

Robust Position Anti-Interference Control for PMSM Servo System With Uncertain Disturbance

Longfei Li, Jie Xiao, Yun Zhao, Kan Liu, Senior Member, *IEEE*, Xiaoyan Peng, Haozhe Luan, and Kaiqing Li

¹Abstract—Aiming to suppress the influence of uncertain disturbances in the drive control of permanent magnet synchronous machines (PMSM), such as the parameter uncertainties and load disturbance, a robust anti-interference control for the angular position tracking control of a PMSM servo system has been proposed in this paper. During the position tracking, uncertain system disturbances being regarded as a lumped unknown term will be online observed by a nonlinear disturbance observer (NDOB), of which the influence will consequently be counteracted by a robust backstepping compensator (RBC). The asymptotical stability of proposed control scheme is analyzed and designed according to the Lyapunov stability criterion, and its convergence against the system uncertain disturbance is verified on a prototype PMSM servo platform and shows good performance in rotor angular position tracking and anti-interference.

Index Terms—Nonlinear disturbance observer (NDOB), permanent magnet synchronous machine (PMSM), position control, robust backstepping compensator (RBC), servo system.

I. INTRODUCTION

WITH the rapid development of power electronic devices, microcomputers and modern control theory, the permanent magnet synchronous machine (PMSM) have been widely used in the textile industry, industrial robots, medical equipment, household appliances, etc. [1], [2], thanks to its simple structure, light weight, high power density and strong driving capability. However, the performance of position and speed controls of PMSMs is usually sensitive to the machine nonlinearity, time-varying parameters, and external load disturbances, and a strong position locking ability is usually needed in applications such as robotic arms and marine steering gears[3]. In this case, the traditional linear control method is usually difficult to ensure the overall performance of servo system [4].

In order to overcome this issue, related nonlinear control methods such as the sliding mode variable structure control [5], [6], adaptive control [7], [8], model predictive control [9], [10],

active disturbance rejection control (ADRC) [11], feedback linearization control [12] and adaptive backstepping control [20], [21] are widely investigated for solving the above problems. For example, A. T. Nguyen et al. [13] has proposed a model reference adaptive control (MRAC) based scheme, including an adaptive compensator and a feedback controller to improve the speed response against unknown external interference. However, the adaptive control scheme is usually difficult to eliminate the influence of unmodeled uncertainty. In this case, once the influence of uncertainty is out of a certain range, the resulting system chaotic responses will cause a deterioration of the system stability [14]. Besides, it is reported in some articles that the sliding mode control (SMC) is of high robustness and simple structure, which can guarantee the tracking accuracy in the presence of parameter uncertainty and external interference, and also can increase the control gain of the sliding mode surface to further improve the system robustness. However, a too large control gain tends to increase the chattering and then result in a deterioration of the control performance [15]-[17]. Similarly, Hebertt Sira-Ramirez et al. [18] employs an active disturbance rejection control (ADRC) scheme to overcome the influence of uncertain disturbance, which has a simple structure and can be easily operated in discrete time. However, it is difficult to properly adjust the parameter setups of the ADRC controller. Recently, a novel adaptive backstepping control (ARBC) is proposed in [19], which has taken into account the estimation of load torque for improving the rotor position tracking. Similarly, the ARBC with extended state observer (ESO) is also used to deal with the parameter uncertainty and external interference for the speed regulation of a wind turbine differential mechanism, while the unmodeled uncertainty of the system without consider [20]. Besides, the ARBC is also used to cooperate with the estimated moment of inertia and load torque to achieve good control performance under ultra-low speed control [21].

In this paper, a RBC combined with a nonlinear disturbance observer (NDOB) is proposed for the anti-interference position control of PMSM servo system, which has accounted for the influences of uncertain modeling errors and external disturbances. The proposed scheme has considered the influence of mismatched parameters of PMSM servo system and the overall system disturbances in current, torque, speed etc. Those disturbances are modeled as a lumped term and properly online compensated by the proposed method. Besides, the asymptotical stability of the proposed control scheme is analyzed and designed according to the Lyapunov stability criterion, which guarantees the stability and convergence of whole system. The performance of proposed method is finally verified on a prototype PMSM servo system, and shows quite good robustness against unknown parameters and external load

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disturbances.

II. PMSM MODEL

The mathematical model of PMSM on the dq -axis can be expressed as

$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = \theta_1 x_3 - \theta_2 x_2 - \theta_3 \\ \dot{x}_3 = -g_1 x_3 - g_2 x_2 x_4 - g_3 x_4 + g_4 u_q + f_1 \\ \dot{x}_4 = -g_1 x_4 + g_2 x_2 x_3 + g_4 u_d + f_2 \end{cases} \quad (1)$$

where

$$\begin{aligned} x_1 = \theta_m, x_2 = \omega_m, x_3 = i_q, x_4 = i_d \\ g_1 = \frac{R_s}{L}, g_2 = P_n, g_3 = \frac{P_n \psi_f}{L}, g_4 = \frac{1}{L} \\ \theta_1 = \frac{3P_n \psi_f}{2J}, \theta_2 = \frac{B}{J}, \theta_3 = \frac{T_L}{J} \end{aligned} \quad (2)$$

where i_d, i_q represent the dq -axis currents; u_d, u_q represent the dq -axis voltages; R_s, L, ψ_f and P_n represent the stator resistance, dq -axis inductances, permanent magnet flux linkage and the number of pole pairs, respectively; J, T_L and B represent the moment of inertia, the load torque and the viscous friction coefficient, respectively; θ_m, ω_m represent mechanical position angle and the rotor angular speed; f_1 and f_2 represent other uncertainties, including modeling errors, external disturbances, etc.

Taking into account the parameter variations, (1) can be expressed as

$$\begin{aligned} \dot{x}_2 &= \theta_1 x_3 - \theta_2 x_2 - \theta_3 \\ &= (\theta_{1n} + \Delta\theta_1)x_3 - (\theta_{2n} + \Delta\theta_2)x_2 - (\theta_{3n} + \Delta\theta_3) \\ &= \theta_{1n} x_3 - \theta_{2n} x_2 + d \\ d &= \Delta\theta_1 x_3 - \Delta\theta_2 x_2 - \theta_3 \end{aligned} \quad (3)$$

where $\theta_{1n}, \theta_{2n}, \theta_{3n}$ are the nominal parameters; $\Delta\theta_1, \Delta\theta_2, \Delta\theta_3$ are the parameter variations, d represents the lumped disturbance, which includes internal parameter variations, external load disturbances, and other uncertainties.

To facilitate the controller design, the system needs to meet the following assumptions

Assumption 1: The lumped disturbance d is bounded and has a slow change with time

$$\begin{cases} |d| < \mu \\ \dot{d} \approx 0 \end{cases} \quad (5)$$

where μ represents the boundary value of d .

Assumption 2: Other uncertain terms f_1, f_2 satisfy the following conditions

$$\begin{cases} |f_1| \leq h_1 \\ |f_2| \leq h_2 \end{cases} \quad (6)$$

where h_1, h_2 represent the boundary values of f_1 and f_2 , respectively

III. DESIGN OF POSITION CONTROL STRATEGY FOR PMSM

A. Design of Nonlinear Disturbance Observer

To reduce the influence of internal parameter variations and

external load disturbances, a NDOB for compensation control is introduced.

To estimate the system disturbances, the NDOB [22] applied in this paper can be given as follows

$$\begin{cases} \dot{\hat{d}} = p + lx_2 \\ \dot{p} = -lp - l[lx_2 - \theta_{2n}x_2 + \theta_{1n}x_3] \end{cases} \quad (7)$$

where \hat{d} , p , and l are estimates of the lumped system disturbance, intermediate variables of NDOB and gains of NDOB, respectively.

Define the disturbance error as e_d which satisfies the following conditions

$$\begin{cases} e_d = \hat{d} - d \\ |e_d| \leq \xi \end{cases} \quad (8)$$

where ξ is the boundary value of the disturbance error e_d .

It can be proved that \hat{d} will asymptotically approach the actual value d by selecting the appropriate parameter l .

Proof: By combined of (3), (7) and *Assumption 1*, \dot{e}_d is got as follows

$$\begin{aligned} \dot{e}_d &= \dot{\hat{d}} - \dot{d} = \dot{p} + l\dot{x}_2 - \dot{d} \\ &= -lp - l[lx_2 - \theta_{2n}x_2 + \theta_{1n}x_3] + l[\theta_{1n}x_3 - \theta_{2n}x_2 + d] \\ &= -l(p + lx_2) + ld = -l(\hat{d} - d) = -le_d \end{aligned} \quad (9)$$

It is worth to note that $l > 0$ in (9) and magnitude of this parameter determines the speed of convergence, which means the convergence speed increases with l . The schematic diagram of NDOB is shown in Fig. 1. The advantage of the NDOB introduced in the paper is that its structure is much simpler and it can effectively estimate and compensate a wide range of uncertainty and interference without sacrificing the overall control performance.

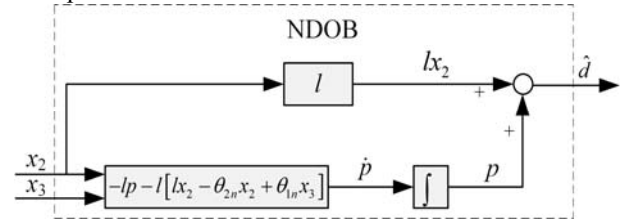


Fig. 1. Schematic diagram of NDOB

B. Design of Robust Backstepping Controller

The RBC can simplify the high-order system to several first-order subsystems by introducing virtual control variables [19], [20]. A RBC based on NDOB is proposed to control PMSM servo system considering parameter uncertainties and external load disturbances. The overall control block diagram of the RCB with NDOB is shown in Fig. 2.

According to the basic design principle of backstepping controller, the error variables z_1, z_2, z_3 and z_4 of the system can be defined as

$$\begin{cases} z_1 = x_1 - x_r \\ z_2 = x_2 - \alpha_1 \\ z_3 = x_3 - \alpha_2 \\ z_4 = x_4 \end{cases} \quad (10)$$

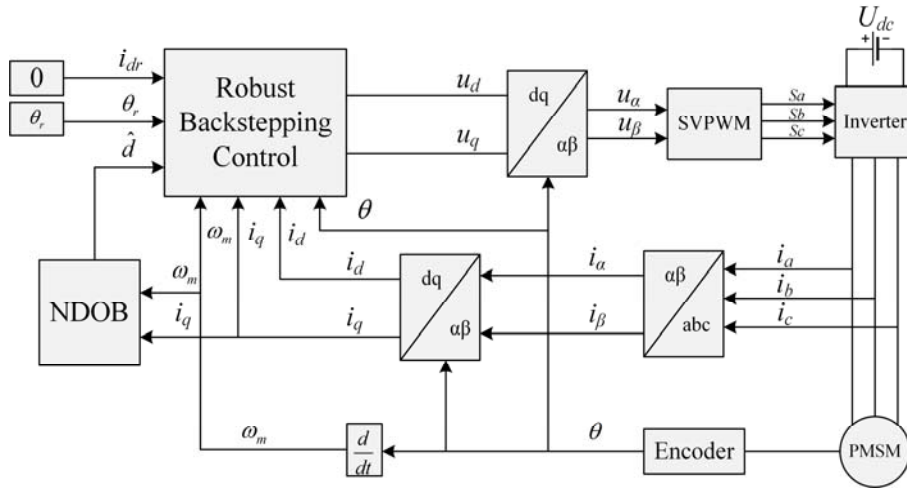


Fig. 2. Overall control block diagram of RBC with NDOB

where x_r is the desired signal; α_1 , α_2 are the desired virtual control signals.

Step 1: Select the first Lyapunov function V_1 as follows

$$V_1 = \frac{1}{2} z_1^2 \quad (11)$$

By differentiating (11), \dot{V}_1 can be obtain from (12)

$$\dot{V}_1 = z_1 \dot{z}_1 = z_1 (z_2 + \alpha_1 - \dot{x}_r) \quad (12)$$

The virtual control variable α_1 is represented as follows

$$\alpha_1 = -k_1 z_1 + \dot{x}_r \quad (13)$$

where k_1 is an adjustable positive parameter, \dot{V}_1 can be written as follows

$$\dot{V}_1 = -k_1 z_1^2 + z_1 z_2 \quad (14)$$

Step 2: Select the second Lyapunov function V_2 as follows

$$V_2 = V_1 + \frac{1}{2} z_2^2 \quad (15)$$

By differentiating (15), \dot{V}_2 can be obtained below

$$\begin{aligned} \dot{V}_2 = \dot{V}_1 + z_2 \dot{z}_2 = & -k_1 z_1^2 + \theta_{1n} z_2 z_3 + z_2 (z_1 + \\ & \theta_{1n} \alpha_2 - \theta_{2n} x_2 + d + k_1 x_2 - k_1 \dot{x}_r - \ddot{x}_r) \end{aligned} \quad (16)$$

The virtual control variable α_2 is represented as follows

$$\begin{cases} \alpha_2 = \alpha_{2a} + \alpha_{2r} \\ \alpha_{2a} = -\frac{1}{\theta_{1n}} (z_1 - \theta_{2n} x_2 + \hat{d} + k_1 x_2 - k_1 \dot{x}_r - \ddot{x}_r + k_2 z_2) \end{cases} \quad (17)$$

where k_2 is an adjustable positive parameter, α_{2a} is the model compensation control to achieve tracking control and α_{2r} is the robust control rate used to reduce the effects of system uncertainties.

Combining (16) and (17) yields (18):

$$\begin{cases} \dot{V}_2 = \dot{V}_2|_{equal} + \theta_{1n} z_2 z_3 \\ \dot{V}_2|_{equal} = -k_1 z_1^2 - k_2 z_2^2 + z_2 (\theta_{1n} \alpha_{2r} - e_d) \end{cases} \quad (18)$$

According to (18), the selection of α_2 should meet the following conditions to ensure the system stability:

$$\begin{cases} z_2 \theta_{1n} \alpha_{2r} \leq 0 \\ z_2 (\theta_{1n} \alpha_{2r} - e_d) \leq \varepsilon_1 \end{cases} \quad (19)$$

where ε_1 is an arbitrarily small positive parameter which represents the attenuation level of system uncertainties.

Therefore, the robust control rate α_{2r} satisfies (19) can be chosen as

$$\alpha_{2r} = -\frac{\xi^2}{4\theta_{1n}\varepsilon_1} z_2 \quad (20)$$

Proof: Combining (8), (19) and (20) yields (21):

$$\begin{aligned} z_2 (\theta_{1n} \alpha_{2r} - e_d) & \leq -\frac{\xi^2}{4\varepsilon_1} z_2^2 + \xi |z_2| \\ & = -\left(\frac{\xi |z_2|}{2\sqrt{\varepsilon_1}} - \sqrt{\varepsilon_1} \right)^2 + \varepsilon_1 \leq \varepsilon_1 \end{aligned} \quad (21)$$

Step 3: Select the third Lyapunov function V_3 as follows

$$V_3 = V_2 + \frac{1}{2} z_3^2 \quad (22)$$

By differentiating z_3 , \dot{z}_3 can be obtained below

$$\begin{aligned} \dot{z}_3 = \dot{x}_3 - \dot{\alpha}_2 = & \dot{x}_3 - \varphi_1 + \varphi_2 \dot{x}_2 \\ = & -g_1 x_3 - g_2 x_2 x_4 - g_3 x_4 + g_4 u_q + f - \varphi_1 + \varphi_2 \dot{x}_2 \end{aligned} \quad (23)$$

where

$$\begin{cases} \varphi_1 = \frac{1}{\theta_{1n}} [(k_1 + k_2) \ddot{x}_r + \ddot{x}_r - \hat{d} - (1 + k_1 k_2)(x_2 - \dot{x}_r) + \frac{\xi^2}{4\varepsilon_1} z_2] \\ \varphi_2 = \frac{1}{\theta_{1n}} \left(\frac{\xi^2}{4\varepsilon_1} + k_1 + k_2 - \theta_{2n} \right) \end{cases} \quad (24)$$

By combining (22) and (23), \dot{V}_3 is derived as follows

$$\begin{aligned} \dot{V}_3 = \dot{V}_2|_{equal} - g_2 x_2 z_3 z_4 + z_3 (\theta_{1n} z_2 - g_1 x_3 - g_3 x_2 \\ + g_4 u_q + f - \varphi_1 + \varphi_2 \theta_{1n} x_3 - \varphi_2 \theta_{2n} x_2 + \varphi_2 \hat{d} - \varphi_2 e_d) \end{aligned} \quad (25)$$

According to (25), the expression of control output u_q is expressed as follows

$$\begin{cases} u_q = u_{qa} + u_{qr} \\ u_{qa} = -\frac{1}{g_4} \left(\theta_{1n} z_2 - g_1 x_3 - g_3 x_2 - \varphi_1 + \varphi_2 \theta_{1n} x_3 - \varphi_2 \theta_{2n} x_2 + \varphi_2 \hat{d} + k_3 z_3 \right) \end{cases} \quad (26)$$

where k_3 is an adjustable positive parameter, u_{qa} is the model

compensation control, u_{qr} is the robust control rate.

By combining (25) and (26), (27) can be obtained as follows

$$\begin{cases} \dot{V}_3 = \dot{V}_3|_{equal} - g_2 x_2 z_3 z_4 \\ \dot{V}_3|_{equal} = \dot{V}_2|_{equal} - k_3 z_3^2 + z_3 (g_4 u_{qr} + f_1 - \varphi_2 e_d) \end{cases} \quad (27)$$

Similarly, the selection of u_{qr} should meet the following conditions

$$\begin{cases} z_3 g_4 u_{qr} \leq 0 \\ z_3 (g_4 u_{qr} + f_1 - \varphi_2 e_d) \leq \varepsilon_2 + \varepsilon_{2r} \end{cases} \quad (28)$$

where $\varepsilon_2, \varepsilon_{2r}$ are arbitrarily small positive parameters.

Therefore, the robust control rate u_{qr} can be chosen as follows

$$u_{qr} = -\frac{z_3}{g_4} \left(\frac{h_1^2}{4\varepsilon_2} + \frac{(\varphi_2 \xi)^2}{4\varepsilon_{2r}} \right) \quad (29)$$

Proof: By combining (6), (8), (28) and (29), (30) can be derived as follows

$$\begin{aligned} & z_3 (g_4 u_{qr} + f_1 - \varphi_2 e_d) \\ & \leq -\left(\frac{h_1^2}{4\varepsilon_2} + \frac{(\varphi_2 \xi)^2}{4\varepsilon_{2r}} \right) z_3^2 + h_1 |z_3| + \xi |\varphi_2| |z_3| \\ & = -\left(\frac{h_1 |z_3|}{2\sqrt{\varepsilon_2}} - \sqrt{\varepsilon_2} \right)^2 - \left(\frac{|\varphi_2| \xi |z_3|}{2\sqrt{\varepsilon_{2r}}} - \sqrt{\varepsilon_{2r}} \right)^2 + \varepsilon_2 + \varepsilon_{2r} \\ & \leq \varepsilon_2 + \varepsilon_{2r} \end{aligned} \quad (30)$$

Step 4: Select the last Lyapunov function V_4 as follows

$$V_4 = V_3 + \frac{1}{2} z_4^2 \quad (31)$$

By differentiating (30), \dot{V}_4 can be obtained as follows

$$\dot{V}_4 = \dot{V}_3|_{equal} + z_4 (-g_2 x_2 z_3 - g_1 x_4 + g_2 x_2 x_3 + g_4 u_d + f_2) \quad (32)$$

Similarly, according to (31), the control output u_d is expressed as follows

$$\begin{cases} u_d = u_{da} + u_{dr} \\ u_{da} = -\frac{1}{g_4} (-g_2 x_2 z_3 - g_1 x_4 + g_2 x_2 x_3 + k_4 z_4) \end{cases} \quad (33)$$

where k_4 is an adjustable positive parameter, u_{da} is the model compensation control to achieve tracking control, u_{dr} is the robust control rate.

By combining (32) and (33), (34) is got as follows

$$\dot{V}_4 = \dot{V}_3|_{equal} - k_4 z_4^2 + z_4 (g_4 u_{dr} + f_2) \quad (34)$$

Similarly, u_{dr} should meet the following conditions:

$$\begin{cases} z_4 g_4 u_{dr} \leq 0 \\ z_4 (g_4 u_{dr} + f_2) \leq \varepsilon_3 \end{cases} \quad (35)$$

where ε_3 is an arbitrarily small positive parameter.

Therefore, the robust control rate u_{dr} can be chosen as follows

$$u_{dr} = -\frac{h_2^2}{4g_4\varepsilon_3} z_4 \quad (36)$$

Proof: Combined (6), (35) and (36), (37) can be derived as follows

$$\begin{aligned} z_4 (g_4 u_{dr} + f_2) & \leq -\frac{h_2^2}{4\varepsilon_3} z_4^2 + h_2 |z_4| \\ & = -\left(\frac{h_2 |z_4|}{2\sqrt{\varepsilon_3}} - \sqrt{\varepsilon_3} \right)^2 + \varepsilon_3 \leq \varepsilon_3 \end{aligned} \quad (37)$$

C. Stability Analysis

Construct a Lyapunov function V as follows

$$V(t) = V_4 = \frac{1}{2} z_1^2 + \frac{1}{2} z_2^2 + \frac{1}{2} z_3^2 + \frac{1}{2} z_4^2 \quad (38)$$

By combining (19), (28), (35) and (34), \dot{V} can be obtained below

$$\begin{aligned} \dot{V}(t) & = -k_1 z_1^2 - k_2 z_2^2 - k_3 z_3^2 - k_4 z_4^2 + z_2 (\theta_{1n} \alpha_{2r} - e_d) \\ & \quad + z_3 (g_4 u_{qr} + f_1 - \varphi_2 e_d) + z_4 (g_4 u_{dr} + f_2) \\ & \leq -k_1 z_1^2 - k_2 z_2^2 - k_3 z_3^2 - k_4 z_4^2 + \varepsilon_1 + \varepsilon_2 + \varepsilon_{2r} + \varepsilon_3 \\ & \leq -2kV(t) + \varepsilon \end{aligned} \quad (39)$$

where

$$\begin{cases} k = \min(k_1, k_2, k_3, k_4) \\ \varepsilon = \varepsilon_1 + \varepsilon_2 + \varepsilon_{2r} + \varepsilon_3 \end{cases} \quad (40)$$

The solution of differential equation of (39) can be expressed as follows

$$V(t) \leq e^{-2kt} V(0) + \frac{\varepsilon}{2k} (1 - e^{-2kt}) \quad (41)$$

According to (41), it can be determined that the closed-loop control system is asymptotically stable, and its exponential convergence rate is k , and the final convergence error e satisfies following formula

$$|e| \leq \lim_{t \rightarrow \infty} V(t) = \frac{\varepsilon}{2k} \quad (42)$$

Therefore, the convergence speed and error of the control system can be changed by adjusting the parameters ε and k .

IV. EXPERIMENTAL VERIFICATION

In this section, a Speed_goat control platform is adopted to drive a 750W surface mounted PMSM to verify the effectiveness of the proposed control strategy. In order to test the performance under different conditions, the external load will be given by the magnetic powder brake, heavy hammer and DC load motor, respectively. The experimental platform and parameters of PMSM are shown in Fig. 4 and TABLE I, respectively. The DC bus voltage is 60V and the IGBT switching frequency is 10kHz. The speed measurement uses the M/T hybrid measurement method integrated in Speed_goat control platform. In addition, a 2500-lines incremental encoder has been applied to achieve an accurate measurement of speed, which cooperates with a low-pass filter for the sake of suppressing high frequency noise. The performance of the proposed control in this paper is compared with the optimal PI control with and without NDOB. The latter two control methods are briefly introduced in the Appendix. The related parameters of the three control methods are given below

(1) *RBC with NDOB:* the control gain of the NDOB is selected as $l = 200$; the boundary value h_1, h_2 of the uncertain

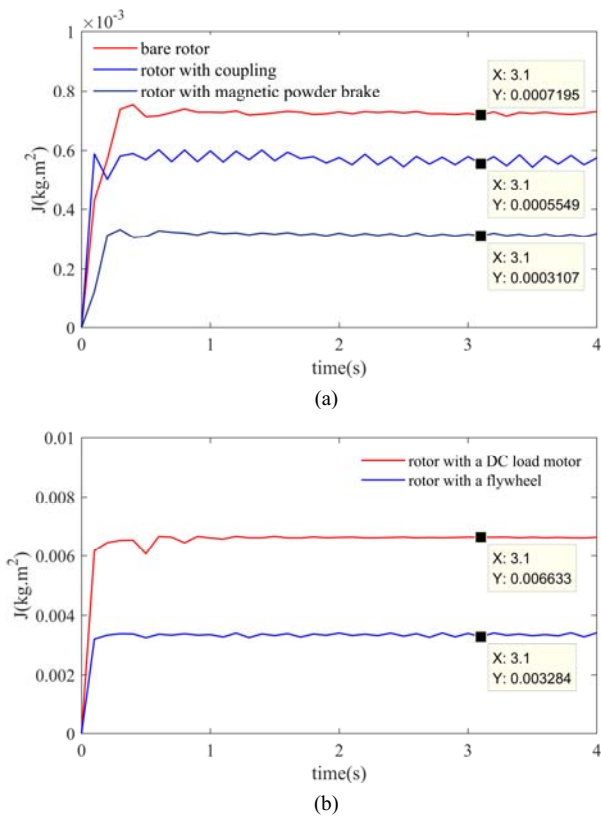


Fig. 3. Estimated combined moment of inertia of PMSM under different conditions. (a) bare rotor, rotor with coupling and with magnetic powder brake, respectively. (b) rotor with a flywheel and with a DC load motor, respectively

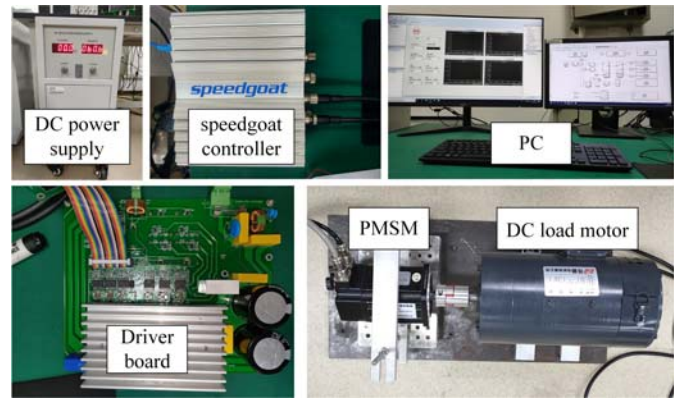


Fig. 4. Speedgoat controller based PMSM test rig.

TABLE I
PARAMETERS OF PMSM

Symbol	PMSM parameters	Value
U_N	rated voltage	220V
I_N	rated current	3A
P_N	rated power	750W
T_N	rated torque	2.4N·m
n_N	rated speed	3000r/min
R_s	stator resistance	1.86Ω
L	dq -axis inductances	2.8mH
ψ_f	permanent magnet flux linkage	0.109Wb
P_n	number of pole pairs	4
J	moment of inertia of bare rotor	$2.95 \times 10^{-4} \text{kg}\cdot\text{m}^2$
B	viscous friction coefficient	0.001

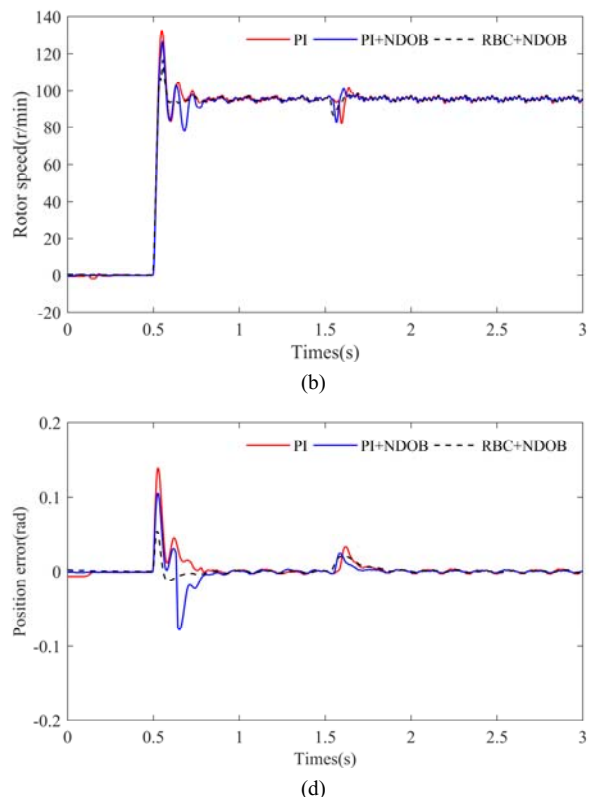
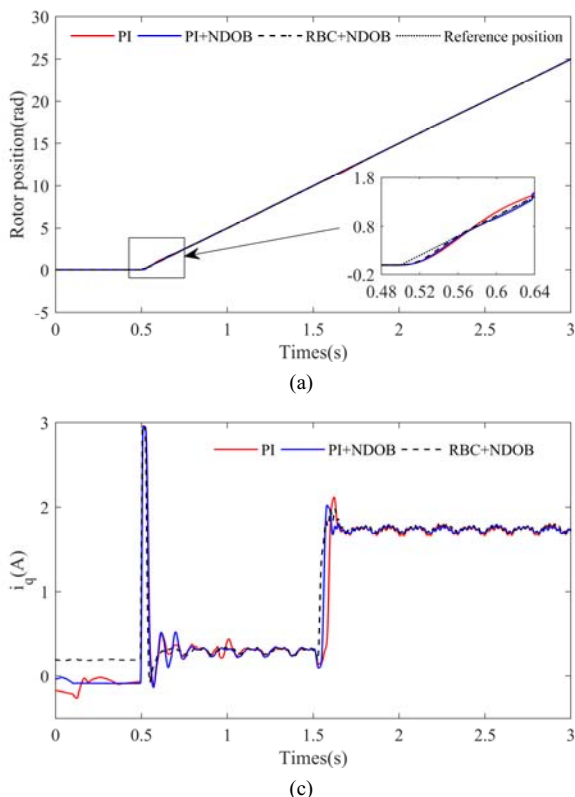


Fig. 5. Comparative experimental tests with an addition of external load $T_L=1\text{N}\cdot\text{m}$ at $t=1.5\text{s}$. (a) rotor position. (b) rotor speed. (c) q -axis current. (d) rotor position error

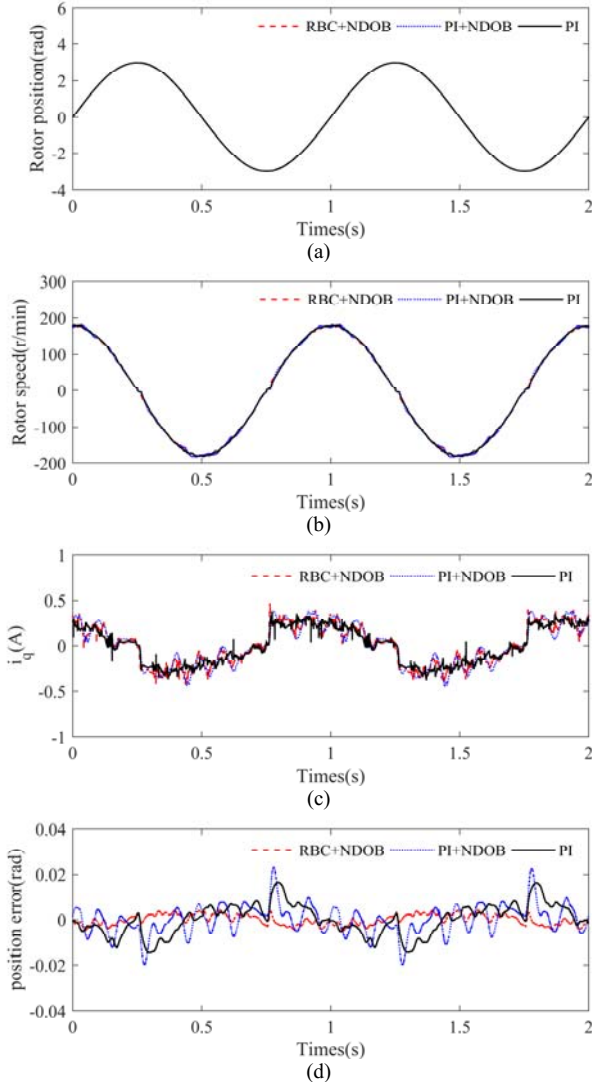


Fig. 6. Sinusoidal position tracking test at no load condition. (a) rotor position. (b) rotor speed. (c) q -axis current. (d) rotor position error.

variable f_1, f_2 is set as $h_1 = h_2 = 20$; the boundary value ζ of the disturbance error is set to $\zeta = 10$; RBC control parameters are set as $k_1 = 50, k_2 = 100, k_3 = 500, k_4 = 200, \varepsilon_1 = 100, \varepsilon_2 = 200, \varepsilon_2 = 0.01, \varepsilon_3 = 0.01$.

(2) *Optimal PI*: according to the optimal PI control method introduced in the Appendix, the control parameters are set to $\omega_i = 2\pi \times 600 \text{ rad/s}, \omega_s = 2\pi \times 30 \text{ rad/s}, \omega_p = 2\pi \times 6 \text{ rad/s}$.

(3) *Optimal PI with NDOB*: this control method is introduced in the Appendix, and its PI control constants and NDOB gain are the same as hereinbefore.

To optimize the performance of PI control, a moment of inertia identification experiment of PMSM is carried out. The identification method is also introduced in the Appendix, and the experimental results are shown in Fig. 3. The moment of inertia of bare rotor, rotor with coupling, rotor with magnetic powder brake, rotor with flywheel and rotor with DC load motor are $3.11 \times 10^{-4} \text{ kg}\cdot\text{m}^2, 5.55 \times 10^{-4} \text{ kg}\cdot\text{m}^2, 7.2 \times 10^{-4} \text{ kg}\cdot\text{m}^2, 3.28 \times 10^{-3} \text{ kg}\cdot\text{m}^2$ and $6.63 \times 10^{-3} \text{ kg}\cdot\text{m}^2$, respectively.

(1) *Tracking the slope change of rotor position*: In the comparative experiment, the PMSM is loaded by a DC load

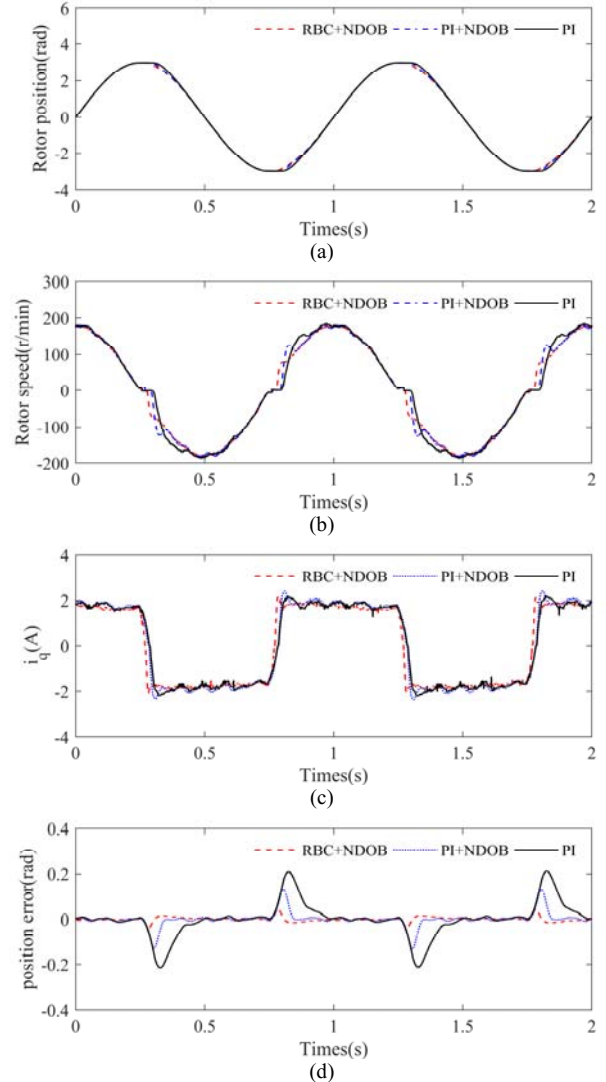


Fig. 7. Sinusoidal position tracking test in loaded condition. (a) rotor position. (b) rotor speed. (c) q -axis current. (d) rotor position error.

motor. As shown in Fig. 5, the reference rotor position of PMSM is $\theta_r = 10 \cdot (t - 0.5) \text{ rad}$ at $t = 0.5 \text{ s}$ and the motor is loaded with $1 \text{ N}\cdot\text{m}$ at $t = 1.5 \text{ s}$. As can be seen from Fig. 5(a) that the proposed method has less fluctuation during the position tracking. It can be seen that the proposed method has high tracking accuracy and can quickly converge to a given position even if there is an abrupt change in the external load as shown in Fig. 5(d). Fig. 8 (a) is the observed torque of NDOB, which settles down to $0 \text{ N}\cdot\text{m}$ between 0 and 0.5 s . Further, it fluctuates around $0.05 \text{ N}\cdot\text{m}$ from 0.5 s to 1.5 s , then after 1.5 s , it fluctuates around $1.05 \text{ N}\cdot\text{m}$. It is obvious that the observed result of the NDOB algorithm can quickly converge to the true value within 0.06 s even there is a load change.

(2) *Tracking the sinusoidal change of rotor position*: As shown in Figs. 6 and 7, the comparative experiments with and without external load torque ($1 \text{ N}\cdot\text{m}$), have the same sinusoidal input $\theta_r = 3 \cdot \sin(2\pi t) \text{ rad}$. It can be seen that the proposed method has smoother speed and current responses compared to the other two control schemes. Fig. 6(d) and Fig. 7(d) show that at no load condition, the dynamic position errors $|e_\theta|$ ($e_\theta = \theta_r - \theta_m$)

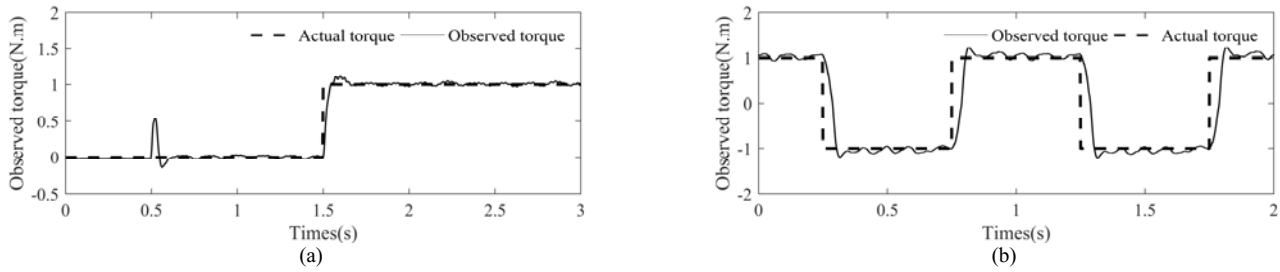


Fig. 8. Observed torque of NDOB. (a) suddenly add a load of $1\text{N}\cdot\text{m}$ at $t = 1.5\text{s}$. (b) square wave load with $1\text{N}\cdot\text{m}$

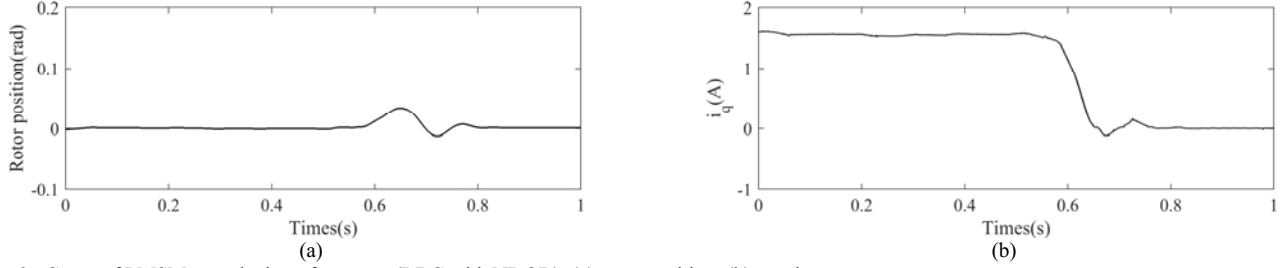


Fig. 9. Curve of PMSM rotor lock performance (RBC with NDOB). (a) rotor position. (b) q -axis current

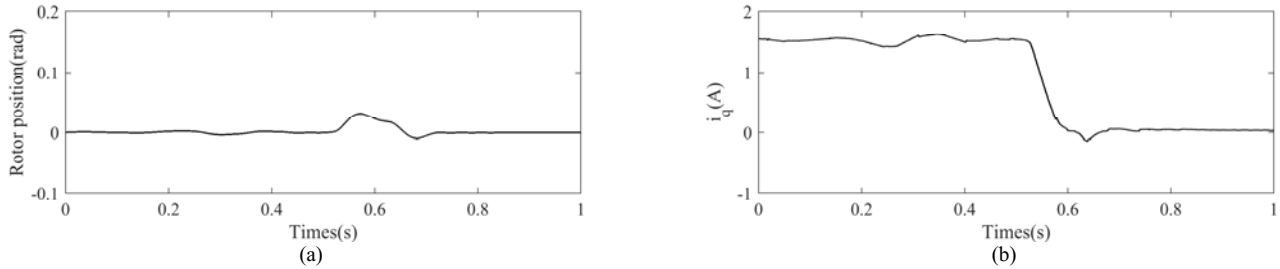


Fig. 10. Curve of PMSM rotor lock performance (PI control with NDOB). (a) rotor position. (b) q -axis current

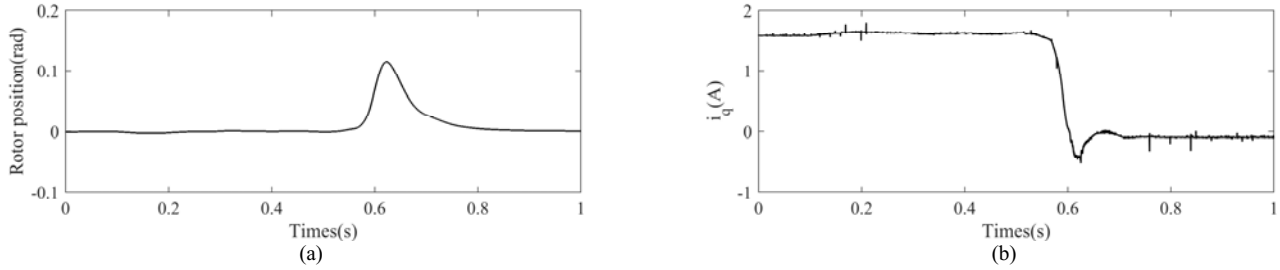


Fig. 11. Curve of PMSM rotor lock performance (PI control). (a) rotor position. (b) q -axis current

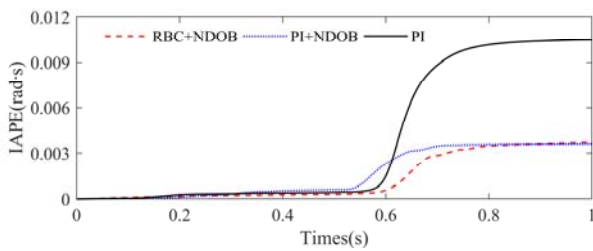


Fig. 12. Curve of integral absolute position error ($\text{IAPE} = \int_0^t |e_\theta(\tau)| d\tau$)

of three control methods (PI, PI+NDOB and RBC+NDOB) are less than 0.02rad , 0.025rad and 0.01rad , respectively, while in loaded condition $T_L = 1\text{N}\cdot\text{m}$, $|e_\theta|$ is less than 0.2rad , 0.15rad and 0.05rad , respectively. Therefore, the proposed method has less position tracking errors at both no load and loaded conditions. Fig. 8(b) is the corresponding observed torque result of NDOB under load ($1 \text{ N}\cdot\text{m}$). It can be seen from the results of Fig. 8 that the proposed NDOB shows faster convergence speed and accuracy. According to the comparison tests shown in Figs. 5-7,

it can be seen that the proposed method is robust to the variation of the load's moment of inertia.

(3) *Holding the rotor position*: In this experiment, the PMSM rotor will be held to the position of zero rad and connected to a flywheel with a heavy hammer and the weight will be removed at $t = 0.5\text{s}$. The equivalent load of the PMSM after the addition of a heavy hammer is about $T_L = 0.88\text{N}\cdot\text{m}$. From the results shown in Fig. 9-11, it can be seen that the RBC+NDOB and PI+NDOB control have a small position change once the heavy hammer is removed. Furthermore, as can be seen in Fig. 12, which is the curve of integrated absolute position error (IAPE), it shows that the RBC+NDOB and PI+NDOB control have better anti-interference (when $t=1\text{s}$, IAPE of RBC+NDOB is $3.56 \times 10^{-4}\text{rad}\cdot\text{s}$, IAPE of PI+NDOB is $3.66 \times 10^{-4}\text{rad}\cdot\text{s}$ and IAPE of PI is $10.46 \times 10^{-4}\text{rad}\cdot\text{s}$). Besides, since the estimation error of the NDOB may cause an over-compensation or under-compensation in q -axis current, the employed RBC will eliminate this kind of influence to achieve a better control performance.

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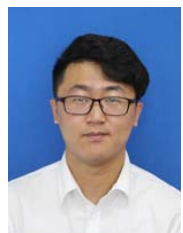
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