

## 유도전동기의 효율적인 회전자 저항 추정 알고리즘에 관한 연구

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### A Study on Efficient Rotor Resistance Identification Algorithm for Induction Motors

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#### Abstract

We propose a nonlinear feedback controller that can control the induction motors with high dynamic performance by means of decoupling of motor speed and rotor flux. A new recursive adaptation algorithm for rotor resistance which can be applied to our nonlinear feedback controller is also presented in this paper. Some simulation results show that the adaptation algorithm for rotor resistance is robust against the variation of stator resistance and mutual inductance. In addition, it is computationally simple and has small estimation errors.

#### I. INTRODUCTION

Along with the rapid growth in microelectronics and power electronics technologies, various advanced control methods have been successfully implemented in real time and shown to be useful in controlling induction motors with high dynamic performance. In this paper, we propose a nonlinear feedback controller that can control the induction motors with high dynamic performance by means of decoupling of motor speed and rotor flux. Our nonlinear feedback controller needs the information on some motor parameters. Among them, rotor resistance varies greatly with machine temperature.

Some efficient estimation algorithms for rotor resistance can be found in [1]-[7]. The estimation methods in [3] and [5] need the information on stator voltages.

Because PWM inverters are commonly used as the voltage sources for the induction motors, such voltage sensing can deliver some error in the estimation of rotor resistance. In [1], the stator voltages are calculated from the stator voltage commands with dead time compensation instead of using voltage sensors.

We present a new recursive adaptation algorithm for rotor resistance, which seems to have some advantages over the previous methods. For instance, our algorithm does not request voltage sensors.

#### II. DECOUPLING CONTROL OF MOTOR SPEED AND ROTOR FLUX

In this section, we describe our approach to control of induction motors whose dynamic equations are described, in the d-q synchronously rotating frame as

$$\begin{aligned} \dot{i}_{ds} &= -a_1 i_{ds} + \omega_s i_{qs} + a_2 \dot{\phi}_{dr} + p a_3 \omega_r \dot{\phi}_{qr} + a_0 v_{ds} \\ \dot{i}_{qs} &= -a_1 i_{qs} - \omega_s i_{ds} + a_2 \dot{\phi}_{qr} - p a_3 \omega_r \dot{\phi}_{dr} + a_0 v_{qs} \\ \dot{\phi}_{dr} &= -a_4 \phi_{dr} + (\omega_s - p \omega_r) \phi_{qr} + a_5 i_{ds} \\ \dot{\phi}_{qr} &= -a_4 \phi_{qr} - (\omega_s - p \omega_r) \phi_{dr} + a_5 i_{qs} \\ \dot{\omega}_r &= -a_6 \omega_r + a_7 K_T (\phi_{dr} i_{qs} - \phi_{qr} i_{ds}) - a_7 T_L \end{aligned} \quad (1)$$

Here, the constants  $K_T$  and  $a_i$ ,  $i = 0, \dots, 7$  are the parameters of the induction motor. See the Reference[7] for the symbols and notations that appear frequently in our development. The angular speed of rotor flux  $\omega_s$  in eq. (1)

is chosen as

$$\omega_s = p\omega_r + \hat{a}_5 i_{qs} / \hat{\phi}_{dr} \quad (2)$$

from which it follows that the slip speed  $\omega_{sl}$  becomes  $\hat{a}_5 i_{qs} / \hat{\phi}_{dr}$ . If  $\hat{a}_5 (\equiv \hat{M}\hat{R}_r / \hat{L}_r)$  is equal to  $a_5$  and  $\hat{\phi}_{dr}$  is equal to  $\phi_{dr}$ , then the q-axis rotor flux  $\phi_{qr}$  will approach to zero. As the result, the dynamic equations in (1) and (2) can be approximated to

$$\begin{aligned} \dot{i}_{ds} &= -a_1 i_{ds} + \omega_s i_{qs} + a_2 \phi_{dr} + a_0 v_{ds} \\ \dot{i}_{qs} &= -a_1 i_{qs} - \omega_s i_{ds} - p a_3 \omega_r \phi_{dr} + a_0 v_{qs} \\ \dot{\phi}_{dr} &= -a_4 \phi_{dr} + a_3 i_{ds} \\ \dot{\omega}_r &= -a_6 \omega_r + a_7 K_r \phi_{dr} i_{qs} - a_7 T_L \end{aligned} \quad (3)$$

To obtain the information on rotor flux, we adopt the following well-known rotor flux simulator.

$$\dot{\hat{\phi}}_{dr} = -\hat{a}_4 \hat{\phi}_{dr} + \hat{a}_5 i_{ds} \quad (4)$$

If the output to be controlled is chosen as [7]

$$y \equiv \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} \phi_{dr}^2 \\ \omega_r \end{bmatrix} \quad (5)$$

then one can find a nonlinear feedback controller that decouples the reduced system consisting of (3) and (5).

$$u \equiv \begin{bmatrix} v_{ds} \\ v_{qs} \end{bmatrix} = \begin{bmatrix} -(\omega_s i_{qs} + \hat{a}_5 i_{ds}^2 / \hat{\phi}_{dr}) / \hat{a}_0 \\ p\omega_r (i_{ds} + \hat{a}_3 \hat{\phi}_{dr}) / \hat{a}_0 \end{bmatrix} + \bar{u} / \hat{\phi}_{dr} \quad (6)$$

where  $\bar{u} \equiv [\bar{u}_1 \ \bar{u}_2]^T$  is the new input and  $\hat{a}_0, \hat{a}_3, \hat{a}_5$  represent the estimated values of  $a_0, a_3, a_5$ , respectively.

Now, we will show that the input-output dynamic characteristics of the system given by (3) - (6) can be linear. Let the new state variable  $z$  be defined as

$$\begin{aligned} z &= [z_1^T \ z_2^T]^T = [z_{11} \ z_{12} \ z_{21} \ z_{22}]^T \\ &= [\phi_{dr} i_{ds} \ \phi_{dr}^2 \ \phi_{dr} i_{qs} \ \omega_r]^T \end{aligned} \quad (7)$$

If  $\hat{a}_0 = a_0, \hat{a}_3 = a_3, \hat{a}_5 = a_5$ , and  $\hat{\phi}_{dr} = \phi_{dr}$ , then the system given by (3) - (6) can be represented in the new state space

as

$$\begin{aligned} \dot{z} &= \begin{bmatrix} \dot{z}_1 \\ \dot{z}_2 \end{bmatrix} = \begin{bmatrix} A_1 z_1 + b\bar{u}_1 \\ A_2 z_2 + b\bar{u}_2 + LT_L \end{bmatrix} \\ y_i &= cz_i, \quad i = 1, 2. \end{aligned} \quad (8)$$

where

$$\begin{aligned} A_1 &= \begin{bmatrix} -a_1 - a_4 & a_2 \\ 2a_5 & -2a_4 \end{bmatrix}, \\ A_2 &= \begin{bmatrix} -a_1 - a_4 & 0 \\ a_7 K_r & -a_6 \end{bmatrix}, \\ b &= [a_0 \ 0]^T, \quad c = [0 \ 1], \quad L = [0 \ -a_7]^T \end{aligned} \quad (9)$$

The block diagram representation of the decoupled linear system (8) is shown in Fig. 1. Note that the responses of motor speed and rotor flux are dynamically decoupled and linear. Also note that the input-output dynamic characteristics of the system given by (3) - (6) are identical with those of the decoupled linear system (8). So, even if the rotor flux is reduced for power efficiency control at

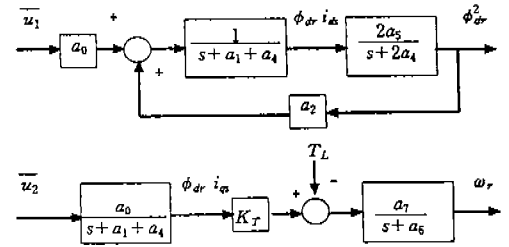


Fig 1. Block diagram of the decoupled linear system

light loads or for field weakening at high speeds, the dynamic performance of the motor speed responses will not be deteriorated because of the complete decoupling of motor speed and rotor flux. For successful set-point tracking of motor speed and rotor flux, the new inputs  $\bar{u}_1$  and  $\bar{u}_2$  are chosen as follows :

$$\begin{aligned} \bar{u}_1 &= k_{p1} (\bar{u}_1 - \hat{\phi}_{dr} i_{ds}) + k_{i1} \int_0^t (\bar{u}_1 - \hat{\phi}_{dr} i_{ds}) dt \\ \bar{u}_2 &= k_{p2} (\bar{u}_2 - \hat{\phi}_{dr} i_{qs}) + k_{i2} \int_0^t (\bar{u}_2 - \hat{\phi}_{dr} i_{qs}) dt \end{aligned} \quad (10)$$

where

$$\begin{aligned}\tilde{u}_1 &= -k_{p3} \hat{\phi}_{dr}^2 + k_{i3} \int_0^t (\phi^* - \hat{\phi}_{dr}^2) dt \\ \tilde{u}_2 &= -k_{p4} \omega_r + k_{i4} \int_0^t (\omega_r^* - \omega_r) dt\end{aligned}\quad (11)$$

Here, constants  $k_{ij}, k_{pi}, j=1, 2, 3, 4$  are controller gains and  $\omega_r^*, \phi^*$  represent the command inputs for  $\omega_r, \hat{\phi}_{dr}^2$ , respectively.

### III. RECURSIVE ADAPTATION OF ROTOR RESISTANCE

As mentioned earlier, if the estimated values  $\hat{a}_0, \hat{a}_3, \hat{a}_4$ , and  $\hat{a}_5$  which are used in the control inputs (2), (6), and the rotor flux simulator (4) are equal to their real values  $a_0, a_3, a_4$ , and  $a_5$ , respectively, the nonlinear dynamic equations of induction motors (1) with the control inputs (2), (6), and the rotor flux simulator (4) can be transformed into the decoupled linear system (8). By the way, the motor parameters  $a_0, a_3, a_4$ , and  $a_5$  are composed entirely of  $M, L_s, L_r$ , and  $R_r$ . Accordingly, the accurate estimation of the motor parameters  $M, L_s, L_r$ , and  $R_r$  is of crucial importance to high dynamic performance. The motor inductances  $M, L_s (=M + L_{so})$ , and  $L_r (=M + L_{ro})$  depend on rotor flux level and do not vary with machine temperature and load conditions. So, they should be exactly known at different rotor flux levels. In practice, they can be obtained by an off-line test. In what follows, we assume that  $\hat{M} = M, \hat{L}_s = L_s$ , and  $\hat{L}_r = L_r$ . This leads us to assume that  $\hat{a}_0 = a_0$  and  $\hat{a}_3 = a_3$ .

On the other hand, rotor resistance varies greatly with machine temperature but is not influenced by rotor flux level. In certain cases, rotor resistance can increase by 100% over its ambient or nominal value. The values of rotor resistance at different machine temperatures can not be obtained by an off-line test and must be estimated instantly during the operation.

Now, we will consider the situation that the rotor resistance is estimated accurately (i.e.  $\hat{R}_r = R_r$ ). Then,  $\hat{\phi}_{dr}$  will be equal to  $\phi_{dr}$ . From the first and second

equations of (1) with the control input (6), we obtain

$$\begin{aligned}i_{ds} &= -a_1 i_{ds} + a_2 \hat{\phi}_{dr} - \hat{a}_5 i_{ds}^2 / \hat{\phi}_{dr} + \hat{a}_0 \bar{u}_1 / \hat{\phi}_{dr} \\ i_{qs} &= -a_1 i_{qs} - \hat{a}_5 i_{ds} i_{qs} / \hat{\phi}_{dr} + \hat{a}_0 \bar{u}_2 / \hat{\phi}_{dr}\end{aligned}\quad (12)$$

Because  $i_{ds}$  and  $i_{qs}$  are zero in the steady-state, this leads to (13).

$$\begin{aligned}i_{ds}^s i_{qs}^s - i_{ds}^s i_{qs}^s &= -a_2 \hat{\phi}_{dr}^s i_{qs}^s + \frac{\hat{a}_0}{\hat{\phi}_{dr}^s} (\bar{u}_2^s i_{ds}^s - \bar{u}_1^s i_{qs}^s) \\ &= 0\end{aligned}\quad (13)$$

where the superscript  $s$  denotes the steady-state value. In the steady-state, the eq. (4) becomes

$$\hat{\phi}_{dr}^s = \hat{M} i_{ds}^s \quad (14)$$

Because it is assumed that  $\hat{L}_s = L_s, \hat{M} = M$ , and  $\hat{L}_r = L_r$ , the motor parameter  $a_2$  can be written as

$$a_2 = a_0 M R_r / L_r^2 = \hat{a}_0 \hat{M} R_r / \hat{L}_r^2 \quad (15)$$

From (13) - (15), we have

$$R_r = \frac{\hat{L}_r^2}{\hat{M}^2 \hat{\phi}_{dr}^s} \left( \frac{\bar{u}_2^s}{i_{qs}^s} - \frac{\bar{u}_1^s}{i_{ds}^s} \right) \quad (16)$$

By the way, the eq. (16) is obtained on the assumption that  $\hat{R}_r$  is equal to  $R_r$ . So, we can have

$$\hat{R}_r = R_r = \frac{\hat{L}_r^2}{\hat{M}^2 \hat{\phi}_{dr}^s} \left( \frac{\bar{u}_2^s}{i_{qs}^s} - \frac{\bar{u}_1^s}{i_{ds}^s} \right) \quad (17)$$

The eq. (17) is effective only in case that  $\hat{R}_r = R_r$ . However, it can be utilized to construct a recursive adaptation algorithm for  $R_r$ . Assume that the  $n$ -th estimated rotor resistance  $\hat{R}_r(n)$  is applied to (2), (4), and (6) and then, in the steady state, the  $(n+1)$ th estimated rotor resistance  $\hat{R}_r(n+1)$  is calculated using (17). Then, we find

that  $\hat{R}_r(n+1)$  is closer to  $R_r$  than  $\hat{R}_r(n)$ , which will be shown in later simulations and experiments.

Now, we describe our recursive adaptation algorithm for  $R_r$ .

Step 1) Calculate  $\hat{R}_r$  using (17) and update  $\hat{a}_4$  and  $\hat{a}_5$  in (2), (4), and (6).

Step 2) Wait until all the state variables reach the steady state.

Step 3) Go to step 1.

Our new recursive adaptation algorithm for rotor resistance has some advantages over the previous methods. As can be seen from (17), it does not need the information on stator voltage, which is required in the prior works [3] and [5]. Therefore, its estimation accuracy is not affected by the use of a PWM inverter as a voltage source. Furthermore, it is independent of stator resistance and stator inductance. However, the performances of our algorithm in (17) depend on the accuracy of  $\hat{M}$  and  $\hat{L}_r$ . In addition, it requires that the stator voltages  $v_{ds}$  and  $v_{qs}$  in (1) should be equal to their commands  $v_{ds}^*$  and  $v_{qs}^*$  in (6), respectively. Some dead-time compensation techniques can solve this problem [1].

#### IV. SIMULATION AND EXPERIMENTAL RESULTS

The performances of our control scheme developed in the preceding sections were studied through some simulations and experiments. For simulation and experimental work, we have chosen a two pole squirrel cage induction motor whose motor data are listed in Table 1.

First, we present the simulation results, which show good performance of our controller with  $R_r$  adaptation. For this aim, the value of  $\hat{R}_r$  was initially assumed to be  $1.425\Omega$ . This corresponds to a 25% estimation error in  $R_r$ . In this simulation, we assumed that  $\hat{R}_s = R_s$ ,  $\hat{M} = M$ , and the induction motor was driven with rated rotor flux, rated load torque, and motor speed of 30r/min. Our

decoupling controller was executed every 0.1msec. At  $t = 1$ sec, our recursive adaptation algorithm in (17) was started. It was calculated and updated every 0.5sec. The simulation result for this case is shown in Fig. 2. We can see that  $\hat{R}_r$  approaches to  $R_r$  step by step. Observe that the motor speed response is disturbed by the abrupt change in  $\hat{R}_r$ . Thus, in the presence of large estimation error in  $R_r$ , our controller fails to decouple the motor speed and the rotor flux.

Second, the recursive adaptation algorithm in (17) was updated every 0.5msec and the rate of change of  $\hat{R}_r$  was restricted within  $0.2\Omega/\text{sec}$ . As can be seen in Fig. 3, the ripple in the motor speed response was decreased. We can see that the slow change in  $\hat{R}_r$  minimize the influence on the motor speed response and it took a shorter time for  $R_r$

Table 1. Data of the induction motor used for simulations and experiments

Nameplate Data	Nominal	Parameters
220 V 50 Hz	$R_s$	$1.09\Omega$
3 phase	$R_r$	$1.14\Omega$
Y connected	$L_s$	100 mH
2 poles	$L_r$	100 mH
Rated power 600W	$L_{ro}(L_{so})$	7.7 mH
Rated speed 3000r/min	M	92.3 mH
Rated rotor flux 0.3Wb	J	$3.2 \times 10^{-4} \text{ kgm}^2$
Rated current 4.2A(r ms)	B	$4.2 \times 10^{-4} \text{ kgm}^2/\text{s}$

adaptation in the case of Fig. 3 than in the case of Fig. 2.

Third, we investigate the robustness of our recursive adaptation algorithm with respect to  $R_s$  and  $M$ . The value of  $\hat{R}_r$  was assumed to be  $1.635\Omega$  and to be unchanged throughout this simulation. This corresponds to a 50% estimation error in  $R_r$ . The simulation result for this case is shown in Fig. 4 (a). This simulation result shows that our recursive adaptation algorithm is perfectly robust against the variation of  $R_s$ . This is because the recursive

adaptation algorithm in (17) does not need the information on  $R_r$ . Now, we will consider the situation that  $\hat{M} = 1.2M$ . We can see from Fig. 4 (b) that the 20% estimation error in  $\hat{M}$  caused about 1.44% increase in the  $R_r$  estimation error. As for this simulation result, we can say that our recursive adaptation algorithm is relatively robust against the uncertainty in  $M$ . This is because the rotor leakage inductance  $L_{ro}$  is negligibly small compared with  $\hat{L}_r (= \hat{M} + L_{ro})$  and  $\hat{L}_r^2$  is divided by  $\hat{M}^2$  in eq. (17).

Fourth, our control algorithm was implemented on a DSP chip TMS320C25. The motor speed was detected by an optical encoder whose resolution was 4000 pulses/rev. The decoupling control algorithm was executed every 0.1msec and the recursive adaptation algorithm in (17) was updated every 0.5msec. The experimental result for this case is shown in Fig. 5. We can see that the ripple in the motor speed response is negligible. As can be seen from Fig. 3 and Fig. 5, the experimental results agree well with the simulation results. Slight differences between the simulation and experimental results are unavoidable due to imperfect hardware implementation, some uncertainties in motor data, and etc.

Finally, we show that our controller with adaptation of rotor resistance can guarantee almost exact decoupling of motor speed and rotor flux. Fig. 6 shows the experimental results for the case of step change in motor speed from 1500r/min to 3000r/min after our  $R_r$  adaptation algorithm was executed. The rotor flux was changed from 0.5p.u. (= 0.0225Wb<sup>2</sup>) to 1.0p.u. (= 0.09Wb<sup>2</sup>) at  $t = 0.4$ sec and from 1.0p.u. to 0.5p.u. at  $t = 1.4$ sec. However, the motor speed response is not affected by the abrupt change in the rotor flux because of the complete decoupling of motor speed and rotor flux.

## V. CONCLUSION

In this paper, a decoupling controller with adaptation of rotor resistance has been proposed. Some simulation results showed that the recursive adaptation algorithm for

rotor resistance proposed in this paper was robust against the variation of stator resistance and mutual inductance. In addition, our adaptation algorithm is computationally simple and has small estimation errors. However, the adaptation algorithm was designed to be applied to our decoupling controller. So, further research will focus on studies related to its application to the general vector control scheme. The mathematical proof for the convergence of the proposed adaptation algorithm is still underway.

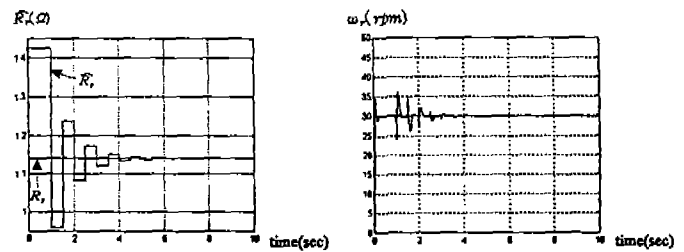


Fig. 2. Simulation of  $R_r$  adaptation and motor speed  $\omega_r$ .

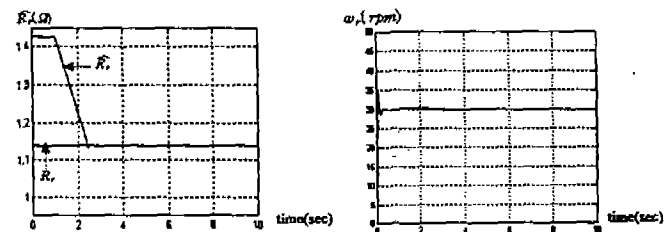


Fig. 3. Simulation of  $R_r$  adaptation with updating period of 0.5msec.

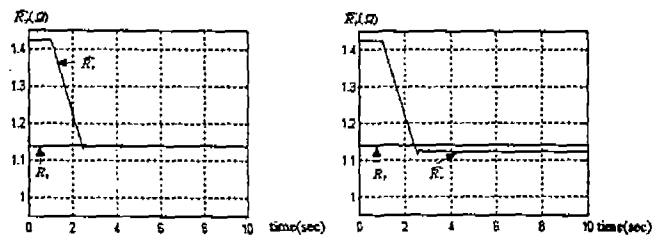


Fig. 4. Simulation of  $R_r$  adaptation (a) when  $\hat{R}_r = 1.5R_r$  and (b) when  $\hat{M} = 1.2M$ .

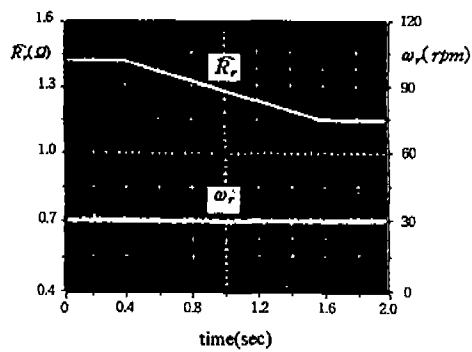


Fig.5. Experimental results of  $R_r$  adaptation and motor speed  $\omega_r$ ,

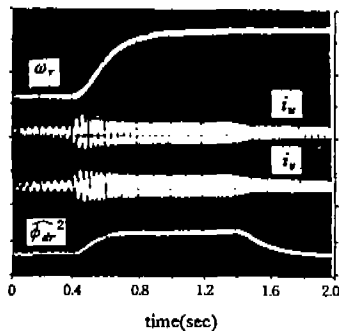


Fig.6. Experimental results of motor speed (700r/min/div), u(v) phase stator currents (10A/div), and rotor flux (0.05 Wb<sup>2</sup>/div).

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