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Sensorless second order sliding mode observer for induction motor

S. SOLVAR, V. LE, M. GHANES, J.P. BARBOT and G. SANTOMENNA

Abstract— In this paper, a second order sliding mode observer for the induction motor without mechanical sensor is presented. This observer converges in finite time and is robust to the variation of parameters. Using Matlab/Simulink, the simulation results show the performance of the proposed observer. Furthermore, an industrial application is presented in order to highlight the technological interest of the proposed method and also show the difficulties due to real time computation constraints.

Index Terms—Induction Motor, Sensorless, Observer, Sliding mode, Super Twisting

I. INTRODUCTION

Nowadays Induction Motors (IMs) have replaced DC motors in the industrial application. There are many different ways to drive an induction motor. The differences between them are the performance of motor and the cost of implementation. The voltage over frequency method is the simplest method. This is a basic scalar control which imposes a constant relation between stator voltage and frequency. It is normally used without speed sensor. However, this control has not a good accuracy in speed and torque response because the flux and the torque are not controlled directly [8]. Contrarily, in the vector control which is based on relations valid for dynamic states, magnitude and instantaneous position of voltage, current and flux are controlled. This control allows to have a good motor's performance with respect to permanent state and dynamic state behavior. But, in the vector control the mechanical speed and the flux are needed. For this reason, in the recent years high performance sensorless induction motor drivers have attracted a great attention in industrial applications due to theirs advantages such as low cost, low maintenance, high reliability etc. The sensorless control helps to reduce the cost and resolves installing problem in many applications: lack of place, dangerous environment. Present effort are devoted to improve the performance of observer at low speed [16] and to develop a robust observer against perturbations and variation of parameters.

Several methods have been proposed to estimate speed and flux of IM such as: Luenberger observer [1] and Kalman filter [14], high gain and adaptive observers [6], [16] neural networks and signal injection [7] and sliding mode observer [15] etc. Compared with other observers, sliding mode

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technics [5] has attractive advantages of robustness to disturbances and insensitivity to variation of parameters when the sliding mode happens. However, the chattering behavior, that is inherent in standard sliding mode techniques, is often an obstacle for practical application if neglected. Higher-Order Sliding Mode [3] is one of the solutions which does not compromise robustness and avoid filtering estimation considered by other methods. In this paper, a new sensorless second sliding mode observer is proposed to estimates flux and speed of IM under unknown load torque. Firstly the observer is designed in continuous time. Thereafter, the work consists to show how to get a sliding mode observer under sampling in order to implement it on an experimental set-up for an industrial application. Here the sliding mode observer is over sampled in order to by pass the difficulties due to slow data acquisition. This paper is organized as follows: the section II recalls the IM model. In section III the observability of the IM is presented. After that the section IV proposes a second order sliding mode observer which is applied for IM. In sections V and VI, the simulation and experimentation results are shown to confirm the good performance of the second order sliding mode observer. Some conclusions and remarks are drawn in section VII.

II. IM PER-UNIT MODEL

In order to construct the proposed observer for an industrial application, the *IM* model given by [9] in the fixed (α, β) frame is rewritten under the following Per-Unit (p.u.) equations:

$$\begin{cases} \dot{x}_{1} = \gamma \cdot x_{1} + \theta \cdot (b \cdot x_{3} + c \cdot x_{5} x_{4}) + \xi \cdot v_{1} \\ \dot{x}_{2} = \gamma \cdot x_{2} + \theta \cdot (b \cdot x_{4} - c \cdot x_{5} x_{3}) + \xi \cdot v_{2} \\ \dot{x}_{3} = a \cdot x_{1} - b \cdot x_{3} - c \cdot x_{5} \cdot x_{4} \\ \dot{x}_{4} = a \cdot x_{2} - b \cdot x_{4} + c \cdot x_{5} \cdot x_{3} \\ \dot{x}_{5} = h \cdot (x_{3} \cdot x_{2} - x_{4} \cdot x_{1}) - d \cdot x_{5} - e \cdot T_{l} \end{cases}$$
(1)

with

with:

$$\begin{aligned} x_1 &= \frac{i_{s\alpha}}{I_{ref}}, \ x_2 &= \frac{i_{s\beta}}{I_{ref}}, \ x_3 &= \frac{\omega_{sref} \cdot \phi_{r\alpha}}{V_{ref}}, \\ x_4 &= \frac{\omega_{sref} \cdot \phi_{r\beta}}{V_{ref}}, \ x_5 &= \frac{p \cdot \Omega}{\omega_{sref}}, \ \sigma &= 1 - \frac{M_{sr}^2}{L_s \cdot L_r}, \\ \gamma &= \frac{R_s \cdot L_r^2 + R_r \cdot M_{sr}^2}{\sigma \cdot L_s \cdot L_r^2}, \\ T_r &= \frac{M_s r \cdot I_{ref} \cdot \omega_{ref}}{T_r \cdot V_{ref}}, \ b &= \frac{1}{T_r}, \ c &= \omega_{ref}, \\ d &= \frac{f_v}{J}, \ e &= \frac{p}{J \cdot \omega_{ref}}, \ h &= \frac{p^2 \cdot M_{sr} \cdot I_{ref} \cdot V_{ref}}{J \cdot \omega_{ref}^2 \cdot L_r}, \\ \theta &= \frac{K \cdot V_{ref}}{I_{ref} \cdot \omega_{ref}}, \ \xi &= \frac{V_{ref}}{\sigma \cdot L_s \cdot I_{ref}} \end{aligned}$$
and

 $i_{s\alpha}$, $i_{s\beta}$, $\phi_{r\alpha}$, $\phi_{r\beta}$, Ω are respectively the stator currents, the rotor fluxes and the speed. I_{ref} , V_{ref} , are the maximal values of stator current and voltage. ω_{sref} is the nominal stator pulsation. R_s and R_r are the stator and rotor resistance. L_s and L_r are the stator and rotor inductance. M_{sr} is mutual

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inductance. p is the number pairs of pole. J is motor inertia and f_v the viscous friction coefficient. T_l is the load torque. Notice that for the sake of homogeneity, hereafter the simulation and experimental results will be given in per-unit (p.u.).

III. BRIEF RECALL OF IM OBSERVABILTY

It is shown in ([10],[16]) that the IM observability cannot be established in the particular case when the fluxes $\phi_{r\alpha}$, $\phi_{r\beta}$ and the speed Ω are constant even if we use the higher derivatives of currents. This is a sufficient and necessary condition for lost of observability. This operating case match to the following physically interpretation :

1) when the fluxes are constant ($\phi_{r\alpha} = \phi_{r\beta} = 0$), or equivalently, the stator pulsation voltage is zero ($\omega_s = 0$), it implies that: $p\Omega + \frac{R_r T_e}{p\phi_d^2} = \omega_s = 0$ or: $T_{em} = -K\Omega$ where T_{em} is the electromagnetic torque, $K = \frac{p^2 \phi_d^2}{R_r}$ and $\phi_d^2 = \sqrt{\phi_{r\alpha}^2 + \phi_{r\beta}^2}$.

2) if the speed motor is constant; thus: $T_{em} = (f_v \Omega + T_l) = -K\Omega$. This last equation defines the **unobservability** curve in the map (T_l, Ω) with $M = \frac{p^2 \phi_d^2}{R_r} + f_v$ (Figure 1).



Fig. 1. Unobservability curve in the map (T_l, Ω) .

Obviously, the observability is lost gradually when we approach this unobservability curve.

IV. SECOND ORDER SLIDING MODE OBSERVER DESIGN

A. Recall on sliding mode observer

The sliding mode technique is used at first in control theory, but in the recent years sliding mode is presented such as a good observer for many applications [2], [11], [13]. Considering a system:

$$\begin{cases}
\dot{x}_{1} = x_{2} \\
\dot{x}_{2} = x_{3} \\
\vdots \\
\dot{x}_{n} = f(x_{1}, \cdots, x_{2}) \\
y = x_{1}
\end{cases}$$
(2)

To this system, a second order sliding mode observer is designed in the following way:

$$\begin{aligned} \dot{\hat{x}}_{1} &= \tilde{x}_{2} + \lambda_{1} \cdot |x_{1} - \hat{x}_{1}|^{\frac{1}{2}} \cdot \operatorname{sign}(x_{1} - \hat{x}_{1}) \\ \dot{\hat{x}}_{2} &= \alpha_{1} \cdot \operatorname{sign}(x_{1} - \hat{x}_{1}) \\ \dot{\hat{x}}_{2} &= E_{1} \cdot \left[\tilde{x}_{3} + \lambda_{2} \cdot |\tilde{x}_{2} - \hat{x}_{2}|^{\frac{1}{2}} \cdot \operatorname{sign}(\tilde{x}_{2} - \hat{x}_{2}) \right] \\ \vdots \\ \dot{\hat{x}}_{n} &= E_{n-2} \cdot \alpha_{n-1} \cdot \operatorname{sign}(\tilde{x}_{n-1} - \hat{x}_{n-1}) \\ \dot{\hat{x}}_{n} &= E_{n-1} \cdot \left[\tilde{\theta} + \lambda_{n} \cdot |\tilde{x}_{n} - \hat{x}_{n}|^{\frac{1}{2}} \cdot \operatorname{sign}(\tilde{x}_{n} - \hat{x}_{n}) \right] \\ \dot{\hat{\theta}} &= E_{n-1} \cdot \alpha_{n} \cdot \operatorname{sign}(\tilde{x}_{n} - \hat{x}_{n}) \\ \text{ with } E_{i} = \begin{cases} 1 \text{ if } \tilde{x}_{i} - \hat{x}_{i} = 0 \\ 0 \text{ if not} \end{cases} \end{aligned}$$

$$\end{aligned}$$

B. Application for IM

1) Continuous observer: Firstly, considering a change of variable like that:

$$\begin{cases} z_1 = x_1 \\ z_2 = x_2 \\ z_3 = b \cdot x_3 + c \cdot x_5 x_4 \\ z_4 = b \cdot x_4 - c \cdot x_5 x_3 \\ z_5 = \dot{z}_3 \\ z_6 = \dot{z}_4 \end{cases}$$
(4)

From the IM model (1) and this change of variable (4), a dynamical system for (4) can be obtained as:

$$\begin{pmatrix}
\dot{z}_1 &= -\gamma \cdot z_1 + \theta \cdot z_3 + \xi \cdot v_1 \\
\dot{z}_2 &= -\gamma \cdot z_2 + \theta \cdot z_4 + \xi \cdot v_2 \\
\dot{z}_3 &= z_5 \\
\dot{z}_4 &= z_6 \\
\dot{z}_5 &= z_7 \\
\dot{z}_6 &= z_8
\end{cases}$$
(5)

Now by applying the second order sliding mode observer (3) recalled on the section (IV-A) for system (5) we get an observer in the following way:

$$\begin{aligned} \hat{z}_{1} &= \theta \cdot \tilde{z}_{3} - \gamma \cdot z_{1} + \xi \cdot v_{1} + \lambda_{1} \cdot |e_{1}|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{1}) \\ \hat{z}_{3} &= \alpha_{1} \cdot \operatorname{sign}(e_{1}) \\ \hat{z}_{2} &= \theta \cdot \tilde{z}_{4} - \gamma \cdot z_{2} + \xi \cdot v_{2} + \lambda_{2} \cdot |e_{2}|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{2}) \\ \hat{z}_{4} &= \alpha_{2} \cdot \operatorname{sign}(e_{2}) \\ \hat{z}_{3} &= E_{1} \cdot E_{2} \cdot (\tilde{z}_{5} + \lambda_{3} \cdot |e_{3}|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{3})) \\ \hat{z}_{5} &= E_{1} \cdot E_{2} \cdot \alpha_{3} \cdot \operatorname{sign}(e_{3}) \\ \hat{z}_{4} &= E_{1} \cdot E_{2} \cdot (\tilde{z}_{6} + \lambda_{4} \cdot |e_{4}|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{4})) \\ \hat{z}_{6} &= E_{1} \cdot E_{2} \cdot \alpha_{4} \cdot \operatorname{sign}(e_{4}) \\ \hat{z}_{5} &= E_{1} \cdot E_{2} \cdot E_{3} \cdot E_{4} \cdot (\tilde{z}_{7} + \lambda_{5} \cdot |e_{5}|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{5})) \\ \hat{z}_{7} &= E_{1} \cdot E_{2} \cdot E_{3} \cdot E_{4} \cdot (\tilde{z}_{8} + \lambda_{3} \cdot |e_{6}|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{6})) \\ \hat{z}_{8} &= E_{1} \cdot E_{2} \cdot E_{3} \cdot E_{4} \cdot \alpha_{6} \cdot \operatorname{sign}(e_{6}) \\ \hat{z}_{8} &= E_{1} \cdot E_{2} \cdot E_{3} \cdot E_{4} \cdot \alpha_{6} \cdot \operatorname{sign}(e_{6}) \\ \text{with } E_{i} \begin{cases} 1 \text{ if } e_{i} = \tilde{z}_{i} - \hat{z}_{i} = 0 \\ 0 \text{ if not} \end{cases} \end{aligned}$$

The functions E_i ensure that the next steps errors do not escape too far before one has the convergent of the last step error.

By choosing the gains α_i, λ_i which respect to the reachability

condition of the Super Twisting algorithm [2], [11], [12] for example: $\alpha_1 > z_{5max}$, $\lambda_1 > 4 \cdot z_{5max} \cdot \frac{\alpha_1 + z_{5max}}{\alpha_1 - z_{5max}}$, we get $e_1 = e_2 = \ldots = e_6 = 0$, ie $\hat{z}_1 = z_1, \hat{z}_2 = z_2...\hat{z}_6 = z_6$ after some finite time T.

To find out the estimated speed and flux, equations (4) are solved as follows.

One has $z_3 = b \cdot x_3 + c \cdot x_5 \cdot x_4$, i.e.,

$$x_{3} = \frac{1}{b} \cdot (z_{3} - c \cdot x_{5} \cdot x_{4})$$
(7)

Furthermore, one has $x_4 = \frac{1}{b} \cdot (z_4 + c \cdot x_5 \cdot x_3)$. If one replaces x_3 by equation (7) one obtains:

$$x_4 = \frac{z_4 + \frac{c}{b} \cdot z_3 \cdot x_5}{b + \frac{c^2 \cdot x_5^2}{L}}$$
(8)

In the similar way, we get:

$$x_3 = \frac{z_3 - \frac{c}{b} \cdot z_4 \cdot x_5}{b + \frac{c^2 \cdot x_5^2}{b}}$$
(9)

If one derives the equation (7) we obtain:

$$z_5 = b \cdot (-z_3 + a \cdot x_1) + c \cdot \dot{x}_5 \cdot x_4 + c \cdot x_5 \cdot (-z_4 + a \cdot x_2)$$
(10)

In the similar way, we have:

$$z_6 = b \cdot (-z_4 + a \cdot x_2) - c \cdot \dot{x}_5 \cdot x_3 - c \cdot x_5 \cdot (-z_3 + a \cdot x_1)$$
(11)

So one has 4 equations (8,9,10,11) with 4 unknowns x_3 , x_4 , x_5 and \dot{x}_5 . Solving these equations we can find out the estimated speed and rotor fluxes.

To estimate the rotor speed we can add a following hypothesis: the speed is constant in comparison with the variation of stator current. From this hypothesis and equations (4) one can estimate the speed and the rotor flux.

$$\hat{x}_5 = \frac{\hat{z}_5 + b \cdot \hat{z}_3 - b \cdot a \cdot \hat{z}_1}{-c \cdot z_4 + c \cdot a \cdot \hat{z}_2}$$
(12)

$$\hat{\phi}_{r\alpha} = \frac{\hat{z}_3 - \frac{c \cdot \hat{x}_5 \cdot \hat{z}_4}{b}}{b + \frac{c^2 \cdot \hat{x}_5^2}{b}}$$
(13)

$$\hat{\phi}_{r\beta} = \frac{\hat{z}_4 + \frac{c \cdot \hat{x}_5 \cdot \hat{z}_3}{b}}{b + \frac{c^2 \cdot \hat{x}_5^2}{b}} \tag{14}$$

The designed second order sliding mode has another advantage because one does not need to know the load torque to estimate the motor's speed.

2) Discrete Observer: For the industrial application in real time, the discrete observer is constructed. The explicit Euler's method is chosen to transform continuous observer to discrete observer. This is due to the simplicity of computation. Nevertheless in order to obtain the request precision this discretization is used n times, doing so the discretization error is in $nO((\frac{T_e}{n})^2)$ instead of $O((\frac{T_e}{n})^2)$ with T_e is the data acquisition period. This method is performed in section (16). Applying the explicit Euler's method for the second order sliding mode observer, the discrete observer is obtained:

$$\begin{split} \hat{z}_{1}(k) &= \hat{z}_{1}(k-1) + T_{e} \cdot (\theta \cdot \tilde{z}_{3}(k-1) - \gamma \cdot z_{1}(k-1)) \\ &+ \xi \cdot v_{1}(k-1) \\ &+ \lambda_{1} \cdot |e_{1}(k-1))|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{1}(k-1))) \\ \hat{z}_{3}(k) &= \tilde{z}_{3}(k-1) + T_{e} \cdot \alpha_{1} \cdot \operatorname{sign}(e_{1}(k-1))) \\ \hat{z}_{2}(k) &= \hat{z}_{2}(k-1) + T_{e} \cdot (\theta \cdot \tilde{z}_{4}(k-1) - \gamma \cdot z_{2}(k-1)) \\ &+ \xi \cdot v_{2}(k-1) \\ &+ \lambda_{2} \cdot |e_{2}(k-1)|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{2}(k-1))) \\ \hat{z}_{4}(k) &= \tilde{z}_{4}(k-1) + T_{e} \cdot \alpha_{2} \cdot \operatorname{sign}(e_{2}(k-1)) \\ &+ \lambda_{3} \cdot |e_{3}(k-1)| + T_{e} \cdot E_{1} \cdot E_{2} \cdot (\tilde{z}_{5}(k-1)) \\ &+ \lambda_{3} \cdot |e_{3}(k-1)|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{3}(k-1))) \\ \hat{z}_{5}(k) &= \tilde{z}_{5}(k-1) + T_{e} \cdot E_{1} \cdot E_{2} \cdot \alpha_{3} \cdot \operatorname{sign}(e_{3}(k-1)) \\ \hat{z}_{4}(k) &= \hat{z}_{4}(k-1) + T_{e} \cdot E_{1} \cdot E_{2} \cdot \alpha_{3} \cdot \operatorname{sign}(e_{3}(k-1)) \\ \hat{z}_{6}(k) &= \tilde{z}_{6}(k-1) + T_{e} \cdot E_{1} \cdot E_{2} \cdot \alpha_{4} \cdot \operatorname{sign}(e_{4}(k-1))) \\ \hat{z}_{5}(k) &= \hat{z}_{5}(k-1) + T_{e} \cdot E_{1} \cdot E_{2} \cdot E_{3} \cdot E_{4} \cdot (\tilde{z}_{7}(k-1)) \\ &+ \lambda_{5} \cdot |e_{5}(k-1)|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{5}(k-1))) \\ \hat{z}_{6}(k) &= \hat{z}_{6}(k-1) + T_{e} \cdot E_{1} \cdot E_{2} \cdot E_{3} \cdot E_{4} \cdot (\tilde{z}_{8}(k-1)) \\ &+ \lambda_{3} \cdot |e_{6}(k-1)|^{\frac{1}{2}} \cdot \operatorname{sign}(e_{6}(k-1))) \\ \hat{z}_{8}(k) &= \tilde{z}_{8}(k-1) \\ &+ T_{e} \cdot E_{1} \cdot E_{2} \cdot E_{3} \cdot E_{4} \cdot \alpha_{6} \cdot \operatorname{sign}(e_{6}(k-1)) \\ &\qquad (15) \\ \text{with } E_{i} \begin{cases} 1 \text{ if } e_{i} = \tilde{z}_{i} - \hat{z}_{i} = 0 \\ 0 \text{ if not} \end{cases} \end{split}$$

For a good accuracy of this discrete observer, one must use a sampling time small enough and it needs a fast DSP. In the industrial application at GS Maintenance, the clock frequency of DSP is only 150MHz, and it does not allow to work on a small sampling time. So in experimentation an over-sample technique is proposed in the following way: $T_{e-new} = \frac{T_e}{n}$ with T_e is the data acquisition time, T_{e-new} is the new computation period, n is the number of over-sample. Considering a differential equation $\dot{X} = f(x)$:

The explicit Euler's method with a sampling time T_e gives: $X(k) = X(k-1) + T_e \cdot f(X(k-1))$ with T_e is the data acquisition period and computation period too. With an over sampling time $T_{e-new} = \frac{T_e}{n}$, the explicit

with an over sampling time $I_{e-new} = \frac{-r}{n}$, the explicit Euler's gives:

$$\begin{array}{rcl} X(kT_e + \frac{T_e}{n}) & = & X(kT_e) + \frac{T_e}{n} \cdot f(X(kT_e)) + O(\frac{T_e}{n})^2 \\ X(kT_e + 2\frac{T_e}{n}) & = & X(kT_e) + \frac{T_e}{n} \cdot f(X(kT_e)) \\ & + & \frac{T_e}{n} \cdot f(X(kT_e + \frac{T_e}{n})) \\ & + & 2O(\frac{T_e}{n})^2 \end{array}$$

•••

$$X(kT_e + T_e) = X(kT_e) + \frac{T_e}{n} \cdot f(X(kT_e)) + \frac{T_e}{n} \cdot f(X(kT_e) + \frac{T_e}{n} \cdot f(X(kT_e) + \frac{T_e}{n})) + \frac{T_e}{n} \cdot f(X(kT_e + 2\frac{T_e}{n})) + \dots + \frac{T_e}{n} \cdot f(X(kT_e + (n-1)\frac{T_e}{n})) + nO(\frac{T_e}{n})^2$$
(16)

Hence it can be remarked that the error is $O(T_e)^2$ for a sample time explicit Euler's while for explicit Euler's with an over sampling time, the error is $nO(\frac{T_e}{n})^2$, which reduce

by n the discretisation error.

In order to reduce again the computation time, one used a table of square root pre-calculated.

V. SIMULATION RESULTS

To verify the effectiveness of the proposed observer, a simulation has been carried out using Matlab/Smulink. The used sampling period is $10^{-4}s$. The simulation results are divided in two part: in one hand the observer is tested under nominal condition (identified parameters), in the other hand the influence of parameters variation is considered (robustness tests). The figures (2,3) illustrate estimated speed and estimated rotor flux. This result present a good performance of the second order sliding mode observer. Error between real speed and estimated speed is very small (0.5%). An error occurs in the time of changed speed because in the simulation one does not respect the hypothesis constant speed, in experimentation this problem does not appear due to system inertia. The figures (4,5,6,7) prove the robustness of proposed observed with respect to the variation of parameters. In this simulation test, the variation of $\pm 50\%$ of R_s , $\pm 50\%$ of R_r , +20% of L_s , +20% of L_r is considered. The results simulation show that the observer is insensible to R_r , L_s , L_r variation. It exists only a small error (1%) in the case of $\pm 50\%$ of R_s .



Fig. 2. Estimated speed (blue) and real speed (red).



Fig. 3. Estimated rotor flux in the fixed frame.



Fig. 4. Estimated speed (blue) and real speed (red) when R_s varies +50%.



Fig. 5. Estimated speed (blue) and real speed (red) when R_s varies -50%.



Fig. 6. Estimated speed (blue) and real speed (red) when R_r varies +50%.



Fig. 7. Estimated speed (blue) and real speed (red) when R_r varies -50%.



Fig. 8. Estimated speed (blue) and real speed (red) when L_s varies +20%.



Fig. 9. Estimated speed (blue) and real speed (red) when L_r varies +20%.

VI. EXPERIMENTAL RESULTS

The proposed second order super twisting sliding mode observer for IM has been implemented at GS Maintenance. The system consists to a 1.1kW/380V/50Hz induction motor fed by an inverter. The observer is implemented in a single fixed-point TMS320F2812. A mechanical sensor is mounted on the motor's shaft to compare measured speed and estimated speed. At first, the observer has been tested at nominal speed. The figures (8,9) show the estimated speed and rotor flux. In comparison the estimated speed with the real speed which is measured by an encoder, one obtains a good performance of the observer. After that the performance of observer is tested at low speed (5Hz) (figure 10). This figure shows a good result of estimated speed. The figure 11 shows that the observer diverges under unobservable condition (very low speed)¹. To ensure the performance of speed observer, at low speed a speed estimator is proposed and tested². The figure 12 shows under unobservable conditions the observer diverges but the estimator give a satisfactory result. For sum up, our practical approach consists in a variable structure observer: under observability conditions a second order sliding mode observer is used and under unobservable

conditions an estimator is used.



Fig. 10. Estimated speed (red) and real speed (black) at nominal speed (p.u.).



Fig. 11. Estimated rotor fluxes in the fixed frame.

¹see section III for more details

²the estimator is simply an integration of normalized motor model



Fig. 12. Estimated speed (blue) and real speed (red) in the dynamic regime (p.u.).



Fig. 13. Speed observer (blue), speed estimator (red) and real speed (black) (p.u.).

VII. CONCLUSION

In this paper a second order sliding mode observer is designed in order to estimate the speed of *IM* without mechanical sensor. An industrial application using a usual DSP shows that it is possible to implement a second order sliding mode observer in a single fixed point DSP. The over-sample technique allows to have a very fast sampling with Euler's method. Furthermore, the speed is estimated without the knowledge of load torque. The simulation and experimentation results affirm a good performance of the proposed second order sliding mode observer such that: finite time convergence, robustness with respect to the variation of parameters. The bad behavior of the observer under unobservable conditions (at very low speed) is remedied by using an estimator of speed.

Annexe: Induction Motor data

Rated power	1.1kW	R_s	8.4Ω
Rated speed	1390 rpm	R_r	5.5Ω
Number of pole pairs	2	L_s	0.349H
Rated voltage	220V/380V	L_r	0.349H
Rated current	4.8A/2.75A	M_{sr}	0.3 H

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