

Detection of bearing faults in asynchronous motors using Luenberger speed observer

Baptiste Trajin, Jérémi Regnier, Jean Faucher

▶ To cite this version:

Baptiste Trajin, Jérémi Regnier, Jean Faucher. Detection of bearing faults in asynchronous motors using Luenberger speed observer. 34th Annual Conference of IEEE Industrial Electronics (IECON 2008), Nov 2008, Orlando, United States. pp.3073-3078, 10.1109/IECON.2008.4758451. hal-02299479

HAL Id: hal-02299479 https://hal.science/hal-02299479

Submitted on 27 Sep 2019

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.



OATAO is an open access repository that collects the work of Toulouse researchers and makes it freely available over the web where possible

This is an author's version published in: http://oatao.univ-toulouse.fr/21675

https://doi.org/10.1109/IECON.2008.4758451

To cite this version:

Trajin, Baptiste and Régnier, Jérémi and Faucher, Jean Detection of bearing faults in asynchronous motors using Luenberger speed observer. (2008) In: 34th Annual Conference of IEEE Industrial Electronics (IECON 2008), 10-13 November 2008 (Orlando, United States)

Any correspondence concerning this service should be sent to the repository administrator: <u>tech-oatao@listes-diff.inp-toulouse.fr</u>

Detection of Bearing Faults in Asynchronous Motors using Luenberger Speed Observer

Baptiste Trajin, *IEEE Student Member* Université de Toulouse; LAPLACE; CNRS, INPT, UPS; 2 rue Charles Camichel BP7122 31071 Toulouse Cedex 07, France Jeremi Regnier, *IEEE Member* Université de Toulouse; LAPLACE; CNRS, INPT, UPS; 2 rue Charles Camichel BP7122 31071 Toulouse Cedex 07, France Jean Faucher, *IEEE Member* Université de Toulouse; LAPLACE; CNRS, INPT, UPS; 2 rue Charles Camichel BP7122 31071 Toulouse Cedex 07, France

Email: baptiste.trajin@laplace.univ-tlse.fr Email: jeremi.regnier@laplace.univ-tlse.fr Email: jean.faucher@laplace.univ-tlse.fr

Abstract— This paper deals with the application of Luenberger speed observer for the detection of damaged rolling bearings in asynchronous machines. Vibration monitoring of mechanical characteristic frequencies related to the bearings is widely used to detect faulty operations. However, vibration measurement is expensive and can not always be performed. An alternative is to base the monitoring on the available electrical quantities e.g. the machine stator current which is often already measured for control and protection purposes. This work investigates another detection approach using estimated variables such as mechanical rotating speed. Considering that bearing faults induce load torque oscillations, a theoretical stator current model in case of load torque oscillations demonstrates the presence of phase modulations. Then, a theoretical estimated rotor flux and speed model in case of load torque oscillations demonstrates the presence of harmonics related to load torque oscillations. These components appear in the estimated speed spectrum and can be used for detection purposes. Measurements show that speed spectral energy is modified in specific frequency ranges corresponding to bearing faults. A fault detector is then proposed. The efficiency of the indicator is studied on long data records of measurements for different operating conditions. The investigated indicator proves that the detection of bearing faults is ensured by using estimated mechanical speed.

I. INTRODUCTION

Electrical drives using induction motors are widely used in industrial applications because of their low cost and high robustness. However, faulty operations could be induced by bearing faults [1, 2]. To improve the availability and reliability of the drive, a condition monitoring could be implemented to favor the predictive maintenance. Traditionally, motor condition is supervised using vibration analysis, but measuring such mechanical quantities to detect bearing faults is often expensive. To overcome this problem, available electrical quantities such as stator current could be used. A general review of monitoring and fault diagnosis schemes using stator current can be found in [3]. Concerning bearing fault detection, several studies demonstrate that specific signatures appear on stator current spectrum [4, 5]. Some papers concern the definition of an automatic indicator performing an extraction of relevant information from the current spectrum [6, 7].

Moreover, closed-loop Luenberger observers are used in several applications for sensorless control [8, 9]. Nevertheless, few papers deal with the application of speed observer for



Fig. 1. State-model of the asynchronous drive Σ

bearing fault condition monitoring. The present paper deals with this specific aspect. It has been demonstrated [6, 7, 10] that bearing faults induce load torque oscillations. As a consequence, mechanical speed oscillations due to load torque variations appear and thus can be detected with a speed observer.

In this paper, the detection of speed oscillations is performed by using Luenberger observer. Section II recalls the construction of a Luenberger observer for asynchronous drives. Section III proposes a model of estimated variables in case of load torque oscillations such as rotor flux and mechanical rotating speed. The model is based on the assumption that load torque oscillations induce phase modulation on stator currents [11]. This approach is validated with simulated signals and experimental measurements. In section IV, an automatic indicator based on the extraction of spectral energy of the estimated speed is proposed to detect bearing faults. The faulty bearing comes from the after-sales service of a motor manufacturer that declares it as faulty. Then, the fault type is representative of realistic faults. The proposed detector is validated with different mechanical speed operating conditions. Experimental results demonstrate the ability of the proposed detection scheme to distinguish healthy and faulty bearings in several conditions.

II. SPEED OBSERVER FOR ASYNCHRONOUS DRIVES

In order to design a speed observer for the asynchronous machine, a model of the drive has to be defined. A general form of asynchronous machine state model is then recalled in figure 1. The model depends on the allowed input measurements and the desired output variables. The matrix B is the input matrix, C the output matrix, A the dynamic matrix, U the input vector, X the state vector and Y the output vector.

A. Asynchronous drive model

Stator currents are often already measured for control purposes. Moreover, in case of bearing fault detection, stator current analysis [6, 7] demonstrates the ability of these quantities to ensure the detection of faulty operations. The supply voltage of the machine can be measured or calculated with the knowledge of the duty-cycle applied on the inverter. Consequently, the supply voltages are defined as the input variables. The estimated stator currents are compared to real measurements in the closed-loop observer and can be defined as output and state variables. Finally, to get the mechanical speed, the rotor flux has to be estimated [8]. Then, the rotor flux is defined as a state variable. As the model is a fourorder system, each state, input or output variable, has to be considered in a two-phase coordinate system. The Concordia transform (α,β) is used in the model. Then, the classical first harmonic equations lead to the state model defined in (1).

$$\dot{X} = \underbrace{\begin{pmatrix} a_1 & 0 & a_2 & -a_3\omega \\ 0 & a_1 & a_3\omega & a_2 \\ a_4 & 0 & a_5 & -\omega \\ 0 & a_4 & \omega & a_5 \end{pmatrix}}_{A} X + \underbrace{\begin{pmatrix} \frac{1}{\sigma L_S} & 0 \\ 0 & \frac{1}{\sigma L_S} \\ 0 & 0 \\ 0 & 0 \\ \end{pmatrix}}_{B} \underbrace{\begin{pmatrix} V_{S\alpha} \\ V_{S\beta} \end{pmatrix}}_{B} \\ \underbrace{\begin{pmatrix} I_{S\alpha} \\ I_{S\beta} \\ \varphi_{r\alpha} \\ \varphi_{r\beta} \end{pmatrix}}_{Y} = \underbrace{\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ C & \underbrace{\begin{pmatrix} I_{S\alpha} \\ I_{S\beta} \\ \varphi_{r\alpha} \\ \varphi_{r\beta} \end{pmatrix}}_{X}$$

with:

$$a_{1} = -\left(\frac{1}{\sigma T_{S}} + \frac{1-\sigma}{\sigma T_{r}}\right), \quad a_{2} = \frac{1-\sigma}{\sigma T_{r} M_{sr}},$$
$$a_{3} = -\frac{1-\sigma}{\sigma M_{sr}}, \quad a_{4} = \frac{M_{sr}}{T_{r}}, \quad a_{5} = -\frac{1}{T_{r}}$$
$$\omega = p\Omega_{mech}$$

where:

- T_s is the stator electrical time constant,
- T_r is the rotor electrical time constant,
- M_{sr} is the mutual stator-rotor inductance,
- σ is the leakage coefficient,
- L_S is the cyclic stator inductance,
- *p* is the number of pole pairs,
- Ω_{mech} is the mechanical rotating speed.

B. Speed observer

The mechanical rotating speed is estimated by the difference (2) between estimated stator electrical pulsation $\hat{\omega}_s$ (3) and estimated rotor electrical pulsation $\hat{\omega}_r$ (4). The notation \hat{x} means that the variables are estimated.

$$\hat{\Omega}_{mech} = \frac{1}{p} (\hat{\omega}_s - \hat{\omega}_r) \tag{2}$$



Fig. 2. State-model of the Luenberger observer

$$\hat{\omega}_s = \frac{\hat{\phi}_{r\alpha}\hat{\phi}_{r\beta} - \hat{\phi}_{r\beta}\hat{\phi}_{r\alpha}}{\hat{\phi}_{r\alpha}^2 + \hat{\phi}_{r\beta}^2} \tag{3}$$

$$\hat{\omega_r} = \frac{M_{sr}}{T_r} \frac{\hat{\phi}_{r\alpha} I_{S\beta} - \hat{\phi}_{r\beta} I_{S\alpha}}{\hat{\phi}_{r\alpha}^2 + \hat{\phi}_{r\beta}^2} \tag{4}$$

The closed-loop observer is designed according to the model of the asynchronous machine (see figure 1). Moreover, the dynamic of the observer is modified to ensure the convergence of the observer variables, comparing to measurements by using a gain matrix K as depicted in figure 2. The poles of the observer are accelerated and the influence of speed variations is reduced. Variables U and Y are measured variables.

III. ESTIMATED SPEED WITH LOAD TORQUE OSCILLATIONS

It has been demonstrated that load torque oscillations induce mechanical speed oscillations and stator current modulations [12], especially phase modulations (PM) [11]. Then, it has to be shown that the presence of PM on the stator current induces oscillations on the estimated speed.

A. Expression of the mechanical speed

(1)

Considering the Luenberger Observer in open-loop conditions, the gain matrix K is null. According to the dynamic matrix A, the rotor flux is dependent on the electrical pulsation i.e. the mechanical rotating speed. However, the rotor flux can be expressed only with the stator current and the supply voltage (5).

$$\hat{\phi}_{r\alpha} = \int \left(\left[\frac{M_{sr}}{T_r} + \frac{a_1}{a_3} \right] I_{S\alpha} + \frac{1}{\sigma L_s a_3} V_{S\alpha} \right) dt - \frac{1}{a_3} I_{S\alpha}$$

$$\hat{\phi}_{r\beta} = \int \left(\left[\frac{M_{sr}}{T_r} + \frac{a_1}{a_3} \right] I_{S\beta} + \frac{1}{\sigma L_s a_3} V_{S\beta} \right) dt - \frac{1}{a_3} I_{S\beta}$$
(5)

Considering that the inverter imposes a sinusoidal supply voltage (6), where $V_{S\alpha\beta}$ is the amplitude of the supply voltage in (α,β) coordinate system and f_s the supply frequency,

$$V_{S\alpha} = V_{S\alpha\beta} \cos(2\pi f_s t)$$

$$V_{S\beta} = -V_{S\alpha\beta} \sin(2\pi f_s t)$$
(6)

Then, the rotor flux frequency components are similar to the stator current harmonics. As a consequence, without load torque oscillations, the stator current is considered as sinusoidal. Consequently, the rotor flux is also a sinusoidal wave and the mechanical speed is constant.

In case of load torque oscillations, the stator current is phase modulated at the load torque oscillation frequency f_{osc} (7) where β_I is the current modulation index, $I_{S\alpha\beta}$ the amplitude of stator current in (α,β) coordinate system, ϕ the phase between stator current and supply voltage and θ_{osc} an arbitrary phase.

$$I_{S\alpha} = I_{S\alpha\beta} \cos(2\pi f_s t + \beta_I \sin(2\pi f_{osc} t + \theta_{osc}) + \phi)$$

$$I_{S\beta} = -I_{S\alpha\beta} \sin(2\pi f_s t + \beta_I \sin(2\pi f_{osc} t + \theta_{osc}) + \phi)$$
(7)

Using the flux equation (5) for the α component and assuming that $\theta_{osc} = 0$, the rotor flux can be generally expressed as (8).

$$\hat{\phi}_{r\alpha} = \frac{A_2}{\omega_s} \sin(2\pi f_s t) + A_3 D(t) + A_1 \int D(t) dt \qquad (8)$$

where:

- $A_1 = I_{S\alpha\beta} \left[\frac{M_{sr}}{T_r} + \frac{a_1}{a_3} \right],$ $A_2 = \frac{V_{S\alpha\beta}}{\sigma L_s a_3},$ $A_3 = -\frac{I_{S\alpha\beta}}{a_3},$ $\omega_s = 2\pi f_s,$

- $D(t) = \cos(2\pi f_s t + \beta_I \sin(2\pi f_{osc} t) + \phi),$

Using the development of phase modulated cosinus in Bessel functions [13], and the assumption that the phase modulated index β_I is quite small, the rotor flux expresses as (9) for $\omega_{osc} \neq \omega_s$.

$$\hat{\phi}_{r\alpha} = \frac{A_2}{\omega_s} \sin(2\pi f_s t) + A_3 D(t) + \frac{A_1 J_0(\beta_I)}{\omega_s} \sin(\omega_s t + \phi) + A_1 J_1(\beta_I) \left[\frac{\sin((\omega_s + \omega_{osc})t + \phi)}{\omega_s + \omega_{osc}} - \frac{\sin((\omega_s - \omega_{osc})t + \phi)}{\omega_s - \omega_{osc}} \right]$$
(9)

where:

- J_0 is the 0-order Bessel function,
- J_1 is the 1-order Bessel function,

• $\omega_{osc} = 2\pi f_{osc}$.

A similar expression can be established for the β component of the rotor flux (10) for $\omega_{osc} \neq \omega_s$.

$$\hat{\phi}_{r\beta} = \frac{A_2}{\omega_s} \cos(2\pi f_s t) + A_3 E(t) + \frac{A_1 J_0(\beta_I)}{\omega_s} \cos(\omega_s t + \phi) + A_1 J_1(\beta_I) \left[\frac{\cos((\omega_s + \omega_{osc})t + \phi)}{\omega_s + \omega_{osc}} - \frac{\cos((\omega_s - \omega_{osc})t + \phi)}{\omega_s - \omega_{osc}} \right]$$
(10)

where:

• $E(t) = -\sin(2\pi f_s t + \beta_I \sin(2\pi f_{osc} t)).$

To simplify the expressions (9) and (10), it can be noticed that $J_0(\beta_I) \simeq 1$ and $J_1(\beta_I) \simeq \frac{\beta_I}{2} \simeq 0$ if β_I is quite small. Then, the rotor flux can also be considered as phase modulated at the load torque oscillation frequency (11).

$$\hat{\phi}_{r\alpha} = \Phi_r \cos(2\pi f_s t + \beta_\phi \sin(2\pi f_{osc} t + \psi_{osc}) + \psi_{\phi_r})$$
$$\hat{\phi}_{r\beta} = -\Phi_r \sin(2\pi f_s t + \beta_\phi \sin(2\pi f_{osc} t + \psi_{osc}) + \psi_{\phi_r})$$
(11)



Fig. 3. Gain Bode diagram of simulated transfer functions between simulated, estimated mechanical speed oscillations and load torque oscillations

where:

- β_{ϕ} is the flux modulation index,
- ψ_{osc} and ψ_{ϕ_r} are arbitrary phases.

According to (11) and (3)-(4) and by neglecting square terms, the stator and rotor electrical pulsations are equal to a constant plus an oscillation at the load torque oscillation frequency. As a consequence, the estimated mechanical rotating speed is also a constant plus oscillations at multiples of the load torque oscillation frequency, whose amplitudes depend on modulation index β_I and β_{ϕ} (12).

$$\hat{\Omega}_{mech} = \Omega_0 + \sum_k \Omega_{osc}(k) \sin(2\pi k f_{osc} t)$$
(12)

B. Simulation results

To ensure that load torque oscillations induce specific harmonics on estimated mechanical speed, the asynchronous machine and the observer defined in section II are simulated using Matlab/Simulink. The mean values of simulated and estimated speed are equal, proving the validity of the observer model. Moreover, in case of load torque oscillations, the estimated speed includes harmonics of higher amplitude than the simulated speed. By varying the load torque oscillations frequency, a Bode diagram is established between simulated, estimated speed and load torque oscillations. Figure 3 shows the gain between the speed oscillations and torque oscillations amplitudes. A first resonance frequency appears on the two transfer functions around $f_{res1} = 20Hz$. A second resonance frequency appears on the estimated speed around $f_{res2} =$ 160Hz and is characteristic of the dynamic of the observer. The electromechanical resonance around 20Hz is useful for the detection of bearing faults using stator current spectral analysis [6, 7]. Moreover, the characteristic resonance of the observer can also be used in the detection scheme to favor the detection of high frequency load torque oscillations due to bearing faults.

C. Measurements results with load torque oscillations

After demonstrating the convergence of the observer on simulated results, the estimated speed is compared to speed measurements on a 5.5kW induction machine. A DC machine



Fig. 4. Measured and estimated speed spectrum with load torque oscillations



Fig. 5. Estimated speed spectrum in healthy and faulty conditions

is used as a load. The DC machine is connected to a resistor through a DC/DC converter which controls the DC motor armature current. The reference of current is composed of an oscillating component at the rotating frequency plus an offset in order to induce load torque oscillations around a mean torque value. The supply frequency of the asynchronous machine is set to 50Hz and the mean torque equals 18N.m. Measured and estimated speed spectra are shown in figure 4. It can be pointed out that the first fault harmonics are sensibly equal. Differences between high order harmonics and noise amplitude are of no interest because the aim is to compare the estimated speed in healthy and faulty conditions when speed oscillations appear. Figure 5 shows spectra of the estimated speed of the induction machine in healthy and load torque oscillation conditions. The two noise amplitudes are similar and a comparison between healthy and faulty operations is possible. The amplitude of the speed oscillation component increases due to the load torque oscillation.

IV. BEARING FAULT DETECTION USING SPEED OBSERVER

Many studies demonstrate that bearing faults induce load torque oscillations [6, 7, 10, 14] and thus lead to speed oscillations. Section III emphasized the ability of a Luenberger speed observer to detect speed oscillations. Consequently, the estimated speed can be used to detect bearing faults. In this section, a 6208-type bearing coming from the aftersales service of a motor manufacturer is mounted in the asynchronous machine. The bearing is declared to be faulty but the fault types or locations are unknown.

As shown in figure 3, the electromechanical system includes a resonance that amplifies the effects of load torque oscillations on mechanical speed and stator currents. As a consequence, the stator current spectral energy detector presented in [6, 7] allows to detect bearing faults preferentially if one of the characteristic bearing fault frequencies [15] equals the resonance frequency. The second resonance frequency, characteristic of the dynamic of the observer, can also be used to detect more efficiently bearing faults on estimated speed. Thus, the detection of bearing faults using speed observer has to be more efficient for several rotating speeds than the detection using stator current monitoring.

A. Bearing characteristic fault frequencies

As a matter of fact, frequencies that could appear in vibration spectrum with appearance of bearings faults are theoretically well known. Harmonics due to defects could appear as combinations of mechanical rotating frequency and characteristic frequencies expressed by (13) [2, 15].

$$f_{orf} = \frac{f_r}{2} N_b \left(1 - \frac{D_b \cos \theta}{D_p} \right)$$

$$f_{irf} = \frac{f_r}{2} N_b \left(1 + \frac{D_b \cos \theta}{D_p} \right)$$

$$f_c = \frac{f_r}{2} \left(1 - \frac{D_b \cos \theta}{D_p} \right)$$
(13)

where:

- f_{orf} outer race fault frequency;
- f_{irf} inner race fault frequency;
- f_c cage frequency;
- f_r mechanical rotating frequency;
- N_b number of balls;
- D_b ball diameter;
- D_p pitch diameter;
- θ contact angle.

The spectral vibration analysis performed on the test bench with faulty bearings shows the appearance of multiples of the characteristic fault frequency related to the inner or outer race. Obviously, combinations of several fault frequencies appear in the vibration spectrum. Strong modulations at f_c and f_r can be noticed around the multiples of outer or inner race fault frequency. Concerning a 6208-type bearing with a rotating frequency of 25Hz, assuming $\theta = 0$, $f_{orf} = 89.4Hz$, $f_{irf} = 136Hz$ and $f_c = 9.94Hz$.

B. Definition of frequency ranges affected by bearing faults

Characteristic fault frequencies depend on the mechanical rotating speed. The speed can be estimated by using the detection of rotor slot harmonics on the stator current spectrum [16]. As the mechanical speed is estimated by the Luenberger observer, a stator current analysis becomes unnecessary. These observations lead us to define an automatic mechanical detector based on estimated speed spectral energy extraction. The estimated speed is computed off-line during 80s. Then, the



Fig. 6. Cumulative sums of relative error in %



Fig. 7. Cumulative sums of relative error in %

fault detector is defined by extracting energies on frequency ranges related to the frequency components at f_{def} , where f_{def} is either the inner or the outer race theoretical fault frequency. Moreover, the frequency ranges are extended to include modulations linked to the mechanical speed and cage frequencies underlined by a vibration spectral analysis. The chosen frequency ranges are given in (14).

$$\begin{bmatrix} nf_{def} - f_c; nf_{def} + f_c \end{bmatrix} \\ \begin{bmatrix} nf_{def} - f_r - f_c; nf_{def} - f_r + f_c \end{bmatrix} \\ \begin{bmatrix} nf_{def} + f_r - f_c; nf_{def} + f_r + f_c \end{bmatrix}$$
(14)

where $n \in [1; 5]$.

The proposed indicator uses the relative energy error in the specified frequency ranges between a faulty case and a healthy reference of estimated mechanical speed spectrum. Then, the relative error of energy is estimated in 15 frequency ranges related to the outer race characteristic fault (Δ_{orf}) and 15 frequency ranges related to the inner race fault frequency (Δ_{irf}). A total relative error is computed by adding relative errors related to inner and outer race fault frequencies (15).

$$\Delta_{tot}(k) = \Delta_{orf}(k) + \Delta_{irf}(k) \text{ with } k \in [1; 15]$$
(15)

A cumulative sum is then applied on the extracted energy

differences from the frequency ranges related to the investigated bearing fault. Finally, the last value of the cumulative sum is defined as the detector value. It is understandable that the detector values have to be close to zero when a healthy case is compared to the healthy reference. In addition, the detector values have to be higher than zero when a faulty case is compared to a healthy reference.

C. Detection of bearing faults using the low frequency resonance point

The supply frequency f_s is tuned to ensure that one of the characteristic frequencies is close to the low frequency resonance point. Figures 6(a)-6(b) show the cumulative sums of relative error for two operating points. In figure 6(a), the inner race fault characteristic frequency equals the resonance frequency. The supply frequency f_s equals 6.7Hz. In figure 6(b), the outer race fault characteristic frequency equals the resonance frequency. The supply frequency equals 13.3Hz.

The detection of the faulty bearing is ensured for the two operating points especially for $f_s = 6.7Hz$. It can be noticed that the bad performances of the Luenberger observer at low speed [17] do not affect the detection efficiency. Table I indicates mean and standard deviation of detectors for healthy and faulty operations.

TABLE I MEAN AND STANDARD DEVIATION OF DETECTORS

| f_s | Healthy case detectors | Faulty case detectors |
|-------|----------------------------|----------------------------|
| (Hz) | (Mean; Standard deviation) | (Mean; Standard deviation) |
| 6.7 | (16.8; 55.2) | (1020; 116) |
| 13.3 | (4.9; 26.7) | (454.7; 163.8) |
| 25 | (-28.9; 99.1) | (378.3; 61.8) |
| 50 | (90; 171.7) | (1464; 176.7) |

D. Detection of bearing faults at nominal and half-nominal speed

Figures 7(a)-7(b) depict the cumulative sums for healthy and faulty operations when the supply frequency equals 25Hzand 50Hz respectively. As the supply frequency equals 50Hz, the inner race characteristic fault frequency is close to the high frequency resonance point on the estimated speed. The detection of the faulty bearing is ensured for the two operating points especially for $f_s = 50Hz$. However the standard deviation increases with the mechanical rotating speed especially in faulty conditions. Table I indicates mean and standard deviation of detectors for healthy and faulty operations. Due to the high frequency resonance point, the efficiency of the detector on the estimated speed is better than the detector efficiency on the stator current [6, 7].

V. CONCLUSION

In this paper, a new method for an automatic detection of bearing faults in induction motors using Luenberger speed observer has been presented. Some mechanical faults induce load torque and mechanical speed oscillations. Thus, the mechanical speed can be used to detect faulty operating conditions. As a consequence, a speed observer has been designed to perform speed harmonic detection for bearing fault diagnosis. A simplified approach of the observer demonstrates the ability of the estimated rotating speed to render the speed oscillations. The simulation of the close-loop Luenberger observer shows a different frequency behavior between simulated and observed mechanical speed with regard to the first fault harmonic. The knowledge of the specific resonance frequencies underlined by the Bode diagram is useful to analyse the detection scheme performances.

The detector is tested on real faulty bearings mounted in an asynchronous drive. According to the knowledge of mechanical and geometrical bearing properties, an automatic indicator has been proposed to extract the estimated speed spectral energy in frequency ranges related to bearing faults.

Experimental results show that the detection of bearing faults is ensured using a mechanical speed observer for large mechanical speed ranges.

Moreover, the dynamic of the observer, especially the high frequency resonance, can be used to improve the fault detection. To demonstrate the efficiency and reliability of the detector, and to be more realistic regarding real time implementation of the detector, studies on short data length will be performed in further work.

REFERENCES

- B. Raison, G. Rostaing, O. Butscher and C.-S. Maroni, *Investigations* of algorithms for bearing fault detection in induction drives, IEEE 28th Annual Conference of the Industrial Electronics Society, vol. 2, Nov. 2002, pp. 1696-1701.
- [2] J. R. Stack, T. G. Habetler and R. G. Harley, *Fault classification and fault signature production for rolling element bearings in electric machines*, IEEE Transactions on Industry Applications, vol. 40, no. 3, May-Jun. 2004, pp. 735-739.
- [3] S. Nandi and H. A. Toliyat, Condition monitoring and fault diagnosis of electrical machines - a review, IEEE Transactions on Energy Conversion, vol. 20, no. 4, Dec. 2005, pp. 719-729.
- [4] J. R. Stack, R. G. Harley and T. G. Habetler, An amplitude modulation detector for fault diagnosis in rolling element bearings, IEEE Transactions on Industry Electronics, vol. 51, no. 5, Oct. 2004, pp. 1097-1102.
- [5] R. R. Obaid, T. G. Habetler and J. R. Stack, *Stator current analysis for bearing damage detection in induction motors*, International Symposium on Diagnostics for Electric Machines, Power Electronics and Drives (SDEMPED '03), Aug. 2003, pp. 182-187.
- [6] B. Trajin, J. Regnier, J. Faucher, *Bearing Fault Indicator in Induction Machine Using Stator Current Spectral Analysis*, Power Electronics Machine and Drives Conference (PEMD '08), York, Apr. 2008, pp. 592-596.
- [7] B. Trajin, J. Regnier, J. Faucher, Indicator for Bearing Fault Detection in Asynchronous Motors using Stator Current Spectral Analysis, IEEE International Symposium on Industrial Electronics (ISIE '08), Cambridge, Jun-Jul. 2008, pp. 570-575.
- [8] M. Cuibus, V. Bostan, S. Ambrosii, C. Ilas and R. Magureanu, Luenberger, Kalman and Neural Network Observers for Sensorless Induction Motor Control, 3rd International Power Electronics and Motion Control Conference (IPEMC 2000), vol. 3, Aug. 2000, pp. 1256-1261.
- [9] V. Bostan, M. Cuibus, C. Ilas and R. Magureanu, *High Performance Sensorless Solutions for Induction Motor Control*, Power Electronics Specialist Conference (PESC '03), vol. 2, Jun. 2003, pp. 556-561.
- [10] M. Blodt, P. Granjon, B. Raison and G. Rostaing, *Models for bearing damage detection in induction motors using stator current monitoring*, IEEE International Symposium on Industrial Electronics (ISIE '04), vol. 1, May 2004, pp. 383-388.
- [11] M. Blodt, J. Faucher, B. Dagues and M. Chabert, *Mechanical load fault detection in induction motors by stator current time-frequency analysis*, IEEE International Conference on Electric Machines and Drives, May 2005, pp. 1881-1888.
- [12] R. R. Schoen and T. G. Habetler, *Effects of time-varying loads on rotor fault detection in induction machines*, IEEE Transactions on Industry Applications, vol. 31, no. 4, Jul-Aug. 1995, pp. 900-906.
- [13] M. Abramowitz, I. A. Stegun, Handbook of Mathematical functions with Formulas, Graphs and Mathematical Tables, Dover Publications, New-York, nineth ed., 1964.
- [14] H. R. Smith, E. Wiedenbrug and M. Lind, Rotating Element Bearing Diagnostics in a Nuclear Power Plant: Comparing Vibration and Torque Techniques, International Symposium on Diagnostics for Electric Machines, Power Electronics and Drives (SDEMPED '07), sep. 2007, pp. 17-22.
- [15] T. A. Harris, Rolling bearing analysis, Wiley, New-York, 3rd ed., 1991.
- [16] S. Nandi, S. Ahmed and H. A. Toliyat, Detection of Rotor Slot and Other Eccentricity Related Harmonics in a Three Phase Induction Motor with Different Rotor Cages, IEEE Transactions on Energy Conversion, vol. 16, no. 3, Sep. 2001, pp. 253-260.
- [17] R. Zhou and J. S. Lai, Low-Speed Performance Comparison of Induction Motor Sensorless Control Methods, 7th Workshop on Computers in Power Electronics (COMPEL 2000), Jul. 2000, pp. 247-252.