



Improvement of the induction motor sensorless control based on the type-2 fuzzy logic

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Abstract

This paper presents MRAS speed sensorless control of induction motor using type-2 fuzzy logic controller (T2FLC). These controllers replace the PI ones, in the new MRAS strategy proposed in Benlaloui et al. (IEEE Trans Energy Convers 30(2):588–595, 2015), in order to improve the induction motor performances and robustness at low speed region. Indeed, the choice of these controllers is made because of their adaptation-based schemes which permit to handle nonlinear uncertain systems without the need of precise mathematical model required when using PI controllers. Comparative study had shown a better rejection of disturbance and high insensitivity to stator resistance compared to the PI and the T1FLC controllers. The effectiveness of the proposed speed-based T2FLC estimation method and its good robustness are validated by simulation and by experimental results.

Keywords Induction machines (IM) · Model Reference Adaptive System (MRAS) · Field-oriented control · Lyapunov Theory · Type-2 Fuzzy Logic Controller (T2FLC)

List of abbreviations

MRAS Model Reference Adaptive System

T2FLC Type-2 Fuzzy Logic Controller

IM Induction machines

FOC Field-oriented control

PI Proportional–integral

IFOC Indirect field-oriented control

s, r Rotor and stator indices

d, q Direct and quadrature indices for orthogonal components

P Number of pairs poles

Ω Rotor speed (rd/s)

ω_r Induced rotor current frequency (rd/s)

J_{in} Inertia

Γ Unknown torque

*

\bar{x}^* Symbol indicating the command value

R_s, R_r Complex conjugate

L_s, L_r Stator and rotor resistances

T_s, T_r Stator and rotor inductances

σ Stator and rotor time constants ($T_{s,r} = L_{s,r}/R_{s,r}$)

M Leakage flux total coefficient ($\sigma = 1 - M^2/L_r L_s$)

ω Mutual inductance

ω_s Mechanical rotor frequency (rd/s)

f Stator current frequency (rd/s)

Γ_e Coefficient of viscous

Γ_e Electromagnetic torque

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1 Introduction

Speed sensorless induction motor (IM) applications are becoming more and more used in industry thanks to their mechanical robustness, higher reliability and to a reduction in bulk and cost [1, 2]. For several years, many speed estimators have been proposed for IM drive [1–10] and permanent magnet synchronous machine (PMSM) [11, 12]. In these last references a binary search algorithm-based phase-locked loop and a finite position set are used.

The obtained results are satisfactory, and the performances of estimators are good.

However, thanks to its simplicity, MRAS-based rotor flux [3, 13–15] remains the most applied strategy. The speed MRAS observer has been widely used in the vector control of induction motors due to its good performance and its simple structure [3, 13–15]. However, in some applications the MRAS observer may not meet the concerned robustness under parameter variations such as rotor and stator resistances and external load disturbances especially at low and zero speed regions. Therefore, many studies have been made in order to overcome this problem which remains one of the most interesting issues to be solved [16–21].

Among them, the new MRAS speed observer is proposed in [21]. Indeed, in this strategy, although PI controllers were used, the results were satisfactory. However, the use of PI controllers requires precise mathematical model, fine tuning of proportional (K_p) and integral (K_i) gain values to achieve good performance drive. However, the lack of knowledge about the actual values of some time varying parameters, such as rotor and stator resistances, limits the use of this kind of controllers. In order to overcome these drawbacks, fuzzy logic and neuro-fuzzy-based controllers were proposed [22–27]. In [22], a comparative study between a fuzzy speed controller and conventional PI controller confirms that the first one leads to advanced performance for all the operating range, such as step variation in speed. Equivalently, fuzzy speed controller in indirect field-oriented control (IFOC) technique is used in [23]. By replacing the conventional PIs in the vector control by fuzzy controllers, the trajectory tracking and the response time are improved. The use of a fuzzy regulator and a neuro-fuzzy in [24–27] for the speed control shows very satisfactory performance for all the operating range.

This paper aims to improve both the robustness and the stability of the MRAS observer especially for low speed regions by replacing conventional PIs with type-2 FLC. Indeed, the choice of these controllers is made because of their adaptation-based schemes which permit to handle nonlinear uncertain systems without the need of precise mathematical model required when using PI controllers. The type-2 fuzzy membership function takes into account uncertainties, while the type-1 fuzzy membership function does not [28].

This paper is organized as follows: In Sect. 2, the mathematical model and the vector control strategy of the induction motor are given. Then, an improved MRAS speed observer is shown in Sect. 3 and its stability is proven using Lyapunov theory. Section 4 details the principle of the proposed T2FLC algorithm. Finally, the results obtained by simulation and confirmed by experimental tests are presented in Sects. 5 and 6.

2 Vector control strategy

The voltage equations in the synchronous frame are given by:

$$\begin{cases} \bar{u}_s = R_s \bar{I}_s + \sigma L_s \frac{d\bar{I}_s}{dt} + \frac{M}{L_r} \frac{d\bar{\phi}_r}{dt} + j\sigma L_s \omega_s \bar{I}_s + j\frac{M}{L_r} \omega_s \bar{\phi}_r \\ 0 = \frac{1}{T_r} \bar{\phi}_r - \frac{M}{T_r} \bar{I}_s + \frac{d\bar{\phi}_r}{dt} + j\omega_r \bar{\phi}_r \end{cases} \quad (1)$$

The mechanical equation is expressed by:

$$\Gamma_e - \Gamma_l = J_{in} \frac{d\Omega}{dt} + f\Omega. \quad (2)$$

The equation of electromagnetic torque is as follows:

$$\Gamma_e = \frac{PM}{L_r} (I_{sq} \phi_{rd} - I_{sd} \phi_{rq}) \quad (3)$$

If we admit that the viscous coefficient and the load are not known, Eq. (2) can be written as:

$$\Gamma_e - \Gamma = J_{in} \frac{d\Omega}{dt}, \quad (4)$$

with $\Gamma = \Gamma_l + f\Omega$.

This strategy of control leads to ensure the decoupling between flux and torque [20, 21]. The flux component is oriented to coincide with the axis (d), so that the following results are obtained:

$$\begin{cases} \phi_{rq} = 0 \\ \phi_{rd} = \phi_r \end{cases} \quad (5)$$

Using (5), the relations between torque, flux and current can be expressed by:

$$\Gamma_e = k_c \phi_r I_{sq} \quad (6)$$

with $k_c = \frac{PM}{L_r}$ and I_{sq} plays the role of a control variable for the torque.

3 Improved MRAS observer

Figure 1 represents the basic configuration of MRAS approach based on adaptive model theory with reference model [20, 21].

The reference and adaptive model equations expressed in the fixed reference frame (α, β) are given, respectively, by:

$$\bar{\phi}_r = \frac{\sigma L_s L_r}{M} \bar{i}_s + \frac{L_r}{M} \int (\bar{u}_s - R_s \bar{i}_s) dt \quad (7)$$

$$\hat{\phi}_r = \int \left(\frac{M}{T_r} \bar{i}_s - \left(\frac{1}{T_r} - jP \cdot \hat{\Omega} \right) \hat{\phi}_r \right) dt. \quad (8)$$

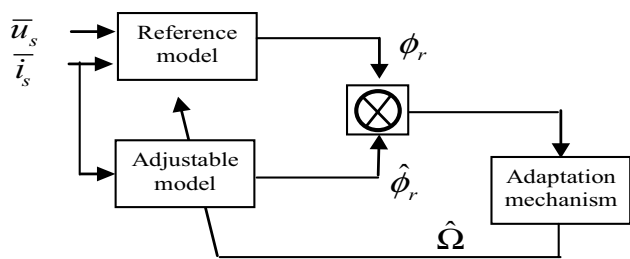


Fig. 1 Basic configuration of MRAS speed observer

The adaptation mechanism can be obtained by Lyapunov’s theorem, and it is easily to extract it from Popov’s criterion:

$$\int_0^t e_\varphi^T W d\tau = \int_0^t (P\Delta\Omega [e_{\varphi\alpha} \ e_{\varphi\beta}] J \hat{\varphi}_r) d\tau \geq -\delta_o^2 \tag{9}$$

A low pass filter is suggested in the adaptation law to give a good speed estimation response [14]:

$$\hat{\Omega} = k_p(\bar{\varphi}_r \otimes \hat{\varphi}_r) + k_i \int (\bar{\varphi}_r \otimes \hat{\varphi}_r) dt \tag{10}$$

where k_p and k_i are positives.

However, the major disadvantage of the MRAS observer is poor performance for low speeds and sensitivity to parametric variations. Several authors have proposed different approaches to maintain performances at satisfactory level for low speed; among these solutions in [21], the speed is estimated by the use of two differences between fluxes and torques.

As, the electromagnetic torque can be described by:

$$\Gamma_e = P \frac{M}{L_r} (\bar{i}_s \otimes \bar{\varphi}_r) \tag{11}$$

where $\bar{\varphi}_r$ is expressed by Eq. (7).

Therefore, the estimated electromagnetic torque can be described by:

$$\hat{\Gamma}_e = P \frac{M}{L_r} (\bar{i}_s \otimes \hat{\varphi}_r) \tag{12}$$

where $\hat{\varphi}_r$ is expressed by Eq. (8).

It can be easily demonstrated that according to the equation of motion (2) each variation of the load torque causes a variation of the speed and consequently the estimated variation of the speed is exactly equal to the estimated variation of the electromagnetic torque.

Then, we replace the electromagnetic torque and the speed by their estimated values in the motion Eq. (2); we obtain:

$$\hat{\Gamma}_e - \Gamma_l = J_{in} \frac{d\hat{\Omega}}{dt} + f\hat{\Omega}. \tag{13}$$

We obtain Eq. (14) by the subtraction of (13) from (2)

$$e_\Gamma = \Gamma_e - \hat{\Gamma}_e = J_{in} \frac{d(\Omega - \hat{\Omega})}{dt} + f(\Omega - \hat{\Omega}). \tag{14}$$

To ensure a good estimate, the following two conditions must be taken into account:

$$\begin{cases} \dot{\bar{e}}_\phi = -\left(\frac{1}{T_r} - jP\Omega\right)\bar{e}_\phi + jP(\Omega - \hat{\Omega})\hat{\phi}_r \\ e_\Gamma = \Gamma_e - \hat{\Gamma}_e = J_{in} \frac{d(\Omega - \hat{\Omega})}{dt} + f(\Omega - \hat{\Omega}). \end{cases} \tag{15}$$

By following the same reasoning with which the adaptation law has been determined and taking into account the torque adaptation error, the speed adaptation law becomes:

$$\hat{\Omega} = \left(K_p + \frac{K_i}{p}\right)(\bar{\phi}_r \otimes \hat{\phi}_r) + K_\Gamma \frac{e_\Gamma}{\tau p + 1}, \tag{16}$$

where τ is chosen to be close to the mechanical time constant.

In this approach, the electromagnetic torque error is filtered by a low pass filter and added to the conventional adaptation law loop. The diagram of the new MRAS estimator is shown in Fig. 2.

4 Design of type-2 fuzzy logic controller

In order to achieve a high dynamic performance of the new MRAS approach, we replace all the conventional PI of vector control regulator and classical MRAS adaptation mechanism of rotor speed estimation by the T2FLC algorithm, which provided satisfactory solutions in [19–22]. The inputs of the proposed algorithm for new MRAS adaptation mechanism are the rotor flux static and dynamic errors $\bar{e}_\phi = (\bar{\phi}_r - \hat{\phi}_r)$, which can be expressed as:

$$\Delta e_\phi(k) = e_\phi(k) - e_\phi(k - 1) \tag{17}$$

These two inputs are multiplied by two scaling factors G_{e_ϕ} and $G_{\Delta e_\phi}$, respectively. The output of the controller is multiplied by a third scaling factor G_u to generate the actual value of the rate of change of estimated rotor speed.

The three quantities, E_ϕ , ΔE_ϕ (inputs), and $\Delta\hat{\Omega}$ (output), are standardized as follows:

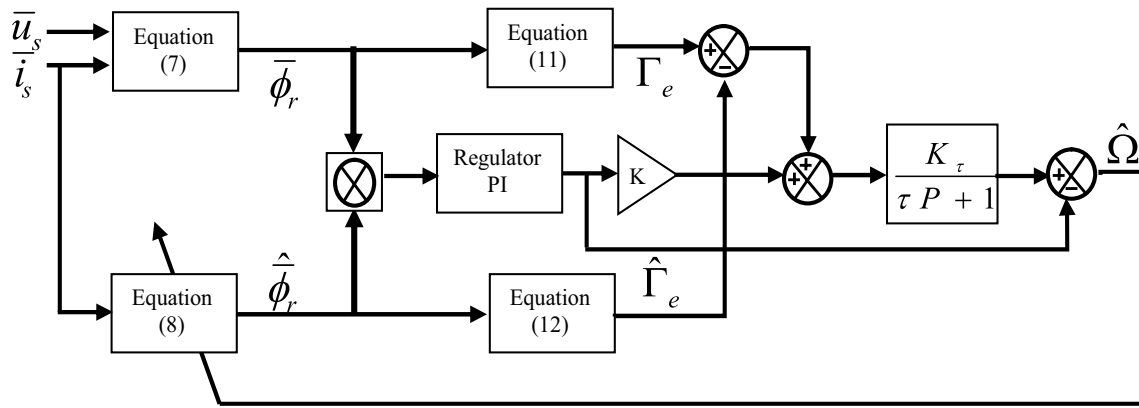


Fig. 2 The novel configuration design of MRAS

Table 1 Scale factors

Scale factor G_{i_j}	Estimated speed $\hat{\Omega}$	Torque Γ_e
$G_{e_{\phi,\Omega}}$	100	100
$G_{\Delta e_\phi, \Delta e_\Omega}$	1000	1000
G_u	0.9	0.9

$$\begin{cases} E_\phi = G_{e_\phi} e_\phi \\ \Delta E_\phi = G_{\Delta e_\phi} \Delta e_\phi \\ \Delta \hat{\Omega} = G_u \Delta \hat{\omega} \end{cases} \quad (18)$$

Therefore, the inputs of vector control regulator are the rotor speed and dynamic errors cited above which can be expressed as:

$$\Delta e_\Omega(k) = e_\Omega(k) - e_\Omega(k-1) \quad (19)$$

These two inputs are multiplied by two scaling factors G_{e_Ω} and $G_{\Delta e_\Omega}$, respectively. The output of the controller is multiplied by a third scaling factor G_u to generate the actual value of the electromagnetic torque Γ_e reference.

The three quantities, e_Ω , Δe_Ω (inputs), and $\Delta \Gamma_e$ (output), are standardized as follows:

$$\begin{cases} E_\Omega = G_{e_\Omega} e_\Omega \\ \Delta E_\Omega = G_{\Delta e_\Omega} \Delta e_\Omega \\ \Delta \Gamma'_e = G_u \Delta \Gamma_e \end{cases} \quad (20)$$

where G_{e_Ω} , $G_{\Delta e_\Omega}$ and G_u are scale factors or normalization, and play an important role on the control performance. The trial-and-error technique is usually used to tune these gains to ensure optimal performance of the controller [28].

The values of scale factors are given in Table 1.

Finally, a discrete integration is performed to get the value of the estimated speed and electromagnetic torque reference Γ_e . Hence, FLC is designed as shown in Figs. 3, 4, 5 and 6. The expressions of the estimated speed and electromagnetic torque reference can be written as:

$$\hat{\Omega}(k) = \hat{\Omega}(k-1) + G_u \Delta \hat{\Omega}(k) \quad (21)$$

$$\Gamma_e(k) = \Gamma_e(k-1) + G_u \Delta \Gamma_e(k) \quad (22)$$

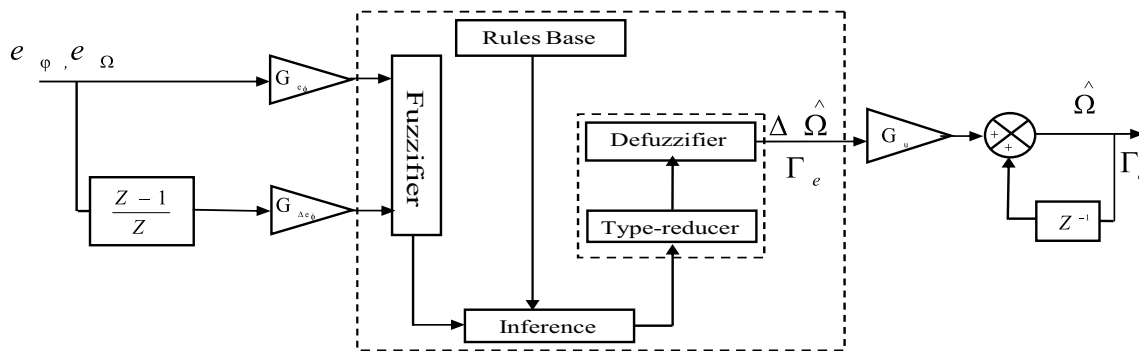


Fig. 3 Proposed type-2 fuzzy logic controllers

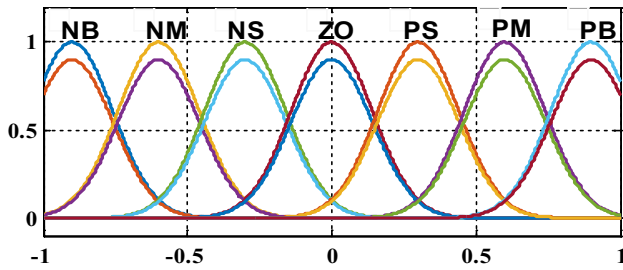


Fig. 4 Fuzzy type-2 membership functions for error variation flux and rotor speed

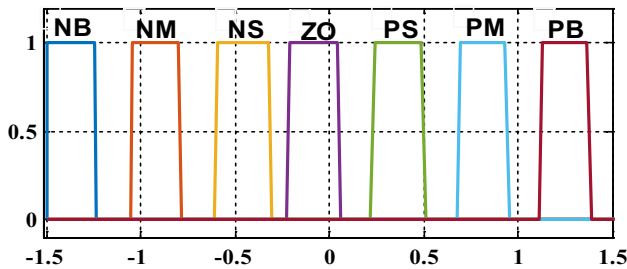


Fig. 5 Fuzzy type-2 membership functions for estimated speed

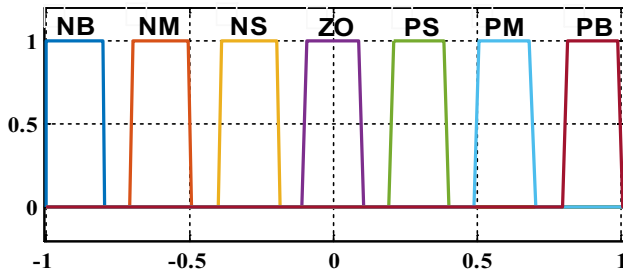


Fig. 6 Fuzzy type-2 membership for reference electromagnetic torque

The type-2 membership functions of errors (e_ϕ, e_Ω) and variation errors ($\Delta e_\phi, \Delta e_\Omega$) of flux and rotor speed are chosen identical with Gaussian forms and are defined on the interval $[-1, 1]$ (Fig. 4). The type-2 belonging functions of the variation Γ_e and $\hat{\Omega}$ are chosen with intervals form on the interval $[-1.5, 1.5]$ (Fig. 5), and the variation $\Delta\hat{\Omega}$ is chosen with intervals form on the interval (Fig. 6).

In Figs. 7 and 8, we present the block diagrams of the new MRAS observer and the general scheme of the proposed control.

5 Simulation and experimental results

In what follows, the simulation and experimental results best embody the performance of the proposed observer structure. First, the performances of the proposed approach with PI and T2FLC are analyzed and compared by simulation. Second, dSPACE DS 1104 was used to validate the simulation results through extensive experimentations.

The values of all controllers' gains, and the rating and parameters of the induction motor are listed in Table 1 and in "Appendix," respectively.

5.1 Simulation results

In order to demonstrate the efficiency of the control structure, a sensorless field-oriented controlled induction motor drive, as shown in Fig. 8 is used. Full nonlinear simulations were carried out for the speed and for load torque variation (Figs. 9, 10, 11, 12 and 13).

Figures 10 and 11 represent responses of the actual rotor speed and the estimated one following the specified reference. We can see in Figs. 11, 12 and 13 a fast convergence toward zero of speed estimation error and that of

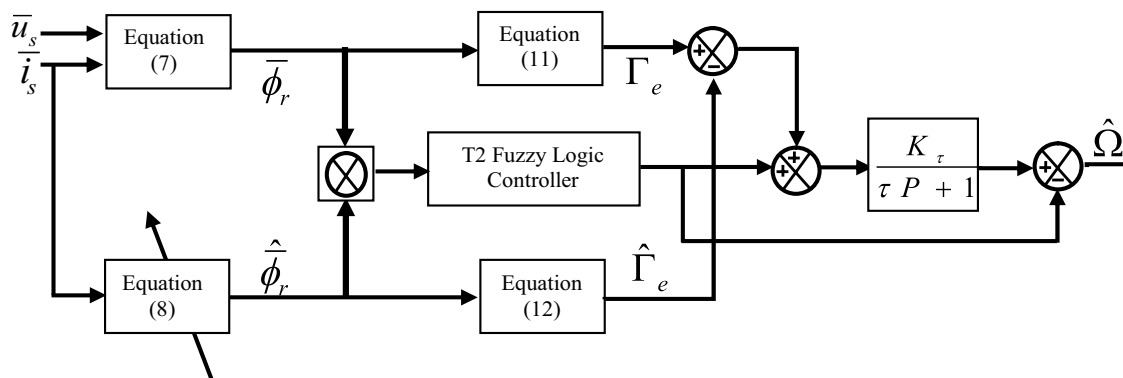


Fig. 7 Block diagram of MRAS speed observer with T2FLC

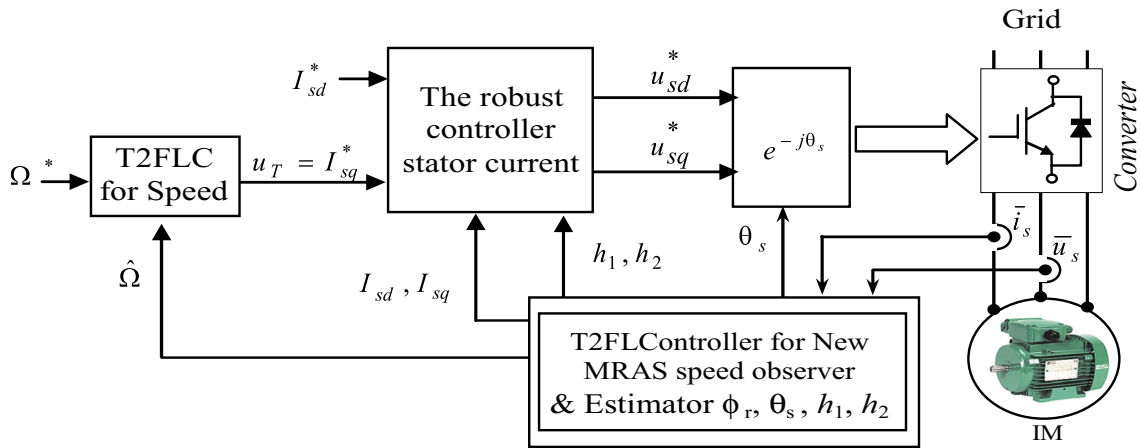


Fig. 8 General scheme of the proposed control

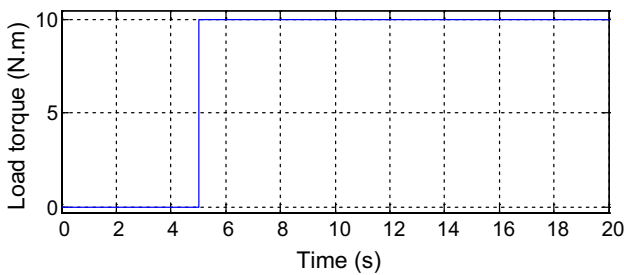


Fig. 9 Load torque variation

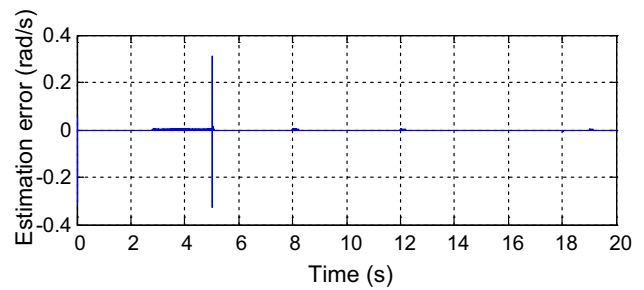


Fig. 12 Speed estimation error

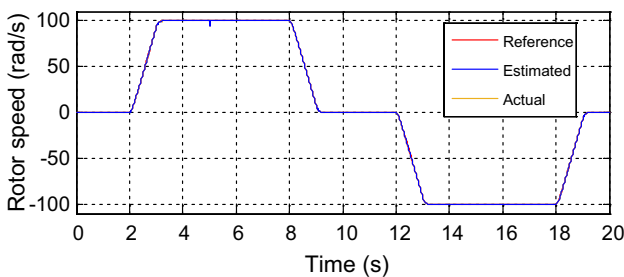


Fig. 10 Speed of induction motor

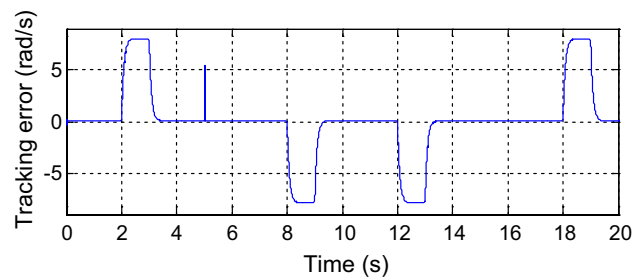


Fig. 13 Speed tracking error

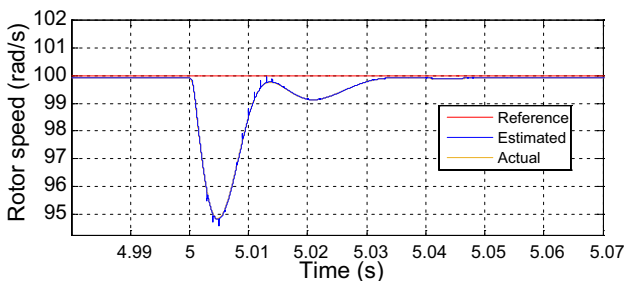


Fig. 11 Zoom of reference, actual and estimated rotor speed

the tracking; also, estimated speed is very satisfactory at zero regions. A robustness test for a load torque variation at a constant speed reference shows the efficiency of the type-2 FLC, where a step load variation of 10 Nm (Fig. 9) is applied at $t=5$ s. In addition, in order to confirm the field orientation, d and q axis fluxes are separately shown in Fig. 14. We can see that the q axis flux is maintained at zero value. In Fig. 15, we can see the influence of the load variation on the I_{sq} current at $t=5$ s and on the other hand no effect on the direct current I_{sd} of the machine. Figures 16 and 17 show that the current peaks remain within acceptable limits.

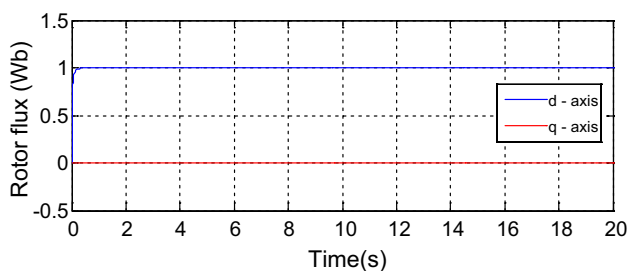


Fig. 14 Rotor flux

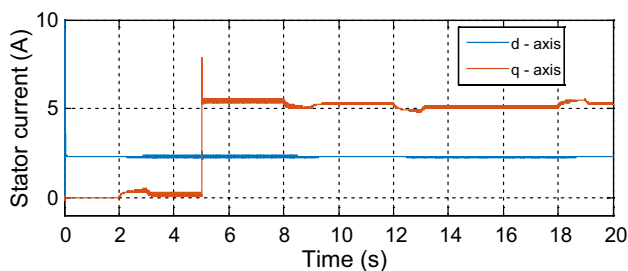


Fig. 15 I_{sd} and I_{sq} stator currents

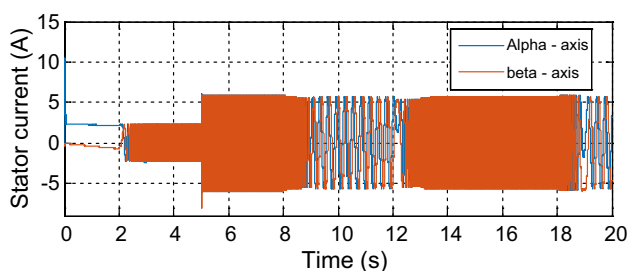


Fig. 16 $I_{s\alpha}$ and $I_{s\beta}$ stator currents

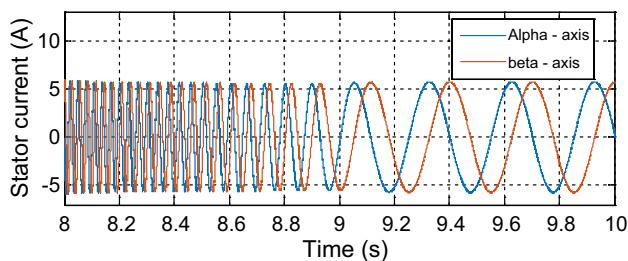


Fig. 17 Zoom of $I_{s\alpha}$ and $I_{s\beta}$ stator currents

5.2 Description of the laboratory setup

Experimental tests are conducted by using dSPACE DS1104 to implement the novel proposed analysis and design approach of rotor MRAS speed observer. Figure 18

shows the basic structure of the laboratory setup with all its parts. The IM stator is fed by a converter under direct control of the DS1104 board. In order to measure the mechanical speed, the encoder is used. The sensors used for the currents and voltages measure are LA-55NP and LV-25P, respectively. The DS1104 PPC controller connected signals are provided galvanic isolation with the Interface.

5.3 Experimental results

Figure 20 shows estimated, actual and reference speeds. As we can see in Figs. 19 and 20 that despite of the load torque at $t=5$ s speed estimation and tracking errors are small and converge swiftly to zero. We also notice, as can be seen in Fig. 21, that the current i_{sd} is kept constant for all the operating range. Figures 22, 23, 24 and 25 show that the stator voltage and current phase are maintained at their rate value. This proves the efficiency proposed control.

In order to confirm the performances of the new control structure at low speed regions respect to load torque and stator resistance R_s variations, a load torque of 10 Nm and $+20\%R_s$ variations are applied to the motor at $t=3.5$ s and $t=6.6$, respectively; it is clearly shown in Figs. 26 and 27 that the new structure with T2FLC controller is capable to provide a quick response and a better rejection of disturbance. Then, it confirms the insensitivity to stator resistance of the speed response.

6 Comparative study between PI, T1FL and T2FL controllers

In order to confirm the performances of the new control structure using T2FLC, a comparison between simulation results obtained at low speed regions by T2FLC, classical PI and T1FLC controllers was carried out with respect to load torque and stator resistance variations. A load torque of 10 Nm and $+30\%R_s$ variations are applied to the motor at $t=5$ s and $t=7$ s, respectively. It is clearly shown in Figs. 28 and 29 that the new structure with T2FLC controller is able to provide a quick response, a better rejection of disturbance and high insensitivity to stator resistance compared to the PI and the T1FLC controllers.

Also comparative study between T2FLC and T1FLC was made by using the performance indexes, integral squared error (ISE), integral absolute error (IAE) and the integral of time multiplied absolute error (ITAE) as given in [25].

The results of the study are given in Table 2.

As can be seen, these results confirm the improved performance of the type-2 fuzzy logic controller.

Fig. 18 Structure of the laboratory setup

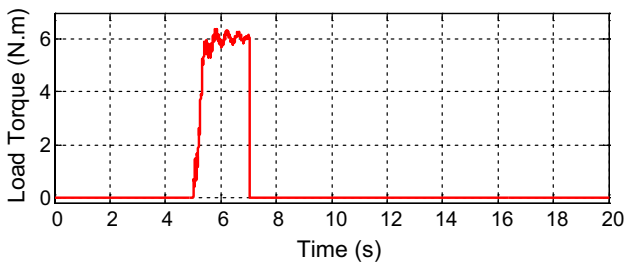
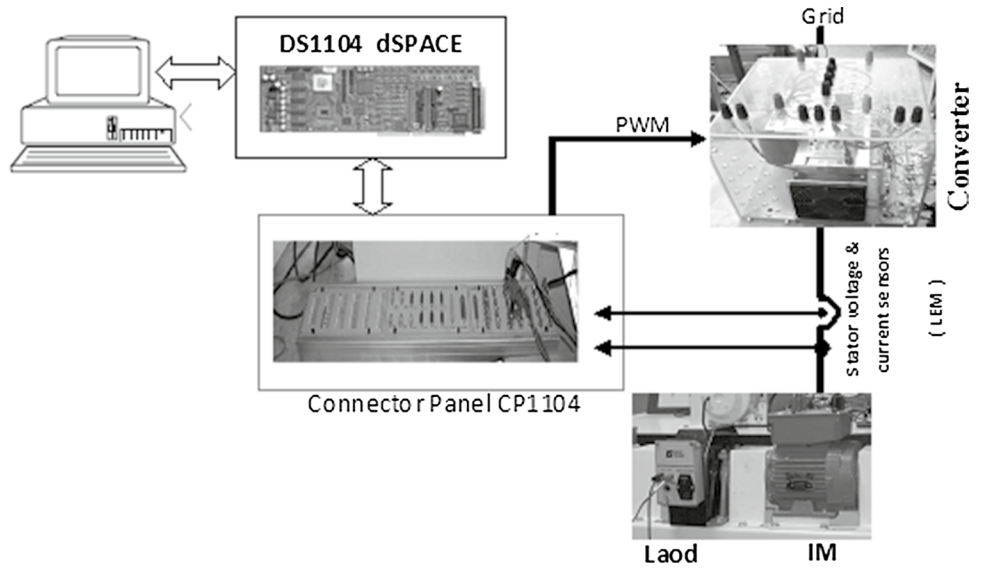


Fig. 19 Load torque variation

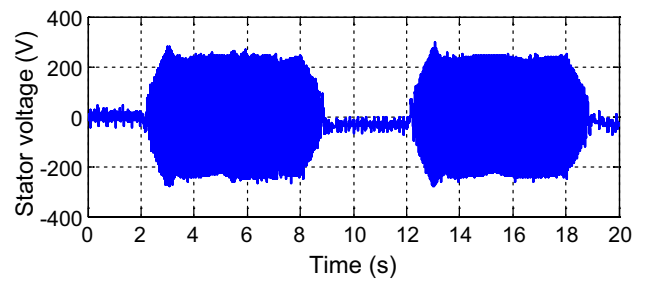


Fig. 22 The stator phase voltage

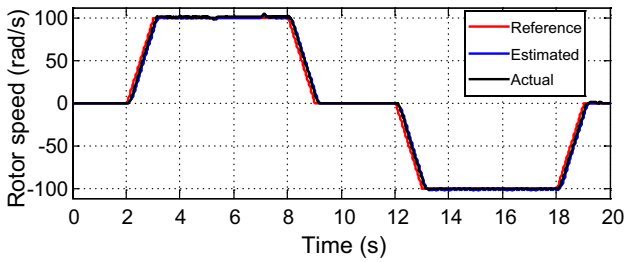


Fig. 20 Speed of induction motor

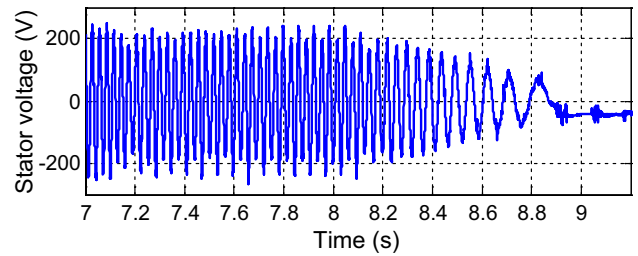


Fig. 23 Zoom of the stator phase voltage

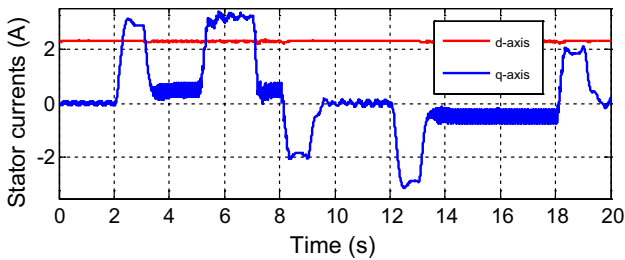


Fig. 21 I_{sd} and I_{sq} stator currents

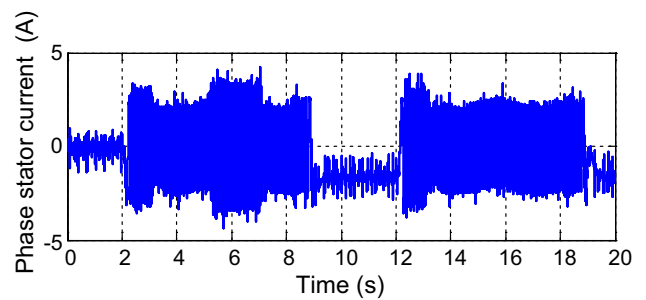


Fig. 24 The stator phase current

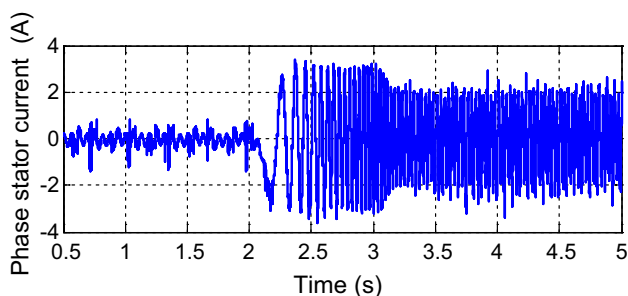


Fig. 25 Zoom of the stator phase current

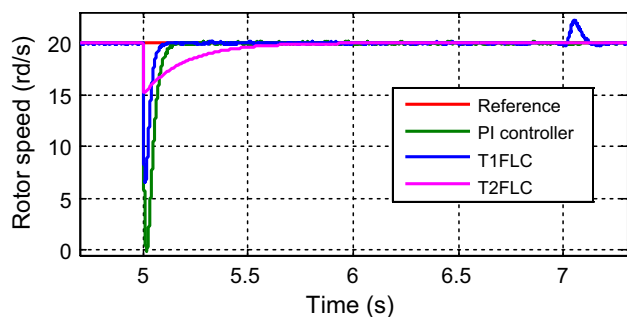


Fig. 29 Zoom of rotor speed response with T2FLC, T1FLC and PI controllers

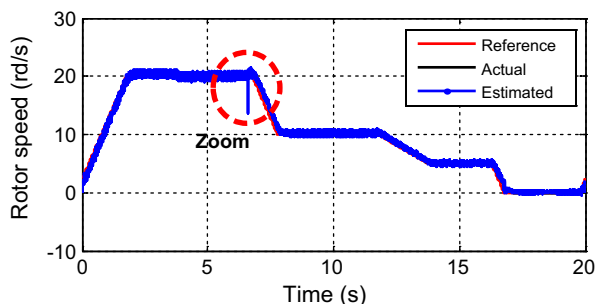


Fig. 26 Rotor speed

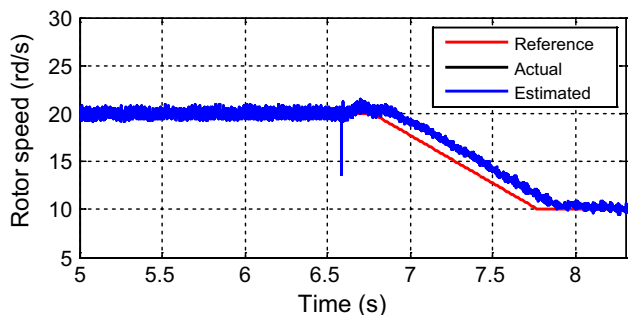


Fig. 27 Zoom of rotor speed

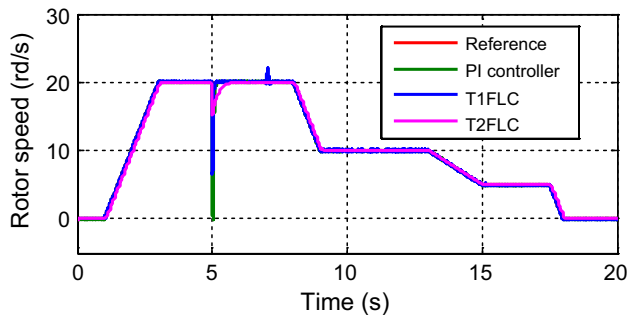


Fig. 28 Rotor speed response with T2FLC, T1FLC and PI controllers

Table 2 Performance indexes

	Rotor speed estimated (rad/s)	
	Fuzzy type 1	Fuzzy type 2
IAE	0.3155	0.05393
ISE	1.491	0.1176
ITAE	10.35	0.8214

7 Conclusions

In this paper, we have presented an improved MRAS T2FLC controller and observer for sensorless rotor speed control of induction machine. When the conventional PI has been replaced by type-2 fuzzy logic controllers, simulation and experimental results showed a clear improvement in performance regarding robustness with respect to load torque and stator resistance variations, compared with classical PI and fuzzy logic type-1 controller-based method. The simulation and experimental results reveal and demonstrate high performances of the induction motor control.

Appendix

Machine parameters and rated values

$R_r = 5.1498 \Omega$; $R_s = 12.75 \Omega$; $M = 0.4331 \text{ H}$; $L_s = 0.4991 \text{ H}$; $L_r = 0.4331 \text{ H}$; $J = 0.0035 \text{ kg m}^2$; $f = 0.001 \text{ Nm s/rd}$; $P_n = 0.9 \text{ kW}$; $n = 1400 \text{ rpm}$; $p = 2$; $f = 50 \text{ Hz}$; load torque = 10 Nm.

References

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