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An Minimum-Energy-Based High-Degree Polynomial Trajectory Planning and Tracking Control for an LCD Glass-handling Robot

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Abstract: A new method of minimum-energy point-to-point (PTP) trajectory planning is proposed for an LCD glass-handing robot, which is driven by a permanent magnet synchronous motor (PMSM). The variable structure controller (VSC) is designed to track the trajectories and compensate LuGre model of frictional torque effects. The robot system is described by a mechanical equation and an electrical equation. To generate the minimum-energy trajectory, we employ a high-degree polynomial with suitable end-point conditions. The real-coded genetic algorithm (RGA) is used to search for the coefficients of the polynomial with the fitness function of minimum-energy input. Finally, numerical simulations of the minimum-energy inputs are compared for various degrees of polynomials. It is concluded that the proposed methodology can effectively design the minimum-energy trajectory, and the nonlinear VSC can track the designed trajectories for the robot system driven by a PMSM.

Keywords: LCD glass-handling robot; minimum-energy; permanent magnet synchronous motor (PMSM); real-coded genetic algorithm (RGA); trajectory planning

1. Introduction

The subject about trajectory planning has been extensively studied. The problem of robot trajectory planning is a very complex task, and plays a crucial role in design and application of robots in task space. The robot trajectory planning can be accomplished through the way of generating and optimizing ruled surfaces under constraints of kinematics, dynamics and control performances. Based on an algorithm which is capable of obtaining a sequence of feasible robot configurations, a path between the given initial and the goal configurations to be reached is obtained [1]-[3]. Biagiotti and Melchiorri [4] developed the trajectory planning of the computation of desired motion pro-

files for the actuation systems of automatic machines.

Huang et al. [5] have proposed a novel minimumenergy point-to-point (PTP) trajectory planning method for a motor-toggle servomechanism. In order to generate a robot's trajectory, the genetic algorithm (GA) was developed to search optimal solutions for the trajectories [6, 7]. The dynamic formulation of an LCD glass-handling robot driven by a PMSM was formulated by Hamilton's principle [8, 9], where the identification and tracking control were also performed for the system. A survey [10] of controlling ac machines for high dynamic performance including different types of machines and converters with their specific control characteristics was performed.

Canudas de Wit et al. [11] proposed a dynamic fric-

tional model that combines the stiction behavior, i.e., the Dahl effect, with arbitrary steady-state frictional characteristics which include Stribeck effect. Frictional force is a nonlinear behavior and can cause some difficult problems such as static errors, limit cycles and stick-slip in mechanical systems [12]. The variable structure controller (VSC) can enhance the robustness under plant disturbances and uncertainties [13, 14]. One of the most recent topics in the variablestructure system theory is represented by the secondorder sliding-mode control methodology [13]. Lee et al. [14] proposed an integral sliding-mode control method to obtain zero steady-state error under an external step disturbance and to show its robust control with respect to the parametric variations of the input channel.

In this paper, we adopt the RGA to obtain the polynomial coefficients with a fitness function of minimum-energy input for the PTP motion profile of an LCD glass-handling robot. In this paper, the mathematical model includes the electrical and mechanical equations. The VSC is designed to compensate the frictional torque. Various degrees of polynomials with the minimum-energy inputs are compared in numerical simulations. It is found that the proposed methodology can effectively design the minimum-energy trajectory, and the VSC can track the minimum-energy trajectory for the robot system driven by a PMSM.

2. Dynamic equations of the mechatronic syste

In order to completely formulate the dynamic equations, we use the electrical and mechanical equations to describe the LCD glass-handling robot driven by a PMSM.

For the electrical equation, we adopt [10] and have

$$v_q = R_s i_q + \frac{d}{dt} \lambda_q + \omega_s \lambda_d \tag{1}$$

where v_q is the stator voltage, R_s is the stator resistance, i_q is the stator current, λ_q is the stator flux linking, ω_s is the inverter frequency and λ_d is the stator flux linking.

In Figure 1, the mechatronic system consists of a PMSM, two timing belt drives and three arms. An LCD glass-handling robot is driven with a maximum travel distance 260 mm in the X_1 direction by a PMSM. The principle of the motion is that the PMSM drives Arm 1 at point O_1 to convert into rotation of Arm 1 and then the timing belt 1 curl Gear 1 to pull rotation of Gear 2. At this time, Arm 2 also rotates along with

rotation of Gear 2 and then the timing belt 2 curl Gear 4 to pull rotation of Gear 3.

In order to simplify the dynamic model, we assume that (A) the timing belts have no difference in slipping in the driving and driven gears, (B) the timing belts are rigid (i.e. the stiffness of timing belts are infinite), and (C) the motor shaft and all of the gears and arms are rigid. Here, one degree-of-freedom dynamic model was formulated by Hamilton's principle [8] as follows

$$[J_0 - J_1 \cos(n\phi_1) + J_m] \ddot{\phi}_1 + B_m \dot{\phi}_1 + \mu \sin(n\phi_1) \dot{\phi}_1^2 = K_\tau i_a,$$
 (2)

where

$$J_{0} = I_{1} + (n-1)^{2}I_{2} + (1-2n)^{2}I_{3} + (n-1)^{2}I_{4} + 4I_{5}$$

$$+ (1-2n)^{2}I_{6} + m_{2}[l_{1}^{2} + (n-1)^{2}b^{2}]$$

$$+ m_{3}[l_{1}^{2} + (n-1)^{2}l_{2}^{2}] + m_{4}l_{1}^{2} + m_{5}l_{1}^{2}$$

$$+ m_{6}[l_{1}^{2} + (n-1)^{2}l_{2}^{2}],$$

$$J_{1} = 2(n-1)[m_{2}b + (m_{3} + m_{6})l_{2}]l_{1},$$

$$\mu = n(n-1)[m_{2}b + (m_{3} + m_{6})l_{2}]l_{1}.$$

 ϕ_1 , $\dot{\phi}_1$, and $\ddot{\phi}_2$ are the angular displacement, angular speed and angular acceleration, respectively. J_m is the moment of inertia of the rotor, B_m is the damping coefficient, K_t is the motor torque constant of the PMSM, and the detailed notations can also be seen in [8]. Equation (2) is a nonlinear ordinary differential equation, and the coefficients of the moment inertial and the nonlinear damping are functions of the rotor angle.

Two kinematic equations [8] of the geometric constraints in the X_1 and Y_1 directions can be described as

$$X_1 = l_1 \cos \phi_1 + l_2 \cos(\phi_1 - n\phi_1), \tag{3}$$

$$Y_1 = l_1 \sin \phi_1 + l_2 \sin(\phi_1 - n\phi_1). \tag{4}$$

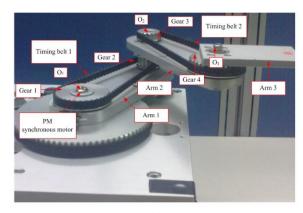
where n is the radius ratio of γ_1/γ_2 . Equations (3) and (4) are the relationships between the translations of point O_3 and rotation of Arm 1.

3. Point-to-point trajectory design

The minimum-energy PTP trajectory [4], [5] is designed on the basis of a high-degree polynomial, which is able to satisfy desired constraints of angular displacement, speed, acceleration and jerk at the start time t_0 and end time T. From a mathematical point of view, the problem is then to find a function such that

$$\phi_l = \phi_1^*(t), t \in [t_0, T], \tag{5}$$

where the superscript * indicates the designed trajectory, $\phi_1^*(t)$ and is the designed angular displacement of Arm 1.



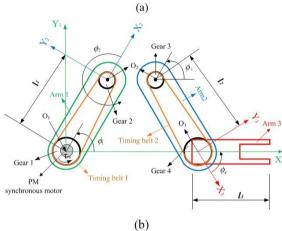


Figure 1 The LCD glass-handling robot driven by a PMSM. (a) The experimental equipment. (b) The geometrical mode.

3.1 Polynomial trajectories

The trajectory planning problem can be easily solved by considering a polynomial function as

$$\phi_l^*(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3 + \dots + a_m t^m,$$
 (6)

where each coefficient a_i , $i = 0, 1, \dots, m$, is a real number; $a_m \neq 0$, and m is a positive integer. There are m+1 coefficients to be determined such that the constraints are satisfied at the start and end time. The degree m of the polynomial depends on the number of constraints to be satisfied and on the desired smoothness of the resulting motion.

Considering a polynomial of degree twelve, and the angular displacement, speed, acceleration and jerk are zero at the start time, we have

$$\phi_1^*(0) = a_0 = 0, \tag{7}$$

$$\dot{\phi}_1^*(0) = a_1 = 0, \tag{8}$$

$$\ddot{\phi}_1^*(0) = a_2 = 0, \tag{9}$$

$$\dot{\phi}^* \text{H1}(0) = a_3 = 0. \tag{10}$$

The angular displacement is Φ , and the angular speed, acceleration and jerk are zero at the end time T, thus we have

$$\phi_1^*(T) = \Phi, \tag{11}$$

$$\dot{\phi}_1^*(T) = 0, \tag{12}$$

$$\ddot{\phi}_1^*(T) = 0, \tag{13}$$

$$\dot{\ddot{\phi}}_{1}^{*}(T) = 0. \tag{14}$$

where Φ is the maximum angular displacement. In this paper, we have $\Phi = 1.57 \ rad$, which is obtained from (3) with the maximum distance $X_l = 260mm$ in the X_l direction of the LCD glass-handling robot. Substituting the eight conditions (7)-(14) into the twelve-degree polynomial (6), we have the four analytic solutions

$$a_4 = 35\frac{\Phi}{T^4} + T^4 a_8 + 4T^5 a_9 + 10T^6 a_{10} + 20T^7 a_{11} + 35T^8 a_{12},$$
 (15)

$$a_5 = -84 \frac{\Phi}{T^5} - 4T^3 a_8 - 15T^4 a_9 - 36T^5 a_{10}$$

$$-70T^6 a_{11} - 120T^7 a_{12},$$
(16)

$$a_6 = 70\frac{\Phi}{T^6} + 6T^2a_8 + 20T^3a_9 + 45T^4a_{10} + 84T^5a_{11} + 140T^6a_{12},$$
(17)

$$a_7 = -20\frac{\Phi}{T^7} - 4Ta_8 - 10T^2a_9 - 20T^3a_{10}$$

$$-35T^4a_{11} - 56T^5a_{12}.$$
(18)

The coefficients $a_4 - a_7$ are to be determined once $a_8 - a_{12}$ are known. The coefficients $a_8 - a_{12}$ in (15)-(18) for the twelve-degree polynomial trajectory are to be determined by the following RGA method.

3.2 Input energy for the PTP motion

The whole PMSM may be considered thermodynamically as an energy converter. It takes electrical input energy from control command and then outputs the mechanical work to drive the Arm. The absolute value of input energy to the PMSM during the whole time *T* is defined as

$$E(z) = \int_0^T |i_q^* v_q^*| dt,$$
 (19)

where

$$i_q^* = \frac{l}{K_t} ([J_0 - J_1 \cos(n\phi_1^*) + J_m] \ddot{\phi}_1^* + B_m \dot{\phi}_1^* + \mu \sin(n\phi_1^*) \dot{\phi}_1^{*2})$$

$$(20)$$

$$v_{q}^{*} = R_{s}i_{q}^{*} + L_{q}\frac{d}{dt}i_{q}^{*} + p\lambda_{d}\dot{\phi}_{l}^{*}.$$
 (21)

The i_q^* is electric current, and can be obtained by the designed trajectory ϕ_1^* , $\dot{\phi}_1^*$ and $\ddot{\phi}_1^*$ from (2). The v_q^* is the control command, and can be obtained by substituting i_q^* and $\dot{\phi}_1^*$ into (21). Therefore, it can be shown that the absolute value of input energy in (19) can be obtained once the designed trajectory is obtained.

4. The real-coded genetic algorithm

The RGA is an optimization searching algorithm, which simulates evolution mechanism on a computerbased platform in conjunction with natural selection and genetic mechanism. The chromosomes are expressed by vectors and each element of vectors is called a gene. The initial real-valued genes in the chromosomes are gotten through generating a sequence of real-valued variable by a randomly limited range. All chromosomes compose of a population and are evaluated according to the pre-given evaluation index and given fitness values. According to the fitness values, the reproduction, crossover and mutation operations are carried out among or between the chromosomes. The chromosomes having less fitness values are given up, and the chromosomes having larger fitness values are left. Those left chromosomes compose of a new population, which maybe better than the old.

4.1 Fitness function

The RGA continuously searches the better chromosomes in this way until the converging index is satisfied. By using the RGA merit, we can obtain the unknown coefficients a_8 - a_{12} in (15)-(18).

The RGA procedure is shown in Figure 2, where the reproduction procedure adopts the roulette wheel selection, and the crossover and uniform mutation adopt the methods as in [6, 7]. How to define the fitness function is the key point of the genetic algorithm, since the fitness function is a figure of merit, and could be computed by using any domain knowledge.

In this paper, the fitness function f(z) in the RGA method is defined as

$$f(z) = \frac{1}{E(z)} \tag{22}$$

$$z = [a_8 \ a_9 \ a_{10} \ a_{11} \ a_{12}] \tag{23}$$

where z is a parameter vector of polynomial coefficients, and E(z) is the total absolute input energy.

4.2 Increasing function

During the PTP motion profile of the mechanical system, the trajectory for the angular displacement is required to be designed as a monotonously increasing function from the start to the end points. Suppose that $\phi_1^*(t)$ is the designed PTP motion profile defined during $0 \le t \le T$ for $t_i < t_{i+1}$, the monotonously increasing function is defined as

$$\phi_1^*(t_i) < \phi_1^*(t_{i+1}) \tag{24}$$

where the subscript i stands for the i^{th} sampling time.

This constraint for a monotonously increasing function is inserted in the procedure of the RGA method as shown in Figure 2. Without using this constraint (24), the identified coefficients of polynomials are not unique in the numerical simulations by the RGA method [5].

5. Variable Structure Controller Design

In order to achieve good tracking control with the minimum-energy input (19) even if external disturbances exist, we utilize the VSC to obtain the best control performance in this paper.

5.1 Frictional model

To consider the frictional torque, we adopt the Lu-Gre model [11, 12], which contains a state variable representing the average deflection of elastic bristles, which are visualized on the contacting surfaces. The resulting model faithfully captures the majority of known frictional behaviors, such as the pre-sliding displacement, frictional lag, varying break-away torque and stick-slip motion.

The frictional torque of LuGre model [11] is defined as

$$\tau_f = \sigma_0 z + \sigma_1 \frac{dz}{dt} + \sigma_2 \dot{\phi}_1 \tag{25}$$

where σ_0 is the stiffness, σ_1 is a damping and σ_2 is a viscous frictional coefficient. The average deflection of the bristles is denoted by z and is modeled by

$$\frac{dz}{dt} = \dot{\phi}_1 - \frac{|\dot{\phi}_1|}{g(\dot{\phi}_1)}z\tag{26}$$

where $\dot{\phi}_1$ is the relative velocity between the two surfaces. On the right-hand side, the first term gives a deflection that is proportional to the integral of the relative velocity. The second term asserts that the deflection z approaches the value,

$$z = \frac{\dot{\phi}_1}{|\dot{\phi}_1|} g(\dot{\phi}_1) = g(\dot{\phi}_1) sign(\dot{\phi}_1)$$
 (27)

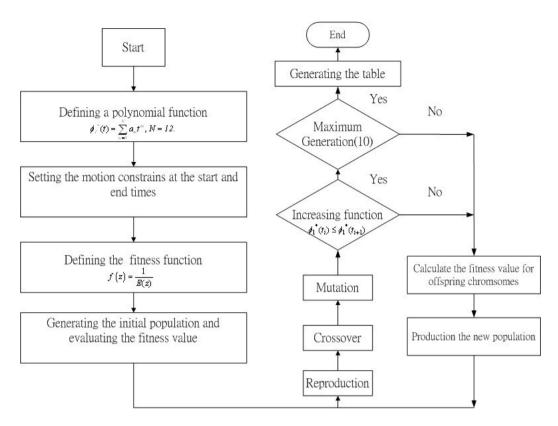


Figure 2 Flow chart of the RGA procedure

for a steady state, as $\dot{\phi}_1$ is constant. The function $g(\dot{\phi}_1)$ is positive and depends on many factors such as material properties, lubrication and temperature. It need not be symmetrical, and direction dependent behavior can therefore be captured. For a typical bearing friction, $g(\dot{\phi}_1)$ will monotonically decrease from g(0) when $\dot{\phi}_1$ increases. A parameterization of $g(\dot{\phi}_1)$ has been proposed to describe the Stribeck effect as

$$\sigma_0 g(\dot{\phi}_1) = \tau_c + (\tau_s - \tau_c) e^{-(\frac{\dot{\phi}_1}{\dot{\phi}_s})^2}$$
 (28)

where τ_c is the Coulomb frictional torque, τ_s is the stiction torque and $\dot{\phi}_s$ is the Stribeck angular velocity. It follows from (25), (27) and (28) that for the steady-state motion the relation between angular velocity and frictional torque can be given by

$$\tau_f = \sigma_0 g(\dot{\phi}_1) sign(\dot{\phi}_1) + \sigma_2 \dot{\phi}_1$$

$$= \tau_c sign(\dot{\phi}_1) + (\tau_s - \tau_c) e^{-(\frac{\dot{\phi}_1}{\dot{\phi}_s})^2} sign(\dot{\phi}_1) \quad (29)$$

$$+ \sigma_2 \dot{\phi}_1$$

Dynamic Equation (2) can be rewritten by adding the frictional torque (29) as

$$[J_0 - J_1 cos(n\phi_1) + J_m] \ddot{\phi}_1 + B_m \dot{\phi}_1 + \mu sin(n\phi_1) \dot{\phi}_1^2 = K_t i_a - \tau_f(30)$$
(30)

5.2 Sliding function

Combining the electrical equation (1) for the PMSM, and the dynamic equation (2) for the mechanical system, and using the state variable $x = [\phi_1 \ \dot{\phi}_1 \ i_q]^T$, we obtain the dynamic state equation for the LCD glasshandling robot as follows

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u} + \mathbf{D} \tag{31}$$

where
$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & \frac{-B_m - \mu sin(n\phi_1)\dot{\phi}_1}{J_0 - J_1 cos(n\phi_1) + J_m} & \frac{K_t}{J_0 - J_1 cos(n\phi_1) + J_m} \\ 0 & -\frac{p\dot{\lambda}_d}{L_q} & -\frac{R_s}{L_q} \end{bmatrix},$$

$$B = \begin{bmatrix} 0 \\ 0 \\ \frac{1}{L_q} \end{bmatrix}, D = \begin{bmatrix} 0 \\ -\frac{\tau_f}{J_0 - J_1 cos(n\phi_1)J_m} \\ 0 \end{bmatrix}, u = v_q$$

To design a VSC [14], we decompose (31) as

$$\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} u \quad (32)$$

where
$$x_1 = \begin{bmatrix} \phi_1 \\ \dot{\phi}_1 \end{bmatrix}$$
, $x_2 = i_q$,

$$A_{11} = \begin{bmatrix} 0 & 1 \\ 0 & \frac{-B_m - usin(n\phi_1)\dot{\phi}_1}{J_0 - J_1 cos(n\phi_1) + J_m} \end{bmatrix}, A_{12} = \begin{bmatrix} 0 \\ \frac{k_t}{J_0 - J_1 cos(n\phi_1) + J_m} \end{bmatrix},$$

$$A_{21} = \begin{bmatrix} 0 & -\frac{p\lambda_d}{L_q} \end{bmatrix}, A_{22} = -\frac{R_s}{L_q}, B_1 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, B_2 = \frac{1}{L_q}.$$

Under the VSC, the system can track the trajectory and approach to the sliding surface, so the sliding surface has to include track object. The sliding function can be expressed as

$$s = ce = c(x - x^*) = \begin{bmatrix} c_1 & c_2 \end{bmatrix} \begin{pmatrix} x_1 - x^*_1 \\ x_2 - x_2^* \end{bmatrix}$$
 (33)

where
$$c_1 \in R^{1 \times 2}, c_2 \in R; x^*_1 = \begin{bmatrix} \phi_1^* \\ \dot{\phi}_1^* \end{bmatrix}, x^*_2 = i^*_q$$

If the system model is on an ideal sliding surface s = ce = 0, we can determine the equivalent system and an associated linear control input.

In order to consider the linear system, we set $\phi_1 = 0$ in A_{11} and A_{12} of Equation (32). On the sliding surface s = 0, we can obtain

$$x_2 = x_2^* - \frac{c_1}{c_2} \left[x_1 - x_1^* \right]$$
 (34)

Substituting (34) into (32), we have

$$x_1^* = A_{11}x_1 + A_{12}(x_2^* - c_2^{-1}c_1 [x_1 - x_1^*])$$
 (35)

Defining $k = c_2^{-1}c_1$ we rewrite (35) as

$$x^*_1 = (A_{11} - A_{12}k)x_1 + A_{12}(x_2^* + kx_1^*)$$
 (36)

It is seen that the locations of poles of the resulting system (36) can be obtained by selecting k and c_2 . Because (A_{11}, A_{12}) is controllable, we use a pole placement method to select the gain k in (36).

5.3 Design the VSC law

The variable structure control [14] can be separated into the equivalent and nonlinear control components as $u = u_{eq} = u_n$. The equivalent control can be selected by the following equations

$$\dot{x} = Ax + Bu \tag{37}$$

$$\dot{s} = c\dot{e} = 0 \tag{38}$$

From (37) and (38), we obtain the equivalent control input as

$$u_{eq} = -(cB)^{-1}(cAx - c\dot{x}^*)$$
 (39)

The approaching condition [14] $s\dot{s} < 0$ can bring the system model to the sliding surface s = 0 Selecting the nonlinear control as

$$u_{eq} = -(cB)^{-1}(\alpha_1 sgn(s) + \alpha_2 s)$$
 (40)

where α_1 and α_2 are positive constants. Finally, it follows that

$$s\dot{s} = sc(Ax + Bu + D - \dot{x}^*)$$

$$= scAx - scB((cB)^{-1}(cAx - c\dot{x}^*)$$

$$+ scD + (cB)^{-1}(\alpha_1 sgn(s) + \alpha_2 s)) - sc\dot{x}^*$$

$$= -\alpha_2 s^2 - \alpha_1 |s| + cDs < 0, if \alpha_1 > |cD|$$
(41)

From (41), the approaching condition is negative except on the sliding surface s = 0.

For suppressing the switch function to create chatter free, we use the saturation function to replace the sign function and write the controller as

$$u = -(cB)^{-1}(cAx - cx^*) - (cB)^{-1}(\alpha_1 sat(s) + \alpha_2 s)$$
(42)

where
$$sat(s, \varepsilon) = \begin{cases} 1 & s > \varepsilon \\ s/\varepsilon & |s| \le \varepsilon \\ -1 & s < -\varepsilon \end{cases} = \begin{cases} sign(s) & |s| > \varepsilon \\ s/\varepsilon & |s| \le \varepsilon \end{cases}$$

6. Numerical simulation

6.1 Trajectory planning

In numerical simulations, the fourth-order Runge-Kutta method is employed to calculate the absolute input energy via a windows supported MATLAB software with a sampling time sec and the total time sec. The identified parameters of the LCD glass-handling robot driven by a PMSM are given in Table 1 [8].

The coefficients of the polynomials with 8-12 degrees are obtained by the RGA and are shown in Table 2. In this Table, the feasible domains are taken -1 to 1 for all the coefficients.

Figure 3 (a-h) show the comparisons among 8-12 degrees of polynomials. Figure 3 (a-d) show the designed trajectories of the angular displacements, speeds, accelerations and jerks of Arm 1, respectively. Taking time derivatives of the designed trajectories, we obtain its time derivatives, substitute into (2), and obtain the input current as shown in Figure 3 (e). By using (21), we can obtain input voltage of the PMSM as shown in Figure 3 (f). Figure 3(g, h) show the input energies and the fitness values, respectively. From Figure 3(h), it is seen that the fitness value increases with increasing the degrees' number of polynomials, and all the genes of the chromosome almost converge before the first 10 generations.

From Table2, it can be found that the minimum absolute input energy occurs for the twelve-degree polynomial. This value is lower -2.574% than that of eight-degree polynomial.

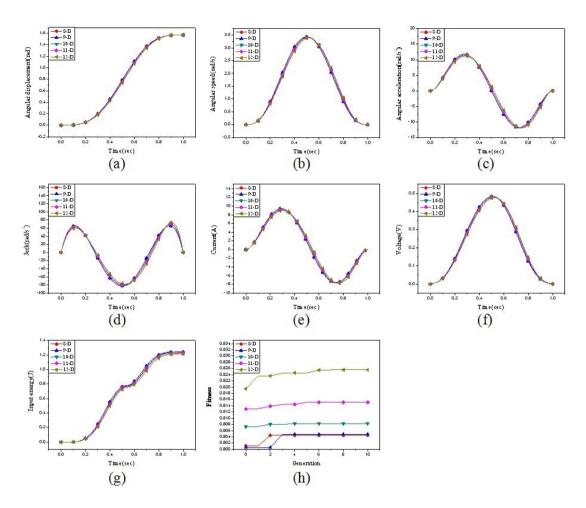


Figure 3 Comparisons among eight- to twelve-degree polynomials in numerical simulations. (a) The angular displacement. (b) The angular speed. (c) The angular acceleration. (d) The angular jerk. (e) The current. (f) The voltage. (g) The input energy. (h) The fitness functions of evolution history.

Table 2 Coefficients of polynomials and total input energy

Tuest 2 coefficients of polynomials and total input energy							
Polynomial	Coefficient	Coefficient	Coefficient	Coefficient	Coefficient	Total input	$\frac{(P_{m+1}-P_8)}{P_8}\%$
function							
of degree	a_8	a_9	a_{10}	a_{11}	a_{12}	energy(J)	
8	0.998	_	_	_	_	1.243	_
9	0.988	-0.740	-	-	-	1.242	-0.080
10	0.836	0.941	-0.926	-	-	1.238	-0.402
11	0.947	0.947	0.793	-0.894	-	1.228	-1.207
12	0.763	0.836	0.810	0.978	-0.977	1.211	-2.574

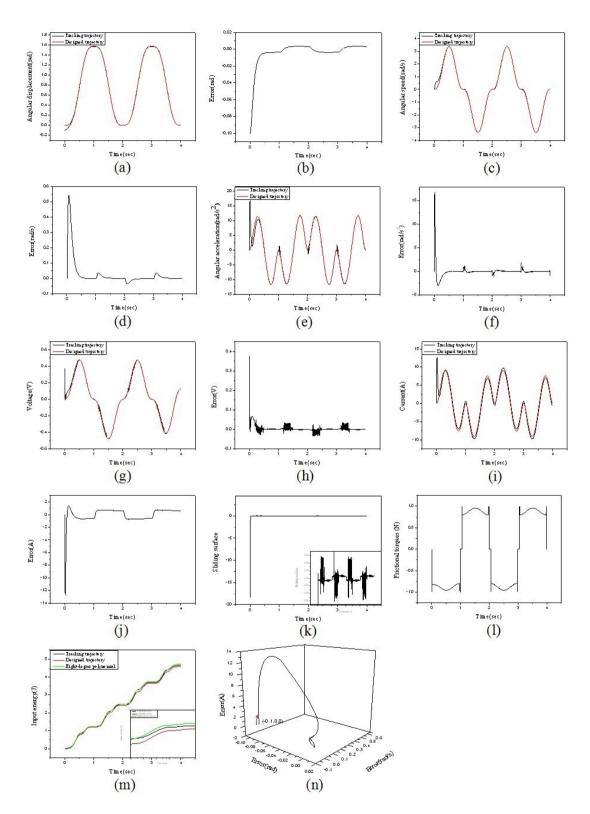


Figure 4 Comparison among the designed and tracking trajectories. (a) The angular displacement. (b) The errors of angular displacements. (c) The angular speed. (d) The errors of angular speeds. (e) The angular acceleration. (f) The errors of angular accelerations. (g) The voltages. (h) The errors of voltages. (i) The currents. (j) The errors of currents. (k) The sliding surface. (l) The frictional torques of LuGre model. (m) The input energies. (n) The phase trajectory in the state space.

Table 1 The identified parameters of the robot system

Parameter	Values
$J_0(kg-m^2)$	0.3585
$J_1(kg-m^2)$	0.0006708
$u(kg-m^2)$	0.0006708
$L_q(H)$	0.00153
λ_d	0.03572
$R_s(\Omega)$	0.00103
n	1.9135
$K_t(Nm/A)$	0.55624
$J_m(kg-m^2)$	0.0679
$B_m(Nms/rad)$	0.2858

6.2 Tracking control

In numerical simulations of the VSC, the fourth-order Runge-Kutta method is also employed with MAT-LAB software. The LuGre model parameters and control gains are shown in Table 3. In order to make system not on the sliding surface at t=0, we set $\phi_1=-0.1$ at the start time. To compare the control performance, we show two work cycles with each cycle 2 second. Here, we just show the case with twelve-degree polynomial.

Figures 4 (a-n) compare the designed and tracking trajectories. Figures 4 (a, c, e) compare the angular displacements, speeds and accelerations, respectively, for the designed and tracking trajectories. Figures 4 (b, d, f) show the errors of the angular displacements, speeds, accelerations, and tracking errors, respectively, between the designed and tracking trajectories. Figures 4 (g, i) compare the input voltages and currents, respectively, for the designed and tracking trajectories. Figures 4 (h, j) show the errors of the input voltages and currents, respectively, between the designed and tracking trajectories. Figure 4 (k) shows the sliding surface. Figure 4 (1) shows the frictional torque of LuGre model. Figure 4 (m) compares the input energies among the designed, tracking and eight-degree polynomial trajectories. It is found that they are very close, and the VSC is proven to track the trajectories well with frictional torques. Figure 4 (n) shows phase trajectory in the state space $(\phi_1, \dot{\phi}_1, \dot{i}_a)$.

7. Conclusion

In this paper, the minimum-energy-based PTP trajectory planning method is performed for an LCD glasshandling robot by utilizing the absolute input energy as the fitness function to determine the polynomial coefficients by the RGA. From the comparisons in nu-

Table 3 The identified parameters of the robot system

c)
)

merical simulations, the minimum-energy input occurs when the highest-degree polynomial is chosen. The nonlinear controller VSC successfully tracks the trajectories and compensates the frictional torques well. Moreover, the proposed methodology in this paper can also be applied to any mechatronic system driven by a PMSM for designing the minimum-energy PTP trajectory using a high-degree polynomial function.

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