Inertia identification based on adaptive interconnected Observer of Permanent Magnet Synchronous Motor

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Abstract: In order to achieve the inertia identification of permanent magnet synchronous motor, this paper presents an adaptive interconnected observer based on model reference adaptive and Rhomberg observer to observe the inertia of permanent magnet synchronous motor and build the simulation model in MATLAB. By the simulation analysis, the observer can identify the moment of inertia of permanent magnet synchronous motor accurately, at the same time has good robustness.

Key words: Inertia, MRAS, Rhomberg observer, PMSM, MATLAB

I. INTRODUCTION

With the development of modern industry, the AC servo system is more and more widely used in industrial production. Permanent magnet synchronous motor has good performance and is widely used in AC servo system. High performance AC servo system not only requires a quick response to the instruction, but also can have good stability when the servo system has a load. In the actual drive system, rotor inertia is an important mechanical parameters, when the load changes, inertia will be changed, which will bring greater influence to drive system’s operating parameters. In order to make the drive system have good robustness and guarantee the high precision control, it is necessary to adjust the parameters of the controller of the drive system, which must be carried out to identify the inertia of the motor. In this paper, Based on model reference adaptive method and Rhomberg observer, an adaptive interconnected observer is established to identify the rotation inertia.

II. PERMANENTMAGNETSYNCHRONOUSMOTORMATHEMATICALMODEL

In this paper, the permanent magnet synchronous motor is SPMSM, that is: $L_d = L_q$, and use with maximum torque current ratio control strategy, that is $i_d = 0$, so the mathematical model of permanent magnet synchronous motor can be turned into:

$$
\begin{align*}
    u_d &= -\omega_r L_q i_q \\
    u_q &= L_q \frac{di_q}{dt} + R_s i_q + e \\
    T_e &= \frac{3}{2} p \varphi_f i_q \\
    T_e &= \int \frac{d\omega_r}{dt} + B \omega_r + T_L 
\end{align*}
$$

(1)

Where $u_d$ and $u_q$ are the d- and q-axis voltage, respectively; $i_d$ and $i_q$ are the d- and q-axis stator current, respectively; $\omega_r$ is the rotor mechanical speed; $L_d$ and $L_q$ are the d- and q-axis inductance; $R_s$ is the winding stator resistance; $\varphi_f$ is the permanent-magnet flux linkage; $e$ is the counter electromotive force; $T_e$ is the electromagnetic torque; $B$ is the viscous friction coefficient; $T_L$ is the load torque; and $p$ is the number of pole.
III. ADAPTIVE INTERCONNECTED OBSERVER

The establishment of the adaptive interconnected observer, which is based on model reference adaptive and Rhomberg observer and combination of the two together. In this paper, through the method of model reference adaptive to estimate the moment of inertia, and using Rhomberg observer for rotor speed estimation, then both estimated values are substituted into the two models, through the iterative process in order to get the precise value.

3.1 Model reference adaptive

The model reference adaptive method is a mature method in the field of parameter identification. It is based on the stability theory of Lyapunov and the ultra-stability theory of Popov, thus, it can guarantee the convergence of parameter identification. The main idea of the model reference adaptive is using the equation which contain the unknown parameter as an adjustable model and without unknown parameters as the reference model, and the two models’ output have the same physical significance. The two models work simultaneously, and adjust the parameters of the adjustable model according to the output error and appropriate adaptive rate, so as to achieve the target's output tracking reference model, and to achieve the purpose of that the output of control object can track the reference model. And the functional block figure is as follows fig.1.

![Fig.1 Model reference adaptive identification principle](image)

In the AC servo system, the system dynamic equation as for:

$$T_e = T_L + B \frac{d\omega_r}{dt} + J \frac{d^2 \theta_r}{dt^2}$$

(2)

The discretization of equation (2) as follows:

$$\omega_r(k) = \omega_r(k - 1) + \frac{T}{J} [T_e(k - 1) - T_L(k - 1) - B\omega_r(k - 1)]$$

(3)

Where $T$ is the system sampling period.

In quick response AC servo system, the system sampling frequency is high, in a single cycle, the load torque can be considered as a constant. Meanwhile, AC servo system viscous friction coefficient is small, so it can be ignored. After the above analysis, the equation (3) can be simplified as:

$$\omega_r(k) = 2\omega_r(k - 1) + b[T_e(k - 1) - T_e(k - 2)]$$

(4)
Were \( b \) is the identification parameter.

The equation (4) as the reference model, and then the adjustable model as follows:

\[
\dot{\omega}_r = 2 \omega_r(k-1) - \omega_r(k-2) + b(k-1) \left[ T_e(k-1) - T_e(k-2) \right] \tag{5}
\]

Where \( \hat{\omega}_r \) is the estimates value of speed signal; \( \hat{b} \) is the estimates value of identification parameter.

According to the parallel model reference adaptive algorithm, we can refer to the inertia identification algorithm design model as follows:

\[
\begin{align*}
\Delta T_e &= T_e(k-1) - T_e(k-2) \\
\dot{\omega}_r &= 2 \omega_r(k-1) - \omega_r(k-2) + b(k-1) \Delta T_e(k-1) \\
\Delta \omega_r &= \omega_r(k) - \omega_r \\
b(k) &= b(k-1) + \lambda \frac{\Delta T_e(k-1) \Delta \omega_r(k)}{1 + \lambda \Delta T_e(k-1)}
\end{align*} \tag{6}
\]

### 3.2 Rhomberg observer

The theory of state observer is proposed by Rhomberg, and its essence is a state estimator. It uses the input and output of the controlled object to estimate the state of the system, thus resolving the problem that the system can not be directly measured.

Through the motor dynamic equation (2) can get:

\[
T_e = T_L + B \frac{d\omega_r}{dt} + J \frac{d^2\theta_r}{dt^2} \tag{7}
\]

Ignoring the viscous coefficient and the motor without load, here is designed by the Rhomberg observer theory, and a closed-loop observer as follows:

\[
\begin{bmatrix}
\dot{\theta}_r \\
\dot{\omega}_r
\end{bmatrix} =
\begin{bmatrix}
0 & 1 \\
0 & 0
\end{bmatrix}
\begin{bmatrix}
\hat{\theta}_r \\
\hat{\omega}_r
\end{bmatrix} +
\begin{bmatrix}
0 \\
\frac{1}{J}
\end{bmatrix} T_e +
\begin{bmatrix}
K_i \\
K_p
\end{bmatrix} \left( \theta_r - \hat{\theta}_r \right) \tag{8}
\]

### 3.3 Interconnected observer design

From the above, the inertia of the motor is obtained by the model reference adaptive method, the rotor position and the speed of the motor can be obtained by the Rhomberg observer. In the process of inertia identification, rotor position and speed identification, they depend on each other. Identification of the inertia need the rotor position and speed of the rotor; Rotor position tracking and rotor speed identification, need to identify the inertia. Thus, the Interconnected observer can be get by equation (6) and (8):
\[
\begin{align*}
\Delta T_e &= T_e(k-1) - T_e(k-2) \\
\omega_r &= 2\omega_r(k-1) - \omega_r(k-2) + b(k-1)\Delta T_e(k-1) \\
\Delta \omega_r &= \omega_r(k) - \omega_r(k-1) \\
b(k) &= b(k-1) + \frac{\Delta T_e(k-1)\Delta \omega(k-1)}{1 + \lambda\Delta T_e(k-1)} \\
\begin{bmatrix}
\dot{\theta}_r \\
\dot{\omega}_r
\end{bmatrix} &= 
\begin{bmatrix}
0 & 1 \\
0 & 0
\end{bmatrix}
\begin{bmatrix}
\hat{\theta}_r \\
\hat{\omega}_r
\end{bmatrix} + 
\begin{bmatrix}
0 \\
0
\end{bmatrix}
T_e + 
\begin{bmatrix}
K_t \\
K_p
\end{bmatrix}
\left(\hat{\theta}_r - \hat{\theta}_r\right)
\end{align*}
\]  

Fig. 2 simulation block diagram

IV. MATLAB SIMULATION ANALYSIS

The basic parameters of permanent magnet synchronous motor are as follows:

\[R_s = 2.875\Omega; \varphi_f = 0.175 wb; L_d = L_q = 8.5 mH; p = 2; J = 0.01 kg \cdot m^2.\]

The simulation waveforms in MATLAB are shown below:

Observation of the position and speed of the rotor:
Fig. 3 is the rotor position tracking waveform and Fig. 4 is the Rotor speed identification. In the figure of rotor position tracking, the horizontal axis represents time, the vertical axis represents curvature, seen from the waveform, the observer can accurately track the position of the rotor; in the figure of rotor speed identification, the horizontal axis represents time, the vertical axis represents angular velocity, similarly, there is a waveform diagram shows that the observer can accurate identification of the rotational speed.

Moment of inertia identification:
The inertia identification by the interconnected observer:

Figures 5, 6, 7 are the inertia identification in different operating conditions of servo system, respectively. Fig. 5 identification of moments of inertia at constant speed; Fig. 6 identification of moments of inertia when a sinusoidal change occurs at a speed; Fig. 7 identification of moments of inertia when the load torque is changed. In Figure 5, the rotating speed is constant, the Internet observer in a very short period of time to achieve the accurate identification of inertia; In Figure 6, due to the sinusoidal changes in the speed, the overshoot of the
overshoot is generated during the process of identification, but the identification results still converge to the correct value after a period of time; In Figure 7, the load torque in 0.05 seconds time step changes occur. The identification results are similarly affected, but in a very short period of time, rapid convergence to the accurate values.

V. CONCLUSIONS

This paper studies the model reference adaptive and Rhomberg observer. Combining Rhomberg observer and the model reference adaptive to establish the adaptive interconnected observer, and eventually realize the online identification of the inertia of the rotor. In the process of identification, the identification of moment of inertia is carried out by changing the different servo system operating conditions. By the simulation analysis, the adaptive interconnected observer can still identify the moment of inertia accurately when the servo system operating conditions is different.

VI. ACKNOWLEDGEMENTS

My deepest gratitude goes first and foremost to Professor ZENG Guohui, my supervisor, for his constant encouragement and guidance. He has walked me through all the stages of the writing of this thesis. Without his consistent and illuminating instruction, this thesis could not have reached its present form. Also I would like to express my heartfelt gratitude to my project team members, and they gave me help and time in listening to me and helping me work out my problems during the difficult course of the thesis. At the same time, this work is supported by the Innovation foundation of SUES under Grant N0.E1-0903-15-01026.

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