Control of Robot Manipulators Using Time-Delay Estimation and Fuzzy Logic Systems

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Abstract – A highly accurate model-free controller is proposed for trajectory tracking control of robot manipulators. The proposed controller incorporates time-delay estimation (TDE) to estimate and cancel continuous nonlinearities of robot dynamics, and exploits fuzzy logic systems to suppress the effect of the TDE error, which is due to discontinuous nonlinearities such as friction. To this end, integral sliding mode is defined using desired error dynamics, and a Mamdani-type fuzzy inference system is constructed. As a result, the proposed controller achieves the desired error dynamics well. Implementation of the proposed controller is easy because the design of the controller is intuitive and straightforward, and calculations of the complex robot dynamics are not required. The tracking performance of the proposed controller is verified experimentally using a 3-degree of freedom PUMA-type robot manipulator.

Keywords: Robot Manipulators, Time-Delay Estimation (TDE), Fuzzy Logic System (FLS), Time-Delay Control (TDC), Model-free control

1. Introduction

Robot manipulators are highly nonlinear, coupled, complex systems [1, 2]. The tracking control of robot manipulators is one of the most challenging problems for control engineers. Accurate robot dynamic model is required to realize high accuracy-tracking control with model-based controllers [3-5]. The calculations of nonlinear robot models are complex to compute correctly, and even impossible if precise models of the robot manipulators are unknown. Therefore, it is difficult to control using conventional model-based techniques.

To realize model-free control, time-delay estimation (TDE) technique was proposed for control of robot manipulators in the late 1980s [6-8]. The TDE technique assumes that the unknown nonlinear function does not change much for a sufficiently small time period, and estimation and cancelation are realized through the TDE by judiciously using time-delayed information of the state derivatives and control inputs. The so-called time-delay control (TDC) for robot manipulators was proposed with this technique [9, 10]. The key concept of the TDC is canceling unknown nonlinear dynamics of robot and injecting the desired error dynamics.

The TDE technique is simple, efficient, and yet effective.

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It has been applied to many plants: a permanent-magnet synchronous motor [11], a DC-DC boost converter [12], shape memory alloy actuators [13, 14], and chaotic systems [15, 16]. Particularly, it has been also implemented for robot manipulators to perform trajectory tracking control, force control, impedance control, and dual arm cooperative control [17-21].

The TDC for robot manipulators consists of two elements [9, 10]: the robot dynamics canceling element using the TDE, and the desired error dynamics injecting element. Mostly, the TDE is computed from one-sample time delayed information. The TDE functions perfectly for continuous nonlinearities (such as gravity, Coriolis and centrifugal torques, viscous friction, disturbance) in robot dynamics, however, pulse type TDE error occurs for estimation of discontinuous nonlinearities such as Coulomb friction and Stiction [19, 22]. The performance of the closed-loop systems are greatly affected by the TDE error.

To suppress the TDE error, a third element has been introduced, which includes ideal velocity feedback (IVF) [19, 21] and terminal sliding mode (TSM) [22]. The combination use of the TDE and TSM for robot control improves tracking performance of the TDC drastically [22]. The TDE-TSM [22] is a superset of the Hsia's formulation [10], and the controller using the TDE and IVF (TDE-IVF) [19]. When the TSM is activated by fractional power, the TDE-TSM provides faster convergence than the TDE-IVF due to the terminal sliding mode.

The TDE-TSM is not flawless. To implement terminal sliding mode, fractional power function pow (x, γ) , where $0 < \gamma < 1$, is required to calculate. The calculation speed of fractional power function pow (x, γ) is dependent on

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operating systems (Windows, Linux, OS for embedded system), libraries versions, or languages (C, C++, Java, or MATLAB Simulink) [23, 24]. The calculation time is also heavily dependent on the input parameters. For example, the power function sometimes runs over 8,000 times slower than normal case. In the worst-case scenario, it takes dozens of millisecond to calculate the power function. Therefore, many users have developed approximate power function instead of exact power function, as a temporary solution [25-27]. The other shortcoming of the TDE-TSM is the control input chattering, which is highly undesirable. Because the slope of the TSM function is nearly 90 degree in the vicinity of equilibrium point when $0 < \gamma < 1$, control input chattering is invoked by the signum-like function near the equilibrium point.

To realize highly accurate tracking control of robot manipulator and to provide chattering reduction near sliding surface, in this paper, we propose a controller formulation using TDE and fuzzy logic systems. A Mamdani type fuzzy inference system (FIS) [28-30] is constructed by using commonsense rules that refer to indefinite quantities. To author's knowledge, this is the first paper that combines both the TDE technique and the intelligent technique (such as fuzzy logic systems) together to cope with the discontinuous nonlinearities for robot manipulators. The proposed controller consists of three elements: a TDE element, a desired error dynamic injection element, and a fuzzy logic compensator (FLC) to suppress the TDE error. Implementation of the proposed controller is easy because the design of the controller is intuitive and straightforward, and calculations of complex robot dynamics are not required. The proposed controller achieves the desired error dynamics well. The proposed controller has shown the better tracking performance compared with several TDE based controllers. The proposed controller is highly intuitive, efficient, and accurate; thus, it is a good candidate for practical use.

This paper is organized as follows. The next section proposes a high-accuracy tracking control of robot manipulators using the TDE and the FLC. In the third section, the proposed controller is compared experimentally with several TDE based controllers using a PUMA-type robot manipulator. Finally, conclusions are given.

2. Proposed Control Law for Robot Manipulators

2.1 TDC for robot manipulators – Hsia's formulation

$$\mathbf{M}(\mathbf{q})\ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}} + \mathbf{G}(\mathbf{q}) + \mathbf{F}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{\tau}_d = \mathbf{\tau}, \quad (1)$$

where $\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}} \in \mathfrak{R}^n$ represent the position, velocity, and acceleration of the joints, respectively, and $\mathbf{M}(\mathbf{q}) \in \mathfrak{R}^{n \times n}$ stands for the generalized inertia matrix, $\mathbf{C}(\mathbf{q}, \dot{\mathbf{q}}) \in \mathfrak{R}^{n \times n}$ the Coriolis/centripetal matrix, $\mathbf{G}(\mathbf{q}) \in \mathfrak{R}^n$ the gravitational vector, $\mathbf{F}(\mathbf{q}, \dot{\mathbf{q}}) \in \mathfrak{R}^n$ the friction forces, $\boldsymbol{\tau}_d \in \mathfrak{R}^n$ the

disturbance torques, and $\tau \in \Re^n$ the joint torques.

Introducing a positive diagonal matrix, $\overline{\mathbf{M}}$, one can obtain another expression of (1) as follows:

$$\overline{\mathbf{M}}\ddot{\mathbf{q}} + \mathbf{N}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}) = \mathbf{\tau} , \qquad (2)$$

where

$$\mathbf{N}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}) = [\mathbf{M}(\mathbf{q}) - \mathbf{M}]\ddot{\mathbf{q}} + \mathbf{C}(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}} + \mathbf{G}(\mathbf{q}) + \mathbf{F}(\mathbf{q}, \dot{\mathbf{q}}) + \mathbf{\tau}_d. \tag{3}$$

The control objective is to make a robot position vector ${\bf q}$ follow the reference input trajectory ${\bf q}_d$. To this end, we first define ${\bf e}={\bf q}_d-{\bf q}$, $\dot{{\bf e}}=\dot{{\bf q}}_d-\dot{{\bf q}}$, and $\ddot{{\bf e}}=\ddot{{\bf q}}_d-\ddot{{\bf q}}$. The desired error dynamics is defined by

$$\ddot{\mathbf{e}} + \mathbf{K}_D \dot{\mathbf{e}} + \mathbf{K}_P \mathbf{e} = \mathbf{0} , \qquad (4)$$

where $\mathbf{K}_D \in \mathfrak{R}^{n \times n}$, and $\mathbf{K}_P \in \mathfrak{R}^{n \times n}$ are constant diagonal gain matrices. \mathbf{K}_D is a derivative gain, and \mathbf{K}_P is a proportional gain.

The control input can be selected as

$$\tau = \overline{\mathbf{M}}\mathbf{u}_0 + \hat{\mathbf{N}}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}) \tag{5}$$

where

$$\mathbf{u}_0 = \ddot{\mathbf{q}}_d + \mathbf{K}_D \dot{\mathbf{e}} + \mathbf{K}_P \mathbf{e} \tag{6}$$

Here $\hat{N}(q, \dot{q}, \ddot{q})$ can be estimated by the TDE [8-10], as

$$\hat{\mathbf{N}}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}) = \mathbf{N}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}})_{t-1}, \tag{7}$$

where $\hat{\mathbf{N}}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}})$ denotes the estimate of $\mathbf{N}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}})$, \bullet_{t-L} denotes time-delayed value of \bullet , and L is the estimation time delay, which is the sampling period in digital implementation.

From (2), we can obtain

$$\mathbf{N}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}})_{t-L} = \mathbf{\tau}_{t-L} - \overline{\mathbf{M}} \ddot{\mathbf{q}}_{t-L}. \tag{8}$$

Thus, with the combination of (5) - (8), the TDC control law for a robot manipulator is expressed by [10]

$$\boldsymbol{\tau} = \boldsymbol{\tau}_{t-L} - \overline{\mathbf{M}} \ddot{\mathbf{q}}_{t-L} + \overline{\mathbf{M}} (\ddot{\mathbf{q}}_d + \mathbf{K}_D \dot{\mathbf{e}} + \mathbf{K}_P \mathbf{e}). \tag{9}$$

Substituting the control input (9) and (8) into robot dynamics (2) yields the closed-loop dynamics, as

$$\ddot{\mathbf{e}} + \mathbf{K}_{D}\dot{\mathbf{e}} + \mathbf{K}_{P}\mathbf{e} = \overline{\mathbf{M}}^{-1}[\mathbf{N}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}) - \mathbf{N}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}})_{t-L}]. \tag{10}$$

If identity of $N = N_{t-L}$ assumed, the closed loop equation is (4), the desired error dynamics. The well-known stability condition for the TDC is established by Youcef-Toumi and Hsia independently in [31-33], expressed by

$$\left\|\mathbf{I} - \mathbf{M}^{-1} \overline{\mathbf{M}}\right\| < 1. \tag{11}$$

When the closed loop system is stable using the stability criterion (11), the $\bar{\mathbf{N}} - \mathbf{N}_{t-L}$ is bounded because \mathbf{N} is sum of continuous terms and bounded discontinuous terms [18, 19]. The bounded TDE error ε is defined as

$$\boldsymbol{\varepsilon} \triangleq \overline{\mathbf{M}}^{-1}(\mathbf{N} - \mathbf{N}_{t-L}). \tag{12}$$

Then the closed loop system dynamics with the TDC (9) becomes

$$\ddot{\mathbf{e}} + \mathbf{K}_{D}\dot{\mathbf{e}} + \mathbf{K}_{P}\mathbf{e} = \mathbf{\varepsilon}.$$
 (13)

The TDE error ε is close to 0 in most of operating time of robot manipulators; however, it exhibits a pulse-type error due to discontinuity of Coulomb friction at velocity reversal [20].

2.2 Proposed control using the TDE and fuzzy logic

To suppress the effect of the TDE error, ε , we adopted the integral sliding surface (ISS), as

$$\mathbf{s} \triangleq \int_0^t (\ddot{\mathbf{e}} + \mathbf{K}_D \dot{\mathbf{e}} + \mathbf{K}_P \mathbf{e}) dt. \tag{14}$$

Achieving (4) is equivalent to obtain the integral sliding surface s = 0. The control input can be designed as

$$\tau = \overline{\mathbf{M}}\mathbf{u} + \hat{\mathbf{N}}(\mathbf{q}, \dot{\mathbf{q}}, \ddot{\mathbf{q}}), \tag{15}$$

and

$$\mathbf{u} = \mathbf{u}_0 + \mathbf{K} \cdot \mathbf{fuzzy}(\mathbf{s}). \tag{16}$$

Then, with the combination of (7), (8), (15), (16), the proposed control law for a robot manipulator is expressed by

$$\mathbf{\tau} = \mathbf{\tau}_{t-L} - \mathbf{\overline{M}} \ddot{\mathbf{q}}_{t-L} + \mathbf{\overline{M}} [\ddot{\mathbf{q}}_d + \mathbf{K}_D \dot{\mathbf{e}} + \mathbf{K}_P \mathbf{e} + \mathbf{K} \cdot \mathbf{fuzzy}(\mathbf{s})]. \tag{17}$$

In (17), the past acceleration $\ddot{\mathbf{q}}_{t-L}$ is given by numerical differentiation as

$$\ddot{\mathbf{q}}_{t-L} = (\mathbf{q}_t - 2\mathbf{q}_{t-L} + \mathbf{q}_{t-2L})/L^2. \tag{18}$$

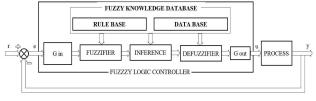


Fig. 1. Fuzzy logic controller

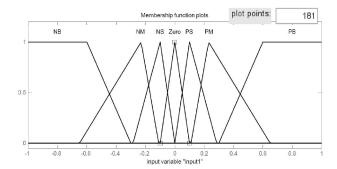


Fig. 2. Fuzzy inference system variables: input

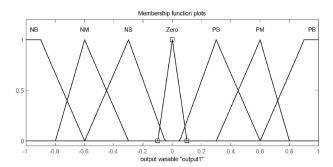


Fig. 3. Fuzzy inference system variables: output

In the proposed control, (17), $\tau_{t-L} - \mathbf{M}\ddot{\mathbf{q}}_{t-L}$ stands for the TDE part, $\overline{\mathbf{M}}(\ddot{\mathbf{q}}_d + \mathbf{K}_D \dot{\mathbf{e}} + \mathbf{K}_P \mathbf{e})$ injects the desired error dynamics, and $\mathbf{K} \cdot \mathbf{fuzzy}(\mathbf{s})$ corrects the TDE error with ISS (14).

Substituting the control input (17), (14) into robot dynamics (2) yields the closed loop dynamics

$$\mathbf{u}_0 - \ddot{\mathbf{q}} + \mathbf{K} \cdot \mathbf{fuzzy}(\mathbf{s}) = \mathbf{\varepsilon} \tag{19}$$

or

$$\dot{\mathbf{s}} + \mathbf{K} \cdot \mathbf{fuzzy}(\mathbf{s}) = \mathbf{\varepsilon}.$$
 (20)

If the fuzzy s-dynamics (21) is asymptotically stable, and ε is bounded, then the close loop dynamics is bounded. Proof of boundedness of ε is given in the Appendix.

$$\dot{\mathbf{s}} + \mathbf{K} \cdot \mathbf{fuzzy}(\mathbf{s}) = \mathbf{0}. \tag{21}$$

Generally, whether a fuzzy control design, shown in Fig. 1, will be stable is a somewhat open question. Structure of a fuzzy controller is determined by the components such as fuzzy sets, fuzzy rules, fuzzy logic operators, the fuzzy inference method, and the defuzzifier. The structure of a fuzzy controller will be different if different types of the components are used. There are various ways to design an asymptotically stable fuzzy logic controller [34, 35]. In this paper, we used a Mamdani-type fuzzy inference system (FIS) to construct an asymptotically stable fuzzy

s-dynamics (21). The Matlab fuzzy logic toolbox is used to design the FIS intuitively.

The fuzzy rules are as follows:

Rule1: If input is Zero then output is Zero.

Rule2: If input is negative big (NB) then output is NB.

Rule3: If input is negative medium (NM) then output is NM

Rule4: If input is negative small (NS) then output is NS.

Rule5: If input is positive big (PB) then output is PB.

Rule6: If input is positive small (PM) then output is PM.

Rule7: If input is positive small (PS) then output is PS.

Input/output functions are mapped into fuzzy numbers using triangular and trapezoidal membership functions (Figs. 2, 3). In Fig. 2, the width of the Zero set membership function is relatively narrow compared with other membership functions, in order to increase control accuracy in the vicinity of the Zero. To obtain the fuzzy surface, singleton fuzzifier minimum inference engine and centroid defuzzifier are used.

Triangle-shaped membership function and trapezoidalshaped membership function are used for input and output membership functions for the FIS, given by

$$f(x;a,b,c) = \begin{cases} 0, & x \le a \\ (x-a)/(b-a), & a \le x \le b \\ (c-x)/(c-b), & b \le x \le c \\ 0, & c \le x \end{cases}$$
 and
$$f(x;a,b,c,d) = \begin{cases} 0, & x \le a \\ (x-a)/(b-a), & a \le x \le b \\ 1, & b \le x \le c \\ (d-x)/(d-c), & c \le x \le d \\ 0, & d \le x \end{cases}$$

And the input membership functions in Fig. 2 are expressed as follows:

$$\mu_{NB}(x) = f(x; -1.5, -1, -0.6, -0.3),$$

$$\mu_{NM}(x) = f(x; -0.65, -0.23, -0.11),$$

$$\mu_{NS}(x) = f(x; -0.285, -0.1, 0),$$

$$\mu_{ZERO}(x) = f(x; -0.1, 0, 0.1),$$

$$\mu_{PS}(x) = f(x; 0.11, 0.285),$$

$$\mu_{PM}(x) = f(x; 0.11, 0.23, 0.65),$$

$$\mu_{PB}(x) = f(x; 0.3, 0.6, 1, 1.5).$$

The output membership functions in Fig. 3 are expressed as follows:

$$\mu_{NB}(x) = f(x; -1.5, -1, -0.9, -0.6),$$

$$\mu_{NM}(x) = f(x; -0.8, -0.6, -0.3),$$

$$\mu_{NS}(x) = f(x; -0.6, -0.3, -0.05),$$

$$\mu_{ZERO}(x) = f(x; -0.1, 0, 0.1),$$

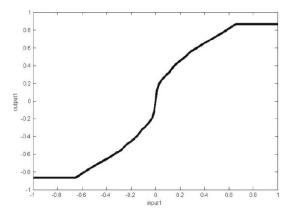


Fig. 4. Control surface of the fuzzy inference system

$$\mu_{PS}(x) = f(x; 0.05, 0.3, 0.6),$$

$$\mu_{PM}(x) = f(x; 0.3, 0.6, 0.8),$$

$$\mu_{PB}(x) = f(x; 0.6, 0.9, 1, 1.5).$$

The control surface of the FIS, depicted in Fig. 4, is an odd function because of the symmetric property of the input/output functions and the use of centroid defuzzifier. Generally, a control surface of FIS is not always an odd function, however, one can carefully select the parameters of input/output membership function to make the control surface function be an odd function; thus, it becomes $s_i \cdot fuzzy(s_i) > 0$. If we consider a Lyapunov function as $V = 0.5s^T s$, then $\dot{V} = \sum \dot{s}_i s_i = -\sum s_i \cdot k_i$ fuzzy(s_i) < 0, which assures asymptotic stability of the fuzzy s-dynamics (21). The convergent speed is faster in the vicinity of the zero than in the distance of the zero.

3. Experimental Studies

A PUMA-type robot, Samsung Faraman AT2 $^{\oplus}$, was used for the experiments (Fig. 5). The maximum payload of the robot is 3 kg. The maximum continuous torques are 0.637, 0.637, and 0.319 Nm for joints 1, 2, and 3, respectively. The gear reduction ratio of each joint is 120:1, the encoder



Fig. 5. A PUMA-type robot manipulator

resolution of each joint is 8192 pulses/rev. Resolution of each robot joint is 3.66×10^{-4} deg.

The control objective is to rotate each joint from the initial position (0°) to the first desired position (30°), and then return to the initial position for $0s \le t < 10s$; this process is repeated for the second desired position $10s \le t < 20s$. The desired trajectory is generated using the fifth polynomial method (Fig. 6). The sampling time is given as L=2 ms.

The desired dynamics error was chosen $\ddot{\mathbf{e}} + 20\dot{\mathbf{e}} + 100\mathbf{e} = \mathbf{0}$; thus, $k_D = 20$, and $k_P = 100$ for each joint. The gain $\overline{\mathbf{M}} = \text{diag}(0.789, 0.713, 0.357)Kg \cdot m^2$ for joints 1, 2, and 3.

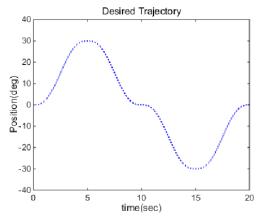


Fig. 6. Desired trajectory for experiments

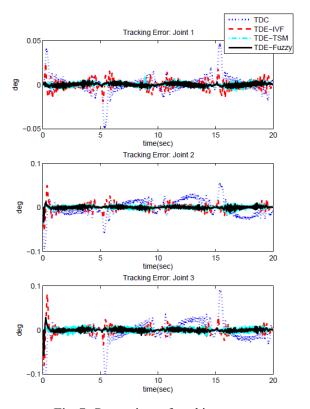


Fig. 7. Comparison of tracking errors

The proposed controller (TDE-Fuzzy) has been compared with the Hsia's TDC control law (9) [10], the TDE-IVF (22) [19], and the TDE-TSM (23) [22].

$$\boldsymbol{\tau} = \boldsymbol{\tau}_{t-L} - \overline{\mathbf{M}} \ddot{\mathbf{q}}_{t-L} + \overline{\mathbf{M}} (\ddot{\mathbf{q}}_d + \mathbf{K}_D \dot{\mathbf{e}} + \mathbf{K}_P \mathbf{e} + \mathbf{K} \cdot \mathbf{s}). \tag{22}$$

$$\boldsymbol{\tau} = \boldsymbol{\tau}_{t-L} - \overline{\mathbf{M}} \ddot{\mathbf{q}}_{t-L} + \overline{\mathbf{M}} [\ddot{\mathbf{q}}_d + \mathbf{K}_D \dot{\mathbf{e}} + \mathbf{K}_P \mathbf{e} + \mathbf{K} \cdot \mathbf{sig}(\mathbf{s})^{\gamma}]. \quad (23)$$

The additional gains for the TDE-IVF and the TDE-TSM are $\mathbf{K} = \text{diag}(10.0, 15..0, 20.0)$ and $\gamma = \text{diag}(0.4, 0.4, 0.4)$.

Fig. 7 shows the tracking errors of the TDC, TDC-IVF, TDE-TSM, and the proposed TDC-Fuzzy controller. The proposed controller shows the smallest size of tracking error among the four controllers. Using a carefully designed fuzzy logic controller to suppress the effect of the TDE error can improve the tracking performance. In Fig. 8, the proposed controller shows the smallest magnitude of strajectories among the four controllers, showing good agreement with the tendency of tracking errors shown in Fig. 7. Because tracking error e is a filtered version of s. bounds on s can be directly translated into bounds on the tracking error. The experimental results confirm that the best tracking performance, as well as s-trajectory tracking, can be achieved through the proposed fuzzy logic based controller. The control input of TDE-TSM (when $0 < \gamma < 1$) shows larger chattering compared with the proposed controller (Fig. 9).

To explain the superiority of the proposed controller to the TDC-IVF and the TDE-TSM, graphical comparisons

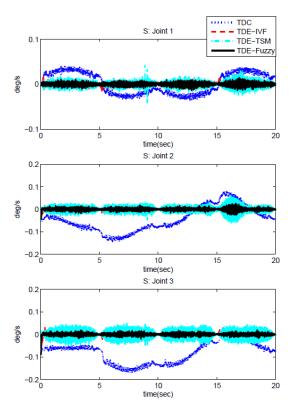


Fig. 8. Sliding variables

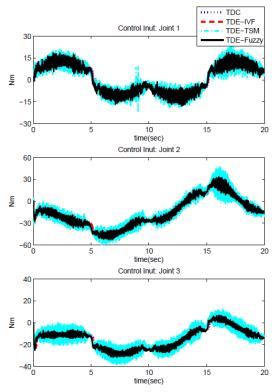


Fig. 9. Control Inputs

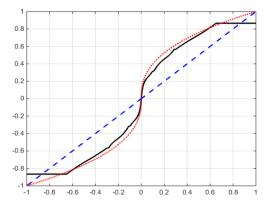


Fig. 10. Graphical comparison of functions: FIS function (black solid), $sig(s)^{0.4}$ (red dot), and s (blue dashed)

of the FIS function, the TSM function, and the linear functions are conducted in Fig. 10 and Fig. 11. The slope of the FIS function and the TSM function are increasing near the equilibrium point, while the slope of the linear function remains constant. Therefore, the tracking error of the TDE-TSM and TDE-fuzzy is smaller than the TDC-IVE.

The slope of the TSM function near the 0 is almost 90 degree, and its shape is almost like signum function which can possibly cause high frequency control chattering (Fig 11). It is well known that the control input chattering harms the tracking performance. In contrast, the slope of the FIS is much smaller than 90° (Fig 11); thus, chattering

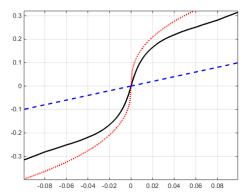


Fig. 11. Magnified graphical comparison of functions: FIS function (black solid), $sig(s)^{0.4}$ (red dot), and s (blue dashed)

is reduced in the proposed controller. Consequently, the tracking performance of the proposed controller is better than that of the TDE-TSM with less control input chattering.

4. Conclusion

In this paper, fuzzy logic is first combined with the TDE technique to suppress the TDE. The proposed controller consists of three parts: the TDE part to estimate and cancel continuous nonlinearities of robot dynamics, the injection part to endow desired dynamics, and the correcting part using integral sliding mode and Mamdani-type FIS. The Mamdani-type FIS is constructed by using common sense rules that refer to indefinite quantities. The proposed controller with well-designed FIS functions is highly intuitive, efficient, and accurate. Thanks to the TDE part, the proposed controller preserves the simplicity and efficiency of TDC. Compared with the TDE-TSM control, better tracking performance is achieved using the proposed controller, while control chattering is reduced significantly. The proposed controller, with the help of FLC, could be a good replacement of existing TDE based controllers to achieve good tracking performance and chattering reduction. Fuzzy Logic Systems used in this paper is simple and standard. One can expect that more sophisticated intelligent systems (such as artificial neural networks (ANN), fuzzy ANN, and neuro-fuzzy systems) can further enhance the control performance of the proposed controller. More researches in the near future, based on this framework, are welcome.

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Appendix: Stability

The boundedness of ε can be proved in the same manner of the stability proof in [18-20].

Using (16), (19) gives

$$\mathbf{\varepsilon} = \mathbf{u} - \ddot{\mathbf{q}}.\tag{24}$$

A combination of (5), (8), (16), (17), and (24) gives

$$M\varepsilon = M(\mathbf{u} - \ddot{\mathbf{q}})$$

$$= M\mathbf{u} + C\dot{\mathbf{q}} + \mathbf{G} + \mathbf{F} + \boldsymbol{\tau}_d - \boldsymbol{\tau}$$

$$= M\mathbf{u} + C\dot{\mathbf{q}} + \mathbf{G} + \mathbf{F} + \boldsymbol{\tau}_d - \overline{\mathbf{M}}\mathbf{u} - \mathbf{N}_{t-L}.$$
(25)

From (3), the delayed nonlinear term is given by

$$\mathbf{N}_{t-L} = [\mathbf{M}_{t-L} - \bar{\mathbf{M}}] \ddot{\mathbf{q}}_{t-L} + (\mathbf{C}\dot{\mathbf{q}})_{t-L} + \mathbf{G}_{t-L} + \mathbf{F}_{t-L} + (\mathbf{\tau}_d)_{t-L}.$$
(26)

Substitute (26) into(25), we have

$$\mathbf{M}\boldsymbol{\varepsilon} = (\mathbf{M} - \overline{\mathbf{M}})\mathbf{u} - (\mathbf{M}_{t-L} - \overline{\mathbf{M}})\ddot{\mathbf{q}}_{t-L} + \boldsymbol{\Delta}, \tag{27}$$

where

$$\Delta = \mathbf{C}\dot{\mathbf{q}} + \mathbf{G} + \mathbf{F} + \mathbf{\tau}_{d} - (\mathbf{C}\dot{\mathbf{q}})_{t-L} - \mathbf{G}_{t-L} - \mathbf{F}_{t-L} - (\mathbf{\tau}_{d})_{t-L}. (28)$$

It is clear that Δ is bounded for a sufficiently small L. Substituting $\ddot{\mathbf{q}}_{t-L} = \mathbf{u}_{t-L} - \boldsymbol{\varepsilon}_{t-L}$ from (24) yields

$$\mathbf{M}\boldsymbol{\varepsilon} = (\mathbf{M} - \overline{\mathbf{M}})\mathbf{u} - (\mathbf{M} - \overline{\mathbf{M}})\ddot{\mathbf{q}}_{t-L}$$

$$+ (\mathbf{M} - \mathbf{M}_{t-L})\ddot{\mathbf{q}}_{t-L} + \boldsymbol{\Delta}$$

$$= (\mathbf{M} - \overline{\mathbf{M}})\mathbf{u} - (\mathbf{M} - \overline{\mathbf{M}})(\mathbf{u}_{t-L} - \boldsymbol{\varepsilon}_{t-L})$$

$$+ (\mathbf{M} - \mathbf{M}_{t-L})\ddot{\mathbf{q}}_{t-L} + \boldsymbol{\Delta}$$

$$= (\mathbf{M} - \overline{\mathbf{M}})\boldsymbol{\varepsilon}_{t-L} + (\mathbf{M} - \overline{\mathbf{M}})(\mathbf{u} - \mathbf{u}_{t-L})$$

$$+ (\mathbf{M} - \mathbf{M}_{t-L})\ddot{\mathbf{q}}_{t-L} + \boldsymbol{\Delta}$$

$$(29)$$

Therefore, ε is given by

$$\mathbf{\varepsilon} = \mathbf{E}\mathbf{\varepsilon}_{t-L} + \mathbf{E}\mathbf{\eta}_1 + \mathbf{\eta}_2, \tag{30}$$

where

$$\mathbf{E} = \mathbf{I} - \mathbf{M}^{-1} \overline{\mathbf{M}} \tag{31}$$

$$\mathbf{E} = \mathbf{I} - \mathbf{M}^{-1} \overline{\mathbf{M}} \mathbf{\eta}_1 = \mathbf{u} - \mathbf{u}_{t-1}$$
 (32)

$$\mathbf{\eta}_2 = \mathbf{M}^{-1} [(\mathbf{M} - \mathbf{M}_{t-L}) \ddot{\mathbf{q}}_{t-L} + \Delta]. \tag{33}$$

For a sufficiently small time delay L, η_1 , and η_2 are bounded.

In the discrete-time domain (30) can be represented as

$$\mathbf{\varepsilon}(k) = \mathbf{E}(k)\mathbf{\varepsilon}(k-1) + \mathbf{E}(k)\mathbf{\eta}_1(k) + \mathbf{\eta}_2(k). \tag{34}$$

Therefore, if $\|\mathbf{E}\| < 1$, then roots of $\mathbf{E}(k)$ reside inside a unit circle, and (34) is asymptotically bounded with bounded forcing function $\mathbf{\eta}_1$, and $\mathbf{\eta}_2$.



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