ADAPTIVE FEEDBACK LINEARIZATION CONTROL FOR ASYNCHRONOUS MACHINE WITH NONLINEAR FOR NATURAL DYNAMIC COMPLETE OBSERVER

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This paper studies the nonlinear adaptive control of an induction motor with natural dynamic complete nonlinear observer. The aim of this work is to develop a nonlinear control law and adaptive performance for an asynchronous motor with two main objectives: to improve the continuation of trajectories and the stability, robustness to parametric variations and disturbances rejection. This control law will independently control the speed and flux into the machine by restricting supply. A complete nonlinear observer for dynamic nature ensuring closed loop stability of the entire control and observer has been developed. Several simulations have also been carried out to demonstrate system performance.

K e y w o r d s: f feedback linearization, adaptive control, parameter estimation, nonlinear observer (NLO), induction motor (IM)

1 INTRODUCTION

The nonlinear control based on the technique of linearization within the meaning of the input-outputs proved reliable on the level of the control of the asynchronous machine [2–7]. It was demonstrated that the nonlinear controller is adept nonlinearities and maintaining its performance in a wide operating range long and as long as the machine parameters are not changed. However, it completely loses its performance when the model of the machine is subject to uncertainties in the parameters. In the case of the induction machine, the uncertain parameters are mainly the stator and rotor resistance (depending on temperature), inductors (which depend on the level of saturation), the moment of inertia and load torque (which are difficult to quantify). In addition, by examining the model, one can notice that two of these parameters fall in a linear fashion in the model (resistance and load torque) while the other two returned from a nonlinear fashion (inductors and timing of inertia). At first, researchers tried to solve this problem by developing algorithms for identification of uncertain parameters that tend to change during operation [8]. However, despite the results obtained, they were discarded because of their complexity. Parallel to this, many researchers have been developing methodologies for adaptive nonlinear linearization technique combining with adaptive methods [9–14]. All these methods have led to satisfactory results, but they limit the model structures and how they depend on uncertain parameters. Indeed, most methods proposed in the literature are for the case where the uncertain parameters of a linear fit in the model [15]. The algorithm

that we propose to apply in our case is the one proposed in [1]. This algorithm is not limited to the case where the uncertain parameters of linear fit in the model but it also applies to models with uncertain parameters returned from a nonlinear fashion. In this article we present the principle of the linearization adaptive input-output intended for the control of the asynchronous machine or the parameters return in a linear way in the model of the machine (resistances and the torque of load). We then used a nonlinear observer of flux and current. The performances of the nonlinear controller adaptive will be discussed by digital simulation.

2 ADAPTIVE NONLINEAR CONTROL OF IM

In this section, one carries out an adaptive nonlinear order which ensures the regulation the speed and the flux of the asynchronous machine as well as decoupling between the latter. In this part, one will hold account only parameters which return in a linear way in the model of the machine. It is about rotor resistance R_r and the torque of load T_L . To be done, one starts by designing a controller based on the technique of linearization within the meaning of the input-output applied to the nominal model, then one calculation the law of adaptation which will make it possible to estimate the vector of the dubious parameters [8, 9].

The dynamics of an induction motor under the assumptions of equal mutual inductances and linear mag-

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netic are given by the fifth-order model.

$$\frac{\mathrm{d}x_1}{\mathrm{d}t} = -\lambda x_1 + \frac{k}{T_r} x_3 + p k x_4 x_5 + \frac{1}{\sigma} u_{s\alpha} ,$$

$$\frac{\mathrm{d}x_2}{\mathrm{d}t} = -\lambda x_2 + \frac{k}{T_r} x_4 + p k x_3 x_5 + \frac{1}{\sigma} u_{s\beta} ,$$

$$\frac{\mathrm{d}x_3}{\mathrm{d}t} = \frac{M}{T_r} x_1 - \frac{1}{T_r} x_3 + p x_4 x_3 ,$$

$$\frac{\mathrm{d}x_4}{\mathrm{d}t} = \frac{M}{T_r} x_2 - \frac{1}{T_r} x_4 + p x_5 x_3 ,$$

$$\frac{\mathrm{d}x_5}{\mathrm{d}t} = p \frac{M}{JL_r} (x_2 x_3 - x_1 x_4) - \frac{1}{J} T_L$$
(1)

with

$$\begin{split} \lambda &= \frac{R_s}{\sigma} \frac{1}{L_s} + \frac{M^2}{\sigma} \frac{R_r}{L_s L_r^2} , \ ; \sigma &= 1 - \frac{M^2}{L_s L_r} , \\ T_r &= \frac{L_r}{R_r} , \quad K = \frac{M}{\sigma L_s L_r} . \end{split}$$

We will drop the subscripts r and s since we will only use rotor fluxes $\Phi_{r\alpha}$, $\Phi_{r\beta}$) and stator currents $(is\alpha, i_{s\beta})$. Let

$$x = (x_1, x_2, x_3, x_4, x_5)^\top = (i_{s\alpha}, i_{s\beta}, \Phi_{r\alpha}, Phi_{r\beta}, \Omega)^\top.$$

Considering the vector of uncertain parameters

$$\delta = \begin{bmatrix} \delta_1 \\ \delta_2 \end{bmatrix} = \begin{bmatrix} T_L - T_{LN} \\ R_r - R_{rN} \end{bmatrix}$$
(2)

one can rewrite the equations of the asynchronous machine in the form suggested by

$$\dot{x} = f(x,\delta) + \sum_{i=1}^{2} g_i(x,\delta)u_i$$
$$= f(x) + \sum_{j=1}^{2} \delta_j f_j(x) + \sum_{i=1}^{2} g_i(x)u_i \quad (3)$$

with

$$f(x) = \begin{bmatrix} -\lambda x_1 + \frac{k}{T_r} x_3 + pkx_4 x_5 \\ -\lambda x_2 + \frac{k}{T_r} x_4 - pkx_3 x_5 \\ \frac{M}{T_r} x_1 - \frac{1}{T_r} x_3 - px_4 x_3 \\ \frac{M}{T_r} x_2 + px_3 x_5 - \frac{1}{T_r} x_4 \\ p \frac{M}{JL_T} (x_2 x_3 - x_1 x_4) - \frac{T_L}{J} \end{bmatrix}, \qquad (4)$$

$$g(x) = \begin{bmatrix} g_1(x) & g_2(x) \end{bmatrix} = \begin{bmatrix} \frac{1}{\sigma L_s} & 0 \\ 0 & \frac{1}{\sigma L_s} \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}, \qquad (4)$$

$$f_1(x) = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ -\frac{1}{J} \end{bmatrix}, \qquad f_2(x) = \begin{bmatrix} -\frac{M}{\sigma L_s L_r^2} x_1 + \frac{M}{\sigma L_s L_r^2} x_3 \\ -\frac{M}{\sigma L_s L_r^2} x_2 + \frac{M}{\sigma L_s L_r^2} x_4 \\ \frac{M}{L_r} x_1 - \frac{1}{L_r} x_3 \\ \frac{M}{L_r} x_2 - \frac{1}{T_r} x_4 \\ 0 \end{bmatrix}. \qquad (5)$$

In order to increase the representation, we omit the dependence on x. the derivatives of the outputs become

$$\begin{split} \dot{y}_1 &= y_2 + \delta_1 L_{f_1} h_1 ,\\ \dot{y}_2 &= L_f^2 h_1 + \delta_2 L_{f_2} h_1 + L_{g_1} L_f h_1 u_1 + L_{g_2} L_f h_1 u_2 ,\\ \dot{y}_3 &= y_4 + \delta_2 L_{f_2} h_2 , \end{split} \tag{6} \\ \dot{y}_4 &= L_f^2 h_2 + \delta_2 L_{f_2} L_f h_2 + L_{g_1} L_f h_2 u_1 + L_{g_1} L_f h_2 u_1 ,\\ \dot{y}_5 &= L_f h_3 + \delta_2 L_{f_2} h_3 . \end{split}$$

We can make a change of coordinates (diffeomorphism) using the estimated parameters $\hat{\delta} = [\hat{\delta}_1, \hat{\delta}_2]^{\top}$. The errors on these parameters are

$$e_{\delta} = \begin{pmatrix} e_{\delta_1} \\ e_{\delta_2} \end{pmatrix} = \begin{pmatrix} \delta_1 - \hat{\delta}_1 \\ \delta_2 - \hat{\delta}_2 \end{pmatrix}.$$
 (7)

Now we introduce the estimated parameters $\hat{\delta}$ in the new coordinates z

$$z_{1} = y_{1},$$

$$z_{2} = y_{2} + \hat{\delta}_{1}L_{f_{1}}h_{1},$$

$$z_{3} = y_{3},$$

$$z_{4} = y_{4} + \hat{\delta}_{2}L_{f_{2}}h_{2},$$

$$z_{5} = y_{5}.$$
(8)

The system (8) becomes

$$\begin{aligned} \dot{z}_{1} &= z_{2} + e_{\delta_{1}}L_{f_{1}}h_{1}, \\ \dot{z}_{2} &= L_{f}^{2}h_{1} + \delta_{2}L_{f_{2}}L_{f}h_{1} + \frac{\mathrm{d}\hat{\delta}_{1}}{\mathrm{d}t}L_{f_{1}}h_{1} \\ &+ L_{g_{1}}L_{f}h_{1}u_{1} + L_{g_{2}}L_{f}h_{1}u_{2}, \\ \dot{z}_{3} &= z_{4} + e_{\delta_{2}}L_{f_{2}}h_{2}, \\ \dot{z}_{4} &= L_{f}^{2}h_{2} + \delta_{2}L_{f_{2}}L_{f}h_{2} + \frac{\mathrm{d}\hat{\delta}_{2}}{\mathrm{d}t}L_{f_{2}}h_{2} + \hat{\delta}_{2}L_{f}L_{f_{2}}h_{2} \\ &+ \hat{\delta}_{2}\delta_{2}L_{f_{2}}^{2}h_{2} + u_{1}(L_{g_{1}}L_{f}h_{2} + \hat{\delta}_{2}L_{g_{1}}L_{f_{2}}h_{2}) \\ &+ u_{2}(L_{g_{2}}L_{f}h_{2} + \hat{\delta}_{2}L_{g_{1}}L_{f_{2}}h_{2}), \\ \dot{z}_{5} &= L_{f}h_{3} + \delta_{2}L_{f_{2}}h_{3}). \end{aligned}$$

$$(9)$$

Linearizing control, with the estimated parameters, is given by

$$\begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = D(x, \hat{\delta}_2)^{-1} \begin{bmatrix} -\hat{\xi}_1 + v_1 \\ -\hat{\xi}_2 + v_2 \end{bmatrix}$$
(10)

where

$$\xi_{1} = L_{f}^{2}h_{1} + \hat{\delta}_{2}L_{f_{2}}L_{f}h_{1} + \frac{d\hat{\delta}_{1}}{dt}L_{f_{1}}h_{1},$$

$$\xi_{2} = L_{f}^{2}h_{2} + \hat{\delta}_{2}L_{f_{2}}L_{f}h_{2} + \frac{d\hat{\delta}_{2}}{dt}L_{f_{2}}h_{2}$$

$$+ \hat{\delta}_{2}L_{f}L_{f_{2}}h_{2} + \hat{\delta}_{2}^{2}L_{f_{2}}h_{2}$$
(11)

and

$$D(x, \delta_2) = \begin{bmatrix} L_{g_a} L_f h_1 & L_{g_b} L_f h_1 \\ L_{g_a} L_f h_2 + \hat{\delta}_2 L_{g_a} L_{f_2} h_2 & L_{g_b} L_f h_2 + \hat{\delta}_2 L_{g_b} L_{f_2} h_2 \end{bmatrix}.$$
 (12)
New orders are stabilizing

 $v_1 = -k_{11}(z_1 - z_{1ref}) - k_{12}z_2,$ $v_2 = -k_{21}(z_3 - z_{3ref}) - k_{22}z_4.$ (13)

Control parameters $(k_{11}, k_{12}, k_{21}, k_{22})$ are chosen so that

$$K_1 = \begin{bmatrix} 0 & 1 \\ -k_{11} & -k_{12} \end{bmatrix}, \ K_2 = \begin{bmatrix} 0 & 1 \\ -k_{21} & k_{22} \end{bmatrix}$$
(14)

are asymptotically stable. The matrix gains K_1 and K_2 are given by (14). Then the tracking error is given by

$$e = \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \end{bmatrix} = \begin{bmatrix} z_1 - z_{1d} \\ z_2 - z_{2d} \\ z_3 - z_{3d} \\ z_4 - z_{4d} \end{bmatrix}.$$
 (15)

Dynamics become

$$\dot{e}_{1} = e_{2} + e_{\delta 1}L_{f_{1}}h_{1},$$

$$\dot{e}_{2} = -k_{11}e_{1} - k_{12}e_{2} + e_{\delta 2}L_{f_{2}}L_{f}h_{1},$$

$$\dot{e}_{3} = e_{4} + e_{\delta 2}L_{f_{2}}h_{2},$$
(16)

$$\dot{e}_4 = -k_{21}e_3 - k_{22}e_4 + e_{\delta 2}(L_{f_2}L_fh_2 + \hat{\delta}_2L_{f_2}^2h_2).$$

Or in matrix form

$$\begin{pmatrix} \dot{e}_1 \\ \dot{e}_2 \\ \dot{e}_3 \\ \dot{e}_4 \end{pmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -k_{11} & -k_{12} & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & -k_{21} & -k_{22} \end{bmatrix} \begin{pmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \end{pmatrix} + \begin{bmatrix} L_{f_1}h_1 & 0 \\ 0 & L_{f_2}L_fh_1 \\ 0 & L_{f_2}h_2 \\ 0 & L_{f_2}L_fh_2 + \dot{\delta}_2L_{f_2}^2h_2 \end{bmatrix} \begin{pmatrix} e_{\delta_1} \\ e_{\delta_2} \end{pmatrix}.$$
(17)

In the condensed form we have

$$\dot{e} = Ke + W(z, zhatzdelta_2)e_{\delta}$$
(18)

$$e_{\delta} = [e_{\delta_1}, e_{\delta_2}]^{\top}; \ e = [e_1, e_2, e_3, e_4]^{\top},$$

$$K = \text{bloc} \operatorname{diag}(K_1, K_2)$$
(19)

and

with

$$W(z,\hat{\delta}) = \begin{bmatrix} L_{f_1} & 0 \\ 0 & L_{f_2}L_fh_1 \\ 0 & L_{f_2}h_2 \\ 0 & L_{f_2}L_fh_2 + \hat{\delta}_2L_{f_2}^2h_2 \end{bmatrix}.$$
 (20)

3 NATURAL DYNAMICS COMPLETE OBSERVER

We develop a nonlinear control law with a comprehensive observer. According to [1] we use the complete model of the machine, which implies a relative degree one and two of the output. We develop a control law that controls the torque and flux of the machine (see Fig. 1). We use an estimator with zero gain observers. The control uses only quantities estimated.



Fig. 1. Nonlinear control with dynamic natural observer n

$$Y(x) = \begin{bmatrix} \hat{h}_1(x) \\ \hat{h}_2(x) \end{bmatrix} = \begin{bmatrix} p \frac{M}{L_r} (\hat{x}_2 \hat{x}_3 - \hat{x}_1 \hat{x}_4) \\ \hat{x}_3^2 + \hat{x}_4^2 \end{bmatrix}, \quad (21)$$

$$\hat{h}_3(x) = 2\frac{M}{T_r} (\hat{x}_3 \hat{x}_1 + \hat{x}_4 \hat{x}_2).$$
(22)

3.1 Observer development

The machine model that uses the model with four streams flux it self particularly well to find a Lyapunov function which can show the stability of the estimator. We choose this representation only for easier demonstration of the stability of the observer. The current estimates could be obtained easily with a transformation matrix from invariant flux estimated.

$$\hat{\overline{X}} = A_2(\Omega)\hat{\overline{X}} + Bu,$$

$$\hat{\overline{X}} = \left[\hat{\Phi}_{s\alpha}, \hat{\Phi}_{s\beta}, \hat{\Phi}_{r\alpha}, \hat{\Phi}_{r\beta}\right]^{\top},$$

$$A(\Omega) = -RL^{-1} + p\Omega \begin{bmatrix} 0I & 0I\\ 0I & J \end{bmatrix},$$

$$\hat{A}(\Omega) = -RL^{-1} + p\Omega \begin{bmatrix} 0I & 0I\\ 0I & J \end{bmatrix},$$

$$\hat{A}(\Omega) = -RL^{-1} + p\Omega \begin{bmatrix} 0I & 0I\\ 0I & J \end{bmatrix},$$

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$$\hat{A}(\Omega) = -RL^{-1} + p\Omega \begin{bmatrix} 0I & 0I\\ 0I & J \end{bmatrix},$$

$$B = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}, I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, J = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix},$$
$$R = \begin{bmatrix} R_s I & 0I \\ 0I & R_r I \end{bmatrix}, L = \begin{bmatrix} L_s I & MI \\ MI & L_r I \end{bmatrix}.$$
(24)

We define the estimation errors

$$\overline{e} = [e_4, e_5, e_6, e_7] . \tag{25}$$

Let us choose the function of Lyapunov V_1 following [1]

$$V_1 = \frac{1}{2} \overline{e}^\top R^{-1} \overline{e} \,. \tag{26}$$

Its derivative is equal to

$$\dot{V}_{1} = \frac{1}{2} \overline{e}^{\top} \Big(-(RL^{-1})^{\top} R^{-1} + p\Omega \begin{bmatrix} 0I & 0I \\ 0I & J \end{bmatrix}^{\top} R^{-1} \\ + R^{-1} RL^{-1} + R^{-1} p\Omega \begin{bmatrix} 0I & 0I \\ 0I & J \end{bmatrix} \Big) \overline{e} = -\overline{e}^{\top} (L^{-1}) \overline{e} \quad (27)$$

with L^{-1} positive definite (det $L = L_s L_r \sigma$; $0 < \sigma < 1$). This proves the stability of the observer.



Fig. 2. Decoupled with AIOL

3.2 Control development with observer

We can develop the control law using the observer and the mechanical equation

$$\begin{bmatrix} \dot{\overline{X}} \\ \Omega = x_5 \end{bmatrix} = \begin{bmatrix} A(\Omega)\hat{\overline{X}} \\ \frac{1}{L} \left(p \frac{M}{L_r} (x_3 x_2 - x_4 x_1) - C_{\text{res}} \right) \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} u.$$
(28)

On this model we have the system of tracking errors of torque and flux following

$$\begin{bmatrix} \dot{e}_1\\ \dot{e}_2\\ \dot{e}_3 \end{bmatrix} = \begin{bmatrix} L_f \hat{h}_1\\ L_f \hat{h}_2\\ L_f \hat{h}_3 \end{bmatrix} + \begin{bmatrix} L_g \hat{h}_1\\ 0\\ L_g \hat{h}_3 \end{bmatrix} u - \begin{bmatrix} \dot{h}_{1\text{ref}}\\ \dot{h}_{2\text{ref}}\\ \dot{h}_{3\text{ref}} \end{bmatrix}.$$
(29)

To obtain a further output flux and torque, choose

$$u = \begin{bmatrix} L_g \hat{h}_1 \\ L_g \hat{h}_3 \end{bmatrix}^{-1} \begin{bmatrix} -L_f \hat{h}_1 - k_1 e_1 + \hat{h}_{1\text{ref}} \\ -L_f \hat{h}_3 - k_3 e_3 + \hat{h}_{3\text{ref}} - e_2 \end{bmatrix}.$$
 (30)

By applying the command (30) we obtain the following system errors stability of the control.

$$\begin{bmatrix} \dot{e}_1\\ \dot{e}_2\\ \dot{e}_3 \end{bmatrix} = \begin{bmatrix} -k_1 & 0 & 0\\ 0 & 0 & 0\\ 0 & -1 & -k_3 \end{bmatrix} \begin{bmatrix} e_1\\ e_2\\ e_3 \end{bmatrix} + \begin{bmatrix} 0\\ L_f \hat{h}_2 - h_{2ref}\\ 0 \end{bmatrix} ,$$
(31)
$$\dot{\overline{e}} = RL^{-1}\overline{e}$$
(32)

with

$$L_f \hat{h}_2 = \hat{h}_3 - \frac{2}{T_r} \hat{h}_2 - \dot{h}2 \text{ref}.$$
 (33)

Choosing

$$h_{3\rm ref} = \frac{2}{T_r} \hat{h}_2 + \dot{h}_{2\rm ref} - k_2 e_2 \tag{34}$$

and

$$L_f \hat{h}_2 = e_3 + \dot{h}_{22\text{ref}} - k_2 e_2 \tag{35}$$

we obtain the system of tracking errors

V

$$\begin{bmatrix} \dot{e}_1 \\ \dot{e}_2 \\ \dot{e}_3 \end{bmatrix} = \begin{bmatrix} -k_1 & 0 & 0 \\ 0 & -k_2 & 1 \\ 0 & -1 & -k_3 \end{bmatrix} \begin{bmatrix} e_1 \\ e_2 \\ e_3 \end{bmatrix}.$$
 (36)

The Lyapunov function V_2 with its derivative along the trajectories of the system ensures the asymptotic

$$V_2 = \frac{1}{2} \left(e_1^2 + e_2^2 + e_3^2 \right), \tag{37}$$

$$\dot{V}_2 = -k_1 e_1^2 - k_2 e_2^2 - k_3 e_3^2 \,. \tag{38}$$

We can conclude on the overall stability of the entire control and observer using Lyapunov function candidate V with its derivative

$$V = V_1 + V_2,$$
 (39)

$$\dot{V} = -k_1 e_1^2 - k_2 e_2^2 - k_3 e_3^2 - \overline{e}^\top (L^{-1}) \overline{e}$$
(40)

with V > 0 and $\dot{V} < 0$ we can deduce the stability of the observer and control using a complete model of the motor and a complete observer.

4 SIMULATION RESULTS

The proposed PI for controlling the speed of IM with AIOL decoupling and flux NLO was designed for 1.5kw drive is represented in Fig. 2. We interpret some simulation results with parameter variations of 50% on the rotor resistance (Rr) and the variation of torque (TL). We conduct a comparison of three cases in order to evaluate the performance of adaptive control with or without the use of a flux observer, these are:

1) Nonlinear control sets (without adaptation).

This control is developed based on the nominal model (R_{rN}, T_{LN}) applied to the model characterized by parameters R_r and T_L show a variation of 50 % R_r at t = 2 s to t = 5 s and a torque variation T_L 10N.m equal to t = (1.5s, 2.5s) and -10N.m at t = (5.5s, 6.5s).



Fig. 7. Rotor flux

2) Nonlinear adaptive control.

3) Nonlinear control with adaptive nonlinear observer.

This control law is developed similarly to the previous one using a robust nonlinear observer flux and current. We have chosen to adapt the following gains: $\Gamma = (1000, 1000)^{\top}$ see equation (38), with the weighting Q = (0.005; 1; 0.001) see equation (35). The results found are as follows

- According to the Figs. 3 and 4, we noted that despite the application of the load torque and the rotor resistance variations, speed and flux return to their initial values along very short time. We confirm that the speed and flux perfectly follow their references.



- In the comparison between the fixed nonlinear control and the adaptive nonlinear control without observer (case 1 and case 2), we base ourselves on the Figs. 5 to 7 which show the speed, torques and flux. We notice that the presence of the parametric variations involves a coupling between flux and torque for the fixed nonlinear control.

- Knowing that the errors of continuation speed are due to the presence of a variable load torque of 10 N.m to t = (1.5 to 2.5)s and -10 N.m to t = (5.5 to 6.5)s, we note that only the adaptive nonlinear control allows a good continuation of flux and that it ensures, with effectiveness, the limitation of current.

- Our second comparison between the adaptive nonlinear control and the adaptive nonlinear control with observer (case 2 and case 3), shows the interest to adapt the model used by the observer. The use of an observer allows primarily a better continuation speed, flux and torque (see Figs. 8–11), then giving a good limitation of the current.

5 CONCLUSION

In this article we developed an adaptive nonlinear control which takes account of the parametric variations of rotor resistance and the load torque. The latter makes it possible to ensure an effective limitation of current.

Theoretically we established a proof of stability of the adaptive nonlinear control in the case or flux is measured.





Fig. 10. Stator current

Table 1.

| Designation | Parameter | value |
|----------------------|-----------|---------------------------|
| Rotor resistance | R | 3.81Ω |
| Stator resistance | R | 4.85Ω |
| Mutual inductance | M | 0.258H |
| Stator inductance | L | $0.274 \mathrm{H}$ |
| Rotor inductance | L | $0.274 \mathrm{H}$ |
| Rotor inertia | J | $0.031 \mathrm{Kgm^2}$ |
| Pole pair | p | 2 |
| Viscous frict. Coef. | f | $0.0114 \mathrm{Nm/rd/s}$ |
| Mechanical power | P | $1.5 \mathrm{kW}$ |
| Nominal voltage | V | 220V |
| Nominal current | Ι | 3A |
| Nominal speed | n | $1450 \mathrm{tr/mn}$ |

However in reality we must use an observer because flux is not accessible to measurement. The studies of simulation made it possible to show a better performance when an observer of flux is used.

APPENDIX

Induction motor parameters

The induction motor used in this system is a threephase, Y-connected, four poles, $1.5 \,\mathrm{Kw}$, $50 \,\mathrm{Hz}$, $220 \,\mathrm{V}/3\mathrm{A}$



Fig. 11. Torques

type. The nominal values of the motor used in this simulation are given in the Table 1.

Lie derivation estimation

$$\begin{split} L_f \hat{h}_1 &= -p \frac{M}{L_r} \Big[p x_5 (\hat{x}_3 \hat{x}_1 + \hat{x}_4 \hat{x}_2) \\ &\quad + \Big(\lambda + \frac{1}{T_r} \Big) (\hat{x}_3 \hat{x}_2 + \hat{x}_4 \hat{x}_1) + p x_5 K (\hat{x}_3^2 + \hat{x}_4^2) \Big], \\ L_f \hat{h}_2 &= \frac{2}{T_r} \Big[M (\hat{x}_3 \hat{x}_1 + \hat{x}_4 \hat{x}_2) - (x_3^2 + x_4^2) \Big], \\ L_f \hat{h}_3 &= \frac{2M^2}{T_r^2} (\hat{x}_1^2 + \hat{x}_2^2) - 2M \Big(\frac{\lambda}{T_r} + \frac{1}{T_r^2} \Big) (\hat{x}_3 \hat{x}_1 + \hat{x}_4 \hat{x}_2) \\ &\quad + \frac{2M p x_5}{T_r} (\hat{x}_3 \hat{x}_2 - \hat{x}_4 \hat{x}_1) + \frac{2KM}{T_r^2} (\hat{x}_3^2 + \hat{x}_4^2), \\ L_g \hat{h}_1 &= [-p K \hat{x}_4, -p K \hat{x}_3], \\ L_g \hat{h}_3 &= [2R_r K \hat{x}_3, 2R_r K \hat{x}_4]. \end{split}$$

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