

Induction Motor Flux Estimation using Nonlinear Sliding Observers

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Abstract: A nonlinear sliding flux was proposed for an induction motor. Its dynamic observation errors converge asymptotically to zero, independently from the inputs. The aim of this work was to study the robustness of this observer with respect to the variation of the rotor resistance known to be a crucial parameter for the control. The dynamic performance of this sliding observer was compared to that of Verghese observer via a simulation of an IM driven by U/F control in open loop.

Key words: Nonlinear sliding observer, induction motor, rotor flux estimation

INTRODUCTION

DC motors have been used extensively in the industry because of the simple control strategies required to achieve good performance, in variable speed applications. However, in comparison with their counterparts, IM drives, DC drives result more expensive and less robust devices, not to mention the maintenance they require due to the commutator. Because they are highly nonlinear, thus requiring much more control complex algorithms, IM drives were rarely used in control applications in the past.

Nowadays, as a consequence of the important progress realized in nonlinear control theory and power electronics, the AC drives, by using new control techniques, have proved to outperform the DC ones. Among these techniques, both field oriented control (FOC) and nonlinear input-output decoupling have emerged as powerful tools for high performance control of induction machines^[1,2]. The main drawback of both algorithms is the need of flux sensors, which are to be inserted in the air gap and involve a redesign of the machine, which reduces reliability and implies both additional costs and technological difficulties. For this reason, flux observers have been widely investigated^[3,4]: they are rather sensitive with respect to rotor resistance variations.

Starting with^[5], rotor resistance estimators have been studied^[5,6] but most contributions rely on simplifying assumptions and definite results are still not completely available since, no method applies when the motor is in low-speed regime.

In the literature, several alternative methods exist for the design of different observer structures: linearization by a change of coordinates and output injection^[7-9], variable structure systems^[10] and Lyapunov-based design^[11]. However, closed-loop stability cannot be guaranteed a priori if the control design is based on the separation principle, which is verified only for linear systems. Thus, based on control theory and noting that observability is a dual problem of controllability, sliding mode observers were developed^[12-14]. They derive from a transposition of the switching controllers^[12] to the problem of state observation in nonlinear systems. Sliding control design consists in defining a switching surface in the phase plane that is rendered attractive by the action of the switching terms. The dynamics is determined by the Fillipov solution concept^[15], which indicates that the system dynamic behaviour within the switching surface can be described as a pondered average of the dynamics of each side of the discontinuous surface.

Class of sliding mode observers: Consider the n^{th} - order nonlinear system:

$$\dot{x} = f(x, u), \quad x \in R^n; \quad u \in R^q \quad (1)$$

and for convenience, consider a vector of measurements:

$$y = Cx, \quad y \in R^r \quad (2)$$

The system is assumed to be observable and the observer is defined with the following structure :

$$\dot{\hat{x}} = \hat{f}(\hat{x}, y, u) + KI_s \quad (3)$$

where $\hat{x} \in R^n$, \hat{f} is our model of f , K is $n \times r$ gain matrix to be specified and

$$I_s = [\text{sgn}(s_1), \text{sgn}(s_2), \dots, \text{sgn}(s_r)]^T \quad (4)$$

where

$$[s_1, s_2, \dots, s_r]^T = S = \Gamma[y - C\hat{x}] \quad (5)$$

and Γ is $r \times q$ matrix to be specified. Defining the error vector $\tilde{y} = C\tilde{x}$ and $\tilde{x} = (x - \hat{x})$, one has

$$\dot{\tilde{x}} = \Delta f - KI_s \quad (6)$$

where $\Delta f = f(x, u) - \hat{f}(\hat{x}, y, u)$

The r dimensional surface $S=0$ will be attractive

if: $s_i \dot{s}_i < 0, i \in \{1, \dots, r\}$ During the sliding, the switching term in (4) is keeping $S \equiv 0$; hence, formally $S \equiv 0$. So, \tilde{I}_s , the equivalent switching vector^[16] can be obtained from: $\Gamma C(\Delta f - K\tilde{I}_s) = 0$

$$\text{so that: } \tilde{I}_s = (\Gamma CK)^{-1} \Gamma C \Delta f \quad (7)$$

The $r \times r$ matrix ΓCK is invisible with an appropriate choice for Γ and K . Thus, from (6) and (7) the equivalent dynamics on the reduced order manifold is given by:

$$\dot{\tilde{x}} = (I - K(\Gamma CK)^{-1} \Gamma C) \Delta f \quad (8)$$

with $\Gamma C\tilde{x} = 0$ The structure of Δf must be known before any further analysis can be done

Vergheze observer: The dynamic behaviour of an induction motor working under no saturation of its magnetic circuits can be described in a fixed stator reference $(\alpha - \beta)$ frame^[2] by:

$$\begin{cases} d\phi_{\alpha,r} / dt = -b\phi_{\alpha,r} + ai_{\alpha,s} - \omega p\phi_{\beta,r} \\ d\phi_{\beta,r} / dt = -b\phi_{\beta,r} + ai_{\beta,s} + \omega p\phi_{\alpha,r} \\ di_{\alpha,s} / dt = \gamma_4 V_{\alpha,s} - \gamma_1 i_{\alpha,s} + \gamma_2 \phi_{\alpha,r} + \gamma_3 \omega \phi_{\beta,r} \\ di_{\beta,s} / dt = \gamma_4 V_{\beta,s} - \gamma_1 i_{\beta,s} + \gamma_2 \phi_{\beta,r} - \gamma_3 \omega \phi_{\alpha,r} \\ J_r d\omega / dt = p(M / L_r)[i_{\beta,s} \phi_{\alpha,r} - i_{\alpha,s} \phi_{\beta,r}] - T_L + f_v \omega \end{cases} \quad (9)$$

with $\alpha_1 = pM / J_r L_r$,

$b = R_r / L_r = 1 / T_r$ $\sigma = 1 - (M^2 / L_s L_r)$, $a = R_r M / L_r$

$\gamma_1 = (R_s / \sigma L_s) + (R_r M^2 / \sigma L_s L_r^2)$ $\gamma_2 = MR_r / \sigma L_s L_r^2$,

$\gamma_3 = Mp / \sigma L_s L_r$, $\gamma_4 = 1 / \sigma L_s$

where: $(\phi_{\alpha,r}, \phi_{\beta,r}), (i_{\alpha,s}, i_{\beta,s}), T_L, J_r, \sigma, f_v,$

$R_r, L_r, R_s, L_s, M, T_r, \omega$ and p are respectively the rotor fluxes, the stator currents, the torque load, the moment of inertia, the leakage and sticky friction coefficients, the rotor and stator winding resistances and inductances, the mutual inductance, the rotor time

constant, the mechanical speed and the number of pole pairs. Therefore, setting $x = (\omega, \phi_{\alpha,r}, \phi_{\beta,r}, i_{\alpha,s}, i_{\beta,s})^T$, (9)

is written in the form:

$$\begin{cases} \dot{x}_1 = \alpha_1(x_5 x_2 - x_3 x_4) - \alpha_2 T_L - \alpha_3 x_1 \\ \dot{x}_2 = ax_4 - bx_2 - px_1 x_3 \\ \dot{x}_3 = ax_5 - bx_3 + px_1 x_2 \\ \dot{x}_4 = -\gamma_1 x_4 + \gamma_2 x_2 + \gamma_3 x_1 x_3 + \gamma_4 v_1 \\ \dot{x}_5 = -\gamma_1 x_5 + \gamma_2 x_3 - \gamma_3 x_1 x_2 + \gamma_4 v_2 \end{cases} \quad (10)$$

The Vergheze observer model which is a copy of the first four equations of (9) where added a corrective term due to a prediction error, is written in compact form as^[3]:

$$\begin{bmatrix} \dot{i}'_s \\ \dot{\phi}'_r \end{bmatrix} = \begin{bmatrix} -\gamma_1 I & (M/cT_r)I \\ (M/T_r)I & (-1/T_r)I \end{bmatrix} \begin{bmatrix} \hat{i}_s \\ \hat{\phi}_r \end{bmatrix} + \begin{bmatrix} (L_r/c) \\ 0 \end{bmatrix} v_s + \begin{bmatrix} k_1 I + k_2 \omega_e J \\ k_3 I + k_4 \omega_e J \end{bmatrix} (i_s - \hat{i}_s)$$

where

$$\phi_r = [\phi_{\alpha,r} \ \phi_{\beta,r}]^T, i_s = [i_{\alpha,s} \ i_{\beta,s}]^T, v_s = [V_{\alpha,s} \ V_{\beta,s}]^T, I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, J = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}, c = \sigma L_s L_r$$

$(\hat{i}'_s, \hat{\phi}'_r)$ are the derivatives of $(\hat{i}_s, \hat{\phi}_r)$ estimates of (i_s, ϕ_r) , k_i are scalars and $\omega_e = p\omega$ is the electrical speed of the rotor. The dynamics of the observation error $e = \hat{x} - x$ is given by :

$$\dot{e} = \begin{bmatrix} (k_1 - \gamma_1)I & (M/cT_r)I \\ [k_3 + (M/T_r)]I & (-1/T_r)I \\ +\omega_e \begin{bmatrix} k_2 J & -(M/c)J \\ k_4 J & J \end{bmatrix} \end{bmatrix} e.$$

If k_1 and k_3 are selected such that: $k_1 - \gamma_1 = -k_2 / T_r$ and $k_3 + (M/T_r) = -k_4 / T_r$

the error dynamics becomes : $\dot{e} = A Q e$ (11)

$$\text{where: } A = \begin{bmatrix} k_2 I & -(M/c)I \\ k_4 I & I \end{bmatrix}$$

$$\text{and: } Q = \begin{bmatrix} (-1/T_r)I + \omega_e J & 0 \\ 0 & (-1/T_r)I + \omega_e J \end{bmatrix}$$

The freedom that one has in choosing k_2 and k_4 is used to place the eigenvalues of A in pairs at arbitrary locations, as is verified by noting that the characteristic polynomial of A is:

$$[p^2 - (1 + k_2)p + k_2 + k_4(M/c)]^2$$

If the eigenvalues of A are p_1 (twice) and p_2 (twice), then the eigenvalues of the matrix product in (11) can be shown to be:

$$\left[(-1/T_r) \pm j\omega_e\right] p_1 \text{ and } \left[(-1/T_r) \pm j\omega_e\right] p_2$$

Hence, if the speed is (nearly) constant, the error dynamics is (approximately) governed by these eigenvalues. If it is time-varying, we will attempt a Lyapunov analysis.

Flux sliding mode observer: The proposed type of sliding mode based observer of (9) can be written as:

$$\begin{cases} \dot{\hat{x}}_1 = \alpha_1(x_5\hat{x}_2 - \hat{x}_3x_4) - \alpha_2T_L - \alpha_3x_1 + K_1I_s + q_1(x_1 - \hat{x}_1) \\ \dot{\hat{x}}_2 = ax_4 - b\hat{x}_2 - px_1\hat{x}_3 + K_2I_s \\ \dot{\hat{x}}_3 = ax_5 - b\hat{x}_3 + px_1\hat{x}_2 + K_3I_s \\ \dot{\hat{x}}_4 = -\gamma_1x_4 + \gamma_2\hat{x}_2 + \gamma_3x_1\hat{x}_3 + \gamma_4v_1 + K_4I_s \\ \dot{\hat{x}}_5 = -\gamma_1x_5 + \gamma_2\hat{x}_3 - \gamma_3x_1\hat{x}_2 + \gamma_4v_2 + K_5I_s \end{cases}$$

where $K_iI_s = \lambda_{i1}sign(s_1) + \lambda_{i2}sign(s_2)$. for $i \in \{1, \dots, 5\}$, K_i and q_i are the observers gains. The sliding surface S is given by:

$$S = M \begin{bmatrix} x_4 - \hat{x}_4 \\ x_5 - \hat{x}_5 \end{bmatrix} = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

Setting $e_i = x_i - \hat{x}_i$ for $i \in \{1, \dots, 5\}$, the observation error dynamics is:

$$\begin{cases} \dot{e}_1 = \alpha_1(x_5e_2 - e_3x_4) - K_1I_s - q_1e_1 \\ \dot{e}_2 = -be_2 - px_1e_3 - K_2I_s \\ \dot{e}_3 = -be_3 + px_1e_2 - K_3I_s \\ \dot{e}_4 = \gamma_2e_2 + \gamma_3x_1e_3 - K_4I_s \\ \dot{e}_5 = \gamma_2e_3 - \gamma_3x_1e_2 - K_5I_s \end{cases}$$

The stability analysis consists of determining K_4 and K_5 such that the surface $S \equiv 0$ is the attractive. Then K_1 , K_2 , K_3 and q_1 are determined such that the reduced order system obtained when $S \equiv 0$ is locally stable to 0 in the attractive domain defined as follows:

Let us consider the Lyapunov function $v = \frac{S^T S}{2}$ such that ($S=0 \rightarrow e_4 = e_5 = 0$) with M as a regular matrix. The attractive of sliding surface $S=0$ is given by:

$$\forall S \neq 0, \dot{v} = S^T \frac{\partial S}{\partial t} = S^T M W < 0 \tag{12}$$

where: $W = M^{-1} \begin{bmatrix} e_2 \\ e_3 \end{bmatrix} - \begin{bmatrix} \lambda_{41} & \lambda_{42} \\ \lambda_{51} & \lambda_{52} \end{bmatrix} I_s$ with: $\begin{bmatrix} K_4 \\ K_5 \end{bmatrix} = M^{-1} \Delta$

$$\Delta = \begin{bmatrix} \delta_1 & 0 \\ 0 & \delta_2 \end{bmatrix}, \delta_1, \delta_2 > 0.$$

From the singular perturbation theory^[14], the dynamics of ω is supposed to be a slow variable with respect to the currents and the flux dynamics. The conditions of attraction between s_1, e_2 and s_2, e_3 are decoupled. So (12) is obtained within the set defined by the following inequalities:

$$\begin{cases} \text{if } s_1 > 0 \text{ then } e_2 < \delta_1 \\ \text{if } s_1 < 0 \text{ then } e_2 > -\delta_1 \\ \text{if } s_2 > 0 \text{ then } e_3 < \delta_2 \\ \text{if } s_2 < 0 \text{ then } e_3 > -\delta_2 \end{cases}$$

On the sliding surface, $S=0$ which is invariant, the vector $I_s = \tilde{I}_s$ is given by:

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix} = M^{-1} \left[\begin{pmatrix} e_2 \\ e_3 \end{pmatrix} - \begin{pmatrix} \delta_1 & 0 \\ 0 & \delta_2 \end{pmatrix} I_s \right] \text{ and } \tilde{I}_s = \begin{pmatrix} e_2 / \delta_1 \\ e_3 / \delta_2 \end{pmatrix} = \Delta^{-1} e$$

From the definition of equivalent vector^[16], one obtains the dynamics error after a finite time t_0 which is reduced

$$\text{to: } \dot{e}_1 = \alpha_1(x_5e_2 - e_3x_4) - (\lambda_{11}/\delta_1 \quad \lambda_{12}/\delta_2) \begin{pmatrix} e_2 \\ e_3 \end{pmatrix} - q_1e_1$$

$$\begin{pmatrix} \dot{e}_2 \\ \dot{e}_3 \end{pmatrix} = \left[H - \begin{pmatrix} \lambda_{21} & \lambda_{22} \\ \lambda_{31} & \lambda_{32} \end{pmatrix} \Delta^{-1} \right] \begin{pmatrix} e_2 \\ e_3 \end{pmatrix}, H = \begin{pmatrix} -b & -px_1 \\ px_1 & -b \end{pmatrix}$$

Thus: $[\lambda_{11}/\delta_1 \quad \lambda_{12}/\delta_2] = [\alpha_1x_5 \quad \alpha_1x_4]$ and :

$$\begin{pmatrix} \lambda_{21} & \lambda_{22} \\ \lambda_{31} & \lambda_{32} \end{pmatrix} = \left[H + \begin{pmatrix} q_2 & 0 \\ 0 & q_3 \end{pmatrix} \right] \Delta$$

The reduced system of the observer errors can be written as: $\dot{e}_i = -q_i e_i$ with $q_i > 0$ and $i=1,2,3$. The observation dynamics error is then stable, with an appropriate choice of q_i

Digital simulation: A combination of the two observers presented in Sections 3 and 4, is simulated in Matlab/Simulink for the simultaneous estimation of rotor fluxes. The combined estimation is depicted in the block diagram of Fig. 1

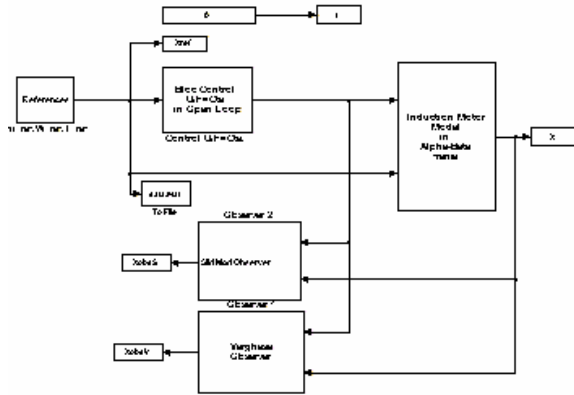


Fig. 1: Flux observers for induction motor

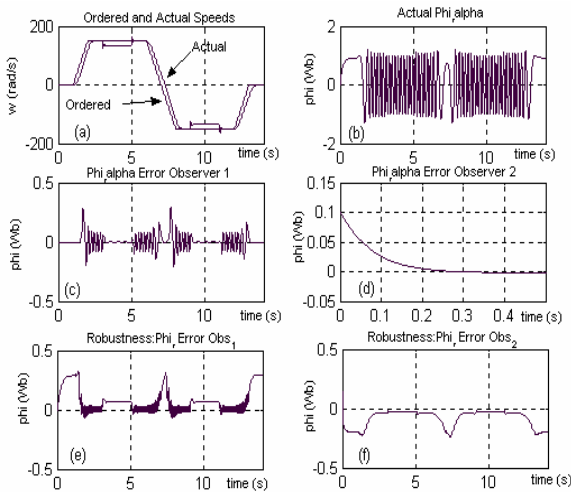


Fig. 2: Simulation results

A digital simulation of the proposed combined scheme illustrates its behaviour when the motor is operating under U/F control in open loop with full load capability (The IM data in simulation are given in the Appendix).

Figure 2a shows the reference speed of the motor with the measured one and Fig. 2b shows the first component of the rotor flux .

From Fig. 2c, we verify that observation errors converge to zero in steady-state operations but, as stated previously, they tend to their maximum value, in low-speed regime, particularly at zero-crossing. Figure 2d illustrates the well-known insensitivity property of sliding modes with respect to disturbances.

Figure 2e and 2f show the robustness of the two observers in case of an instantaneous rotor resistance increase of 50% in the IM drive.

CONCLUSION

In this study, a novel combined scheme for concurrent estimation of the rotor fluxes of an IM was presented. The proposed method is based on two nonlinear observers. The simulation results show a good performance of the sliding mode estimation scheme.

It has been shown that sliding mode observer design methods based on the prescribed form of the Lyapunov function candidate can be successfully applied. The simulation results are suggesting that design can be implemented based on the mechanical motion measurement only, thus avoiding flux variable measurement. In addition, the simplicity of the algorithm makes it suitable for an on-line implementation.

In further work, the authors intend to study high-order sliding modes, in both control and observation for an induction machine and a pneumatic robot, to remove the chattering effect which is known to be the main drawback of the standard sliding modes.

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Appendix

Machines parameters

Total Stator Inductance	0.142 H
Total Rotor Inductance	0.076 H
Mutual Inductance	0.099 H
Stator Resistance	1.633Ω
Rotor Resistance	0.93 Ω
Rotor Inertia (IM + load)	.029 Kg ^m ²
Number of Pole pairs	2

Rated magnitudes

Direct voltage	450 V
Load Torque	+7 & -7Nm
Speed	1430 rpm
Stator Flux	0.59 Wb
Power	1.5 Kw
Coefficient of sticky friction	0.0038 Nms/rd

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