TDE is designed to estimate the unknown nonlinear dynamics of robotic manipulators, including the Striebeck friction and the external disturbances. Due to Striebeck friction, the effect of TDE error may fail to obtain the desired error performance; thus, another TDE loop is devised to compensate for TDE error generated by non-smooth frictions. The Lyapunov criterion is used to investigate the finite-time stability to analyze the behavior of the designed approach. Finally, computer simulations of the proposed method on PUMA 500 robotic manipulators are performed in contrast with FOTSMC and Adaptive Fractional-Order Nonsingular Terminal Sliding Mode Control (AFONTSMC).
\end{abstract}

\begin{keywords}
Model-free control; Time delay estimation; Sliding mode control; Striebeck friction; Robotic manipulators.
\end{keywords}

\section{1. Introduction}

A robotic manipulator is a nonlinear mechanical device and is generally used in processing/manufacturing industries and its applications because it is cost-effective and replaces manual labor for complicated and repetitive tasks \cite{1,2,3,4}. The complex dynamics of robotic manipulators having inherent uncertainties, dynamic coupling between neighboring links, time-varying inertia, gravity, and nonlinear frictions require efficient controlling to obtain high control performance of rapid dynamic convergence, repeatable accuracy, finite-time stability, smooth control input, and minimal vibration at the desired angle \cite{5}.

In general, nonlinear effects such as presliding displacement, backlash hysteresis, Dahl effect, and Striebeck friction are present in every mechanical system. Basically, the Striebeck friction consists of the Coulomb friction, static friction, and viscous friction \cite{6}. Thus, the motorized mechanism is widely affected by these frictions where moving parts make contact with each other, and it is impractical to pay no attention to the control design. Conversely, this severe nonlinearity may degrade the control performance of the closed-loop system.

A variety of control approaches have been designed, such as neural network-based piecewise con-
tinium function control To meet effectual control performance under Stribeck frictions [7], adaptive fuzzy control [8], recursive model-free control [9], Sliding Mode Control (SMC) [10–12]. However, some of these schemes rely on the knowledge of the system dynamics or information of uncertain parameters [13]. Moreover, some intelligent learning control methodologies, such as fuzzy logic and neural network control, require complex calculations caused by the weight training processes of intelligent control.

Toward this front, to avoid complicated formulations and estimate unknown system dynamics and uncertainties, the Time Delay Estimation (TDE) can be employed to achieve high control performance and implemented easily. Fundamentally, TDE is an estimation method that originates from Time Delay Control (TDC) theory. In this way, the unknown dynamics and uncertainties of a system are estimated by exploiting the system’s delayed dynamics. Thus, the constant time delay is inserted with known system parameters, such as control input and the state derivatives, to obtain delayed unknown dynamics. In literature, TDE has been incorporated with well-known control methodologies, for example, SMC, fuzzy control, neural network control, and intelligent proportional-integral-derivative (iPID) control [14–18] to obtain a precise estimation as well as better control performances.

On the other hand, TDE cannot precisely estimate the unmodeled dynamics because of the TDE estimation error, which is inevitable due to inherent non-smooth frictions and nonlinearities. Therefore, in order to tackle estimation error, TDE is usually used with various control schemes such as adaptive control, neural network control, fuzzy logic, Ideal Velocity Feedback (IVF), and anti-windup schemes [17–21]. TDE estimates the unknown dynamics, and the other schemes suppress the estimation error. As mentioned earlier, the intelligent schemes are complex enough, and the adaptive control may not be estimated to some extent due to its constant tuning gain. Therefore, the TDE scheme is suitable for its simplicity and precise estimation. In this work, TDE is employed to estimate the unmodeled dynamics, and then the other TDE is utilized to deal with the estimation error caused by the Stribeck friction’s effect.

SMC is one of the most robust control schemes in control engineering; however, conventional SMC has some drawbacks, such as slow convergence speed, oscillation in control, and singularity [22]. Thus, Terminal Sliding Mode Control (TSMC), fast SMC, and nonsingular TSMC have been designed to overcome these problems [23–27]. Furthermore, to enhance the TSMC performances, such as robustness and dynamic response, TSMC based on PID (TSMC-PID) has been designed [28]. Moreover, Fractional-Order (FO) control is an arbitrary order of generalized calculus that improves the dynamic response of the controller. Thus, FO has been integrated with TSMC to improve the tracking accuracy and transient response of the closed-loop system [29–32]. Therefore, this work considers fractional-order TSMC-PID with the estimation of uncertain dynamics under Stribeck friction through TDE to design the robust model-free scheme Fractional-Order TSMC (FOTS PMC) and Proportional-Integral-Derivative (PID). At the same time, the chattering problem is attenuated by replacing the sgn function with the \( \tanh \) function. The main goal of this work can be marked as follows:

1. Unlike the intelligent learning methods, TDE based FOTS PMC-PID control scheme under Stribeck friction is proposed to obtain model-free control with finite-time convergence, high precision, and robustness;
2. FOTS PMC-PID is utilized to obtain robust and accurate performances, and unknown dynamics are estimated by TDE. In contrast, TDE error is compensated by integrating augmented control via another TDE approach;
3. A proof of finite-time stability is investigated by Lyapunov stability synthesis;
4. The proposed scheme shows a faster convergence rate and robustness with compared approaches.

The article is organized as follows: Mathematical preliminaries are given in Section 2. Section 3 introduces the dynamics of robotic manipulators. In Sections 4 and 5, the proposed TDE framework with FOTS PMC-PID for robotic manipulators and its finite-time stability analysis by Lyapunov synthesis is demonstrated, respectively. Section 6 presents the resultant comparative simulations to validate the efficacy of the developed scheme. This paper is concluded in Section 7.

2. Mathematical preliminaries

2.1. Definition 1

A perturbed nonlinear system with state \( z(t) \) is defined as:

\[
\dot{z}(t) = g(z) + h(z)u(t) + p(t),
\]  \hspace{1cm} (1)

where \( g(z) \) represents the unknown nonlinear state dynamics function, \( h(z) \) denotes a distribution matrix, \( p(t) \) is an unknown external disturbance, and the control input is given by \( u(t) \). By separating known and unknown terms, Eq. (1) can be written as:

\[
\Theta(z, t) = g(z) + p(t) = \dot{z}(t) - h(z)u(t).
\]  \hspace{1cm} (2)

The TDE of an unmodeled term can be computed as [33]:
\( \dot{\Theta}(z, t) \triangleq \dot{g}(z) + \dot{p}(t-d) \triangleq g(z-d) + p(t-d) \)
\( = \dot{z}(t-d) - h(z-d)u(t-d), \) (3)

where \( d \) is the constant delay, and \( \dot{\Theta}(z, t) \) is the estimated expression of unknown dynamics.

2.2. Definition 2

The \( \gamma \)-th-order Riemann-Liouville (RL) fractional differential of function \( z(t) \) with terminal value \( b \) is defined by Podlubny [34]:
\[ \mathbb{I}^\gamma_z z(t) = \mathbb{I}^\gamma_z z(t) = \frac{1}{\Gamma(\gamma)} \int_0^t \frac{z(\tau)}{(t-\tau)^{1-\gamma}} \, d\tau, \]
\[ \mathbb{D}^\gamma_z z(t) = \frac{d^m}{dt^m}(\mathbb{I}^{\gamma-m}_z z)(t) = \mathbb{I}^{\gamma-m}_z \frac{d^m}{dt^m} z(t), \] (4)

where \( \mathbb{I}^\gamma \) and \( \mathbb{D}^\gamma \) represent the fractional integral and derivative, respectively. Fractional value \( \gamma \) ranges \( m-1 < \gamma < m \) and \( m \in \mathbb{N} \), while \( \Gamma(\cdot) \) denotes Euler’s Gamma function as:
\[ \Gamma(\gamma) = \int_0^\infty e^{-t} t^{\gamma-1} \, dt. \]

2.3. Lemma 1

By taking the ordinary derivative \( (d^m/dt^m) \) of fractional operator \( \mathbb{I}^\gamma_z z(t) \) yields [34]:
\[ \frac{d^m}{dt^m}(\mathbb{I}^\gamma_z z(t)) = \mathbb{D}^\gamma \left( \frac{d^m}{dt^m} z(t) \right) = \mathbb{I}^{\gamma+m}_z z(t). \] (5)

2.4. Lemma 2

For Lyapunov function \( \mathcal{V}(t) \) with initial value \( \mathcal{V}(t_0) \), finite-time stability is implied by Tang [35]:
\[ \mathcal{V}(t) \leq -n \mathcal{V}^p(t), \quad \forall t \geq t_0, \quad \mathcal{V}(t_0) \geq 0, \] (8)

where \( n > 0 \) and \( 0 < p < 1 \). The finite-time \( t_f \) can be estimated as:
\[ t_f \leq \frac{1}{n(1-p)} \mathcal{V}^{1-p}(t_0). \] (9)

3. Dynamics of robotic manipulators

The uncertain dynamics of \( n \)-link robotic manipulators can be expressed in the LaGrange form as:
\[ M(q) \ddot{q} + V_c(q, \dot{q}) \ddot{q} + G(q) + \mathcal{F}(\dot{q}) + \tau_d = \tau, \] (10)

where \( q, \dot{q}, \ddot{q} \in \mathbb{R}^n \) represents the joints’ position, velocity, and acceleration vectors, respectively, the inertia matrix is given by \( M(q) \in \mathbb{R}^{n \times n} \), the Coriolis/centripetal matrix is denoted by \( V_c(q, \dot{q}) \in \mathbb{R}^{n \times n} \), the gravitational vector is represented by \( G(q) \in \mathbb{R}^n \), time-varying unknown external disturbance is denoted by \( \tau_d \in \mathbb{R}^n \), and the control torque is symbolized by \( \tau \in \mathbb{R}^n \). Moreover, the Stribeck friction model can be expressed as [7]:
\[ \mathcal{F}(\dot{q}) = \left[ \beta_0 + \beta_1 e^{-\psi_1|\dot{q}|} + \beta_2 (1 - e^{-\psi_1|\dot{q}|}) \right] \text{sgn}(\dot{q}), \] (11)

where \( \mathcal{F}(\dot{q}) \in \mathbb{R}^n \) represents Stribeck friction force, \( \beta_0 \) denotes the Coulomb friction, \( (\beta_0 + \beta_1) \) represents the Static friction, \( \beta_2 \) is the Viscous friction coefficient, and \( \psi_1 \) and \( \psi_2 \) are positive constants.

3.1. Assumption 1

The \( M(q) \) is a uniformly positive definite matrix such that:
\[ \lambda_1 I \preceq M(q) \preceq \lambda_2 I, \]
where \( \lambda_1 > 0 \) and \( \lambda_2 > 0 \) are constants.

System (10) can be rewritten in the following form as:
\[ \ddot{q} + \alpha^{-1}[M(q) \ddot{q} - \alpha \ddot{q} + V_c(q, \dot{q}) \dot{q} + G(q) + F(q)] + \tau_d = \alpha^{-1} \tau. \] (13)

Then Eq. (13) can further be simplified as:
\[ \ddot{q} = \alpha^{-1} \tau + M(q, \dot{q}, \ddot{q}). \] (14)

where \( \alpha \) is a constant diagonal matrix and:
\[ M(q, \dot{q}, \ddot{q}) = -\alpha^{-1}[M(q) \ddot{q} - \alpha \ddot{q} + V_c(q, \dot{q}) \dot{q} + G(q) + F(q)] + \tau_d. \]

Now, we can represent Eq. (14) in the tracking error form as:
\[ \ddot{\tilde{q}}(t) = \alpha^{-1} \tau + M(q, \dot{q}, \ddot{q}) - \ddot{\tilde{q}}. \] (15)

where \( \tilde{q}(t) = q(t) - q_d(t), \dot{\tilde{q}}(t) = \dot{q}(t) - \dot{q}_d(t), \ddot{\tilde{q}}(t) = \ddot{q}(t) - \ddot{q}_d(t), \) and \( q_d, \dot{q}_d, \ddot{q}_d \in L_\infty \) are the desired trajectory vectors.

4. Controller development

In this section, a model-free controller is proposed using TDE with FOTSBC-based PID (FOTSBC-PID) for the uncertain robotic dynamics under Stribeck friction and external disturbances. Then, the finite-time stability of the closed-loop system is investigated using Lyapunov’s theorem analysis.

4.1. FOTSBC-PID surface

The sliding surface, based on the properties of FO and PID, is designed to obtain a fast response, high
robustness, and finite-time convergence under non-smooth friction. Thus, the terminal sliding manifold is defined as:

\[ S = \ddot{q} + K \int |\dot{q}|^\lambda \text{sgn}(\dot{q}) d\tau, \tag{16} \]

where \( K > 0 \) is a diagonal matrix and \( 1 < \lambda < 2 \).

Once the tracking error reaches the sliding surface \( S = 0(S = 0) \), it guarantees the finite-time convergence of the error \( \ddot{q} \): one gets:

\[ \ddot{q} + K \int |\dot{q}|^\lambda \text{sgn}(\dot{q}) d\tau = 0 \]
\[ \Rightarrow \ddot{q} + K |\dot{q}|^\lambda \text{sgn}(\dot{q}) = 0 \]
\[ \Rightarrow \ddot{q} = -K |\dot{q}|^\lambda \text{sgn}(\dot{q}). \tag{17} \]

Consider the positive-definite Lyapunov candidate as \( V = \frac{1}{2}\ddot{q}^T \ddot{q} \), which has:

\[ \dot{V} = \ddot{q}^T \ddot{q}, \tag{18} \]
\[ \dot{V} = -K |\dot{q}|^\lambda \text{sgn}(\dot{q}) \leq -K \|\ddot{q}\|^\lambda + 1, \tag{19} \]
\[ \dot{V} \leq -K(2\dot{V})^{\frac{\lambda}{\lambda+1}} \leq -K(2)^{\frac{\lambda}{\lambda+1}} (\dot{V}]^{\frac{\lambda}{\lambda+1}} \leq -b\dot{V}. \tag{20} \]

With \( \beta = \frac{\lambda+1}{2} \) and \( b = K^2 \beta \). According to Lemma 2, the finite-time convergence can be calculated as follows:

\[ t_s = \frac{V^1-\beta(t_r)}{b(1-\beta)}, \tag{21} \]

where \( t_s \) is the settling time, and reaching time \( t_r \) will be calculated in Theorem 1.

Using terminal sliding surface Eq. (16), the following FOTSMD-PID sliding surface is developed as:

\[ S_{pid} = K_p S + K_i D^\gamma S + K_d \ddot{S}, \tag{22} \]

where \( K_p, K_i, K_d > 0 \) are PID gains diagonal matrix, and \( 0 < \gamma < 1 \) is FO value.

By taking the derivative of Eq. (22), one can obtain:

\[ \dot{S}_{pid} = K_p \dot{S} + K_i D^\gamma \dot{S} + K_d \ddot{S}. \tag{23} \]

Substituting second derivative of Eq. (16) into Eq. (23) yields:

\[ \dot{S}_{pid} = K_p \dot{S} + K_i D^\gamma \dot{S} + K_d (\ddot{q} + \lambda K |\dot{q}|^{\lambda-1} \dot{q}). \tag{24} \]

Substituting \( \ddot{q}(t) \) from Eq. (15) into Eq. (24), one has:

\[ \dot{S}_{pid} = K_p \dot{S} + K_i D^\gamma \dot{S} + K_d (\ddot{q} + \lambda K |\dot{q}|^{\lambda-1} \dot{q}) \]
\[ -\dddot{q} + \lambda K |\dot{q}|^{\lambda-1} \dot{q} \]. \tag{25} \]

The FOTSMD and PID sliding surfaces are merged, and the proposed sliding surface is developed Eq. (22). Therefore, the FOTSMD-PID surface has the advantage of both schemes, for instance, rapid dynamics response, quick finite-time convergence, small steady-state error, and non-singularity. For the uncertain systems, the mentioned properties are elemental because they are robust against uncertainties and parameter variations, and the system stabilizes swiftly.

**4.2. TDE with FOTSMD-PID-based model-free control design**

In this subsection, the design of the proposed control is presented by combining the schemes such as TDE with the enhanced FOTSMD-PID approach to estimate unknown dynamics of robotic manipulators under frictions and external disturbances. For such a problem of model-free tracking control, the proposed controller is designed as follows:

\[ \tau = \tau_{nom} + \tau_{est} + \tau_{aug}. \tag{26} \]

where \( \tau_{nom} \) is nominal control, and \( \tau_{est} \) is estimation control by TDE, while \( \tau_{aug} \) will be discussed later. Here, \( \tau_{nom} \) and \( \tau_{est} \) are respectively defined in Eqs. (27) and (28) as:

\[ \tau_{nom} = \alpha (\ddot{q}_d - \lambda \dot{K} |\dot{q}|^{\lambda-1} \dot{q} - K^{-1} \nabla^T S - K^{-1} \nabla D^\gamma S), \tag{27} \]
\[ \tau_{est} = -\alpha \dot{M}, \tag{28} \]

where \( \dot{M} \) denotes TDE, which can be computed by Eq. (14) as:

\[ \dot{M} = \nabla (q, \dot{q}, \ddot{q}) |_{t-\psi} = \ddot{q} |_{t-\psi} - \alpha^{-1} \tau |_{t-\psi}. \tag{29} \]

where the constant delay \( \psi \) and then the delayed term \( (t - \psi) \) are obtained.

Thus, by substituting Eq. (26) into Eq. (25), simplified sliding surface is obtained as:

\[ \dot{S}_{pid} = K_d \left[ \alpha^{-1} \tau_{aug} + \dot{\zeta} \right], \tag{30} \]

where \( \dot{\zeta} = \nabla (q, \dot{q}, \ddot{q}) - \dot{\dot{M}} \). Since the estimation error \( \dot{\zeta} \) does not precisely converge to zero due to hard nonlinearities. Thus, to deal with this situation, the\( \tau_{aug} \) is designed by:

\[ \tau_{aug} = -\alpha \left[ K_d \nabla |\dot{S}_{pid} + \zeta | \right], \tag{31} \]

where \( \zeta \equiv \zeta |_{t-\psi} = K_d^{-1} \dot{S}_{pid} - \alpha^{-1} \tau |_{t-\psi} \) is formulated using TDE of Eq. (30) and \( K_d > 0 \).

By substituting Eq. (31) into Eq. (30), one can get:

\[ \zeta = -K_d \text{sgn}(\dddot{S}_{pid}) - \dot{\zeta}, \tag{32} \]

where \( \dot{\zeta} \equiv \dot{\zeta} - \zeta \). The complete proposed model structure is shown in Figure 1.
**Remark 1:** The proposed FOPID-SMC has four major parameters in comparison with conventional SMC. The first parameter, $\mathcal{K}_p$, helps to sustain the properties of the TSMC. The second parameter, $\mathcal{K}_i$, helps to obtain high robustness properties similar to integral SMC. The third parameter, $\mathcal{K}_d$, helps to obtain chatter-free control input. The fourth parameter, the parameter of FO control $\gamma$ greatly improves the response of the system. Moreover, the TDE scheme is used to estimate the unknown dynamics of the system.

**Remark 2:** The TDE scheme is applied for an estimation, which means this technique estimates on the basis of delayed parameters/dynamics. For the implementation of the TDE scheme, a sufficiently small delay is used and can be obtained when the sampling period is selected 30 times faster than the controlled system bandwidth [16,36,37].

**Remark 3:** In the practical implementation of the controller, the acceleration must be known where the position/velocity measurements are available. Thus, the Second-Order Exact Differentiation (SOED) approach can be exploited to estimate the acceleration, which is given as follows:

$$
\begin{align*}
\dot{y}_1 &= -\delta_1 |y_1 - q|^{2/3} \text{sgn}(y_1 - q) + y_2, \\
\dot{y}_2 &= -\delta_2 |y_2 - \dot{y}_1|^{1/2} \text{sgn}(y_2 - \dot{y}_1) + y_3, \\
\dot{y}_3 &= -\delta_3 \text{sgn}(y_3 - \ddot{y}),
\end{align*}
$$

where $y_1 = q$, $y_2 = \dot{q}$, $y_3 = \ddot{q}$ and $\delta_1, \delta_2, \delta_3 > 0$.

5. **Stability analysis**

The stability synthesis of the developed scheme is carried out by the following theorem.

5.1. **Theorem 1**

The states of the unknown system under Stibbe friction (11) converge to zero along the manifold $S_{\text{pid}} = 0$ if the designed controller FOTSMC-PID (26) is applied, then it ensures the finite-time stability and convergence of the system.

Let $\mathcal{V}(t)$ be a Lyapunov candidate selected as:

$$
\mathcal{V}(t) = \frac{1}{2} S_{\text{pid}}^T S_{\text{pid}}. 
$$

By taking the time derivative of $\mathcal{V}(t)$, one obtains:

$$
\dot{\mathcal{V}}(t) = S_{\text{pid}}^T \dot{S}_{\text{pid}}. 
$$

Substitution of Eq. (32) into Eq. (35), one can get:

$$
\dot{\mathcal{V}}(t) = -S_{\text{pid}}^T \left( \mathcal{K}_1 \text{sgn}(S_{\text{pid}}) + \dot{\zeta} \right). 
$$

According to $||S(t)|| = S(t)^T \text{sgn}(S(t))$ yields:

$$
\dot{\mathcal{V}}(t) \leq -\left( \mathcal{K}_1 + \dot{\zeta} \right) ||S_{\text{pid}}||. 
$$

Since the TDE error $\dot{\zeta}$ is bounded by $|\dot{\zeta}| \leq \eta$ with $\eta > 0$ [16] and $\dot{\zeta} \ll \mathcal{K}_1$, one can express the Relation (37) as:

$$
\dot{\mathcal{V}}(t) \leq -\mathcal{K}_1 ||S_{\text{pid}}|| \leq 0. 
$$

Since $\mathcal{K}_1 > 0$, the system (10) will converge to the origin. Hence, the derived Lyapunov analysis shows the
stability of the system is ensured under the proposed control design.

To compute the finite-time convergence, Relation (38) can be expressed as:

\[ \dot{V}(t) \leq -\varrho V^{1/2}(t) \leq 0, \]

with \( \varrho = \sqrt{2K_1} \). Thus, implementing Lemma 2 on Relation (39), the finite-time \( t_r \) can be formulated as:

\[ t_r = \frac{2V^{1/2}(t_0)}{\varrho}. \]

Therefore, the total finite convergence time can be computed using \( t_f = t_r + t_s \) as [38]:

\[ t_f = \frac{2V^{1/2}(t_0)}{\varrho} + \frac{V^{1-\beta}(t_r)}{b(1 - \beta)}. \]

Hence, the tracking error will converge to the origin in a finite-time \( t_f \), and the trajectory will maintain converging to the surface manifold when \( b > 0 \) and \( \varrho > 0 \).

Remark 4: The finite-time and the control torque are dependent on the constant gain \( K_1 \), which is explicitly seen that it is proportional to the \( \tau \) and reciprocal of \( t_f \) in Eqs. (26) and (41), respectively. Thus, better tracking performance, fast finite-time response, and overall dynamic stability can be obtained by selecting the appropriate value of \( K_1 \).

Remark 5: The parameters of the FOTSMC-PID method have been chosen according to the specified range, such as \( K > 0 \), \( K_\varphi > 0 \), \( K_\gamma > 0 \), \( K_\xi > 0 \), \( K_\psi > 0 \), \( 1 < \lambda < 2 \), and \( 0 < \gamma < 1 \). If these parameters are not selected within the given range, then there could be a singularity problem, and the stability of the closed-loop system cannot be achieved. Hence, by selecting the suitable parameters, the desired trajectory tracking and closed-loop system stability can be obtained simultaneously.

6. Simulation evaluations

For the applicability of the theoretical results, the performance of the proposed FOTSMC-PID is validated by implementing 3-DOF dynamics of PUMA 560 robotic manipulators under Stribeck friction. The simulations are performed in the Matlab/Simulink environment with a Runge-Kutta solver under 0.001 sec fixed step size. Moreover, to demonstrate the efficacy of FOTSMC-PID, its results are compared with the proposed scheme without PID (FOTSMC) and Adaptive Fractional-Order Nonsingular TSMC (AFONTSMC) [31].

The dynamics of the considered PUMA 560 manipulators, which were developed in [39], are used. Moreover, the parameters of the proposed and compared controllers are given in Table 1. Initial conditions of joint positions are chosen as \( q_1(0) = q_2(0) = 0.1 \) and \( q_3(0) = 0.05 \), and the parameters of SOED Eq. (33) are chosen as \( \delta_1 = 20 \) and \( \delta_2 = \delta_3 = 5 \). To counteract the chattering problem, the \( \text{sgn} \) function in Eq. (26) is replaced by the \( \tanh \) function. Further, the desired inputs are selected as \( q_{d1} = q_{d2} = 0.2 \cos(0.7t) + 0.2 \cos(0.5t - 0.2) \) and \( q_{d3} = 0.2 \cos(0.5t - 0.2) - 0.2 \cos(0.7t) \).

6.1. Case-1 (without Stribeck friction)

The proposed method is compared with FOTS/MC and AFONTSMC. Thus, simulations of joint position tracking, tracking errors, and control torques without Stribeck friction are depicted in Figures 2–4. Moreover, the Root Mean Square (RMS) results of position errors are illustrated in Table 2.

These results clearly show the high tracking performance of the proposed method in terms of fast convergence speed, quick response, and chatter-free control inputs.

| Controller | Parameters | Values |
|------------|------------|--------|
| FOTSMC-PID | \( K \)   | Diag(30,30,30) |
|            | \( K_p \) | Diag(0,2,0,2,0,2) |
|            | \( K_\xi \) | Diag(0,2,0,2,0,2) |
|            | \( K_\varphi \) | Diag(0,2,0,2,0,2) |
|            | \( \alpha \) | Diag(0.77,0.77,0.77) |
|            | \( \gamma, \psi, \lambda, K_1 \) | 0.9,0.001,1.9,100 |
| FOTSMC     | \( K \)   | Diag(20,20,200) |
|            | \( \gamma \) | 0.1 |
| AFONTSMC   | \( k_1 \) | Diag(50,50,50) |
|            | \( k_2 \) | Diag(10,10,10) |
|            | \( K \)   | Diag(10,10,10) |
|            | \( \alpha \) | 0.1 |

| Controller | \( e_{\text{rms}} \) | \( e_{\text{rms}} \) | \( e_{\text{rms}} \) | \( \sum_{i} x_i^2 \) |
|------------|-----------------|-----------------|-----------------|-----------------|
| FOTSMC-PID | 0.0057 | 0.0036 | 0.0057 | 0.015 |
| FOTSMC     | 0.0029 | 0.0032 | 0.0035 | 0.0096 |
| AFONTSMC   | 0.0077 | 0.0112 | 0.0073 | 0.0262 |
6.2. Case-2 (under striebeck friction)

In this case, comparative analyses of the proposed method with FOTS CMC and AFON TSMC under Striebeck friction are given. Thus, the Striebeck friction parameters are given as $\beta_0 = 22, \beta_1 = 1, \beta_2 = 0.96, \psi_1 = 55$, and $\psi_2 = 50$. The corresponding comparisons of position tracking, tracking errors, control torques,
and RMS position errors under Striebeck friction are depicted in Figures 5–7 and Table 3, respectively.

According to the results of simulations, Figure 5 depicts the actual joint position of the robotic manipulator and precisely tracks the desired trajectory. Figure 7 shows the satisfactory chatter-free control input performance of the proposed method. Therefore, the comparisons of the proposed method with FOTS

\textbf{Figure 4.} Control torque for Adaptive Fractional-Order Nonsingular TSMC (AFONTSMC), Fractional-Order TSMC (FOTS\textsc{mc}), and Fractional-Order TSMC-PID (FOTS\textsc{mc-PID}).

\textbf{Figure 5.} Position Tracking under Striebeck friction for Adaptive Fractional-Order Nonsingular TSMC (AFONTSMC), Fractional-Order TSMC (FOTS\textsc{mc}), and Fractional-Order TSMC-PID (FOTS\textsc{mc-PID}).
Figure 6. Tracking error under Strubeck friction for Adaptive Fractional-Order Nonsingular TSMC (AFONTSMC), Fractional-Order TSMC (FOTSMC), and Fractional-Order TSMC-PID (FOTSMC-PID).

and AFONTSMC show that the performance of all controllers is good without Strubeck friction. As can be seen from Figures 5–7, Strubeck friction considerably affects the dynamics of the robotic manipulator. However, the results under the friction explicitly show that the proposed FOTSMC-PID robustly suppresses the effect of Strubeck friction and obtains effective tracking and fast convergence performances.
Table 3. Comparative tracking error performance under Striebeck Friction for Fractional-Order TSMC-PID (FOTSMC-PID), Fractional-Order TSMC (FOTSMC), and Adaptive Fractional-Order Nonsingular TSMC (AFOTSMC).

| Controller     | $\epsilon_{1RMS}$ | $\epsilon_{2RMS}$ | $\epsilon_{3RMS}$ | $\sum_{i=1}^{3} \epsilon_{i}$ |
|----------------|-------------------|-------------------|-------------------|-------------------------------|
| FOTSMC-PID     | 0.0060            | 0.0048            | 0.0060            | 0.0168                        |
| FOTSMC         | 0.1309            | 0.0043            | 0.2080            | 0.7392                        |
| AFOTSMC        | 0.0098            | 0.0218            | 0.0138            | 0.0454                        |

7. Conclusion

A model-free controller based on Time Delay Estimation (TDE) and Fractional-Order TSMC (FOTSFC) and Proportional-Integral-Derivative (PID) is proposed for robotic manipulators under Striebeck friction. Robust dynamic response and precise trajectory tracking are achieved by FOTSFC-PID, whereas unmodeled uncertain dynamics are estimated by TDE. TDE estimation error is generated because of nonlinear friction, which is compensated by designing augmented torque input. Moreover, the developed scheme is equipped with Second-Order Exact Differentiation (SOED) to estimate joint acceleration, which is impractical to measure. Numerical simulation demonstrates the performance of FOTSFC-PID applied on PUMA 560 manipulators. Moreover, the compared simulation results are illustrated with FOTSFC and Adaptive Fractional-Order Nonsingular Terminal Sliding Mode Control (AFOTSFC), which shows the effectiveness of the proposed scheme.

This paper presents unknown dynamics of the robotic manipulators under Striebeck friction. Therefore, the TDE-based Sliding Mode Control (SMC) approach can be designed to control the system under nonlinearities such as saturation, backlash, hysteresis, and dead-zone.

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