# $H_{\infty}$ DESIGN FOR AN INDUCTION MOTOR \*

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Abstract: A controller for an induction motor is designed using  $H_{\infty}$  control theory and input-output feedback linearization. Because of the special structure of the state-space equations governing the induction motor a Linear Parameter Varying (LPV) feedback controller, scheduled with rotor speed, is used for the inner current loop. An LPV observer is also used to estimate the flux vector. The application is based on the nonlinear model and tracking requirements of a recently published benchmark. The proposed controller delivers high performance over the entire operating range of the induction motor.

Keywords: Induction Motor, Input-Output Linearization; State-Feedback; Linear Matrix Inequalities;  $H_{\infty}$  Optimisation; LPV Gain Scheduling Control.

## 1. INTRODUCTION

Induction motors are mainly used in industry to transform electrical energy into mechanical energy. However, induction motors are rarely used as actuators because they are significantly more difficult to control than d.c. motors. Nowadays, therefore, there is great interest in developing high performance control laws to make induction motor performance rival that of the d.c. motor in a number of high precision applications (e.g. robotic applications). Induction motors are theoretically challenging for control engineers because as dynamical systems they are highly nonlinear, the flux which is to be controlled (and sometimes the rotor speed) is not available. Also, physical parameter uncertainties, such as the variation of the rotor resistance with temperature, affect significantly the dynamics of the system. Overviews of the important induction motor control techniques are given in (Bose, 1998) and (Taylor, 1994).

In this paper, the design of a controller for an induction motor is based on Linear Parameter

Variant  $H_{\infty}$ , the so-called LPV theory (Apkarian and Gahinet, 1995), (Gahinet and al, 1995), and input/output feedback linearisation. It is assumed that only the stator currents and the rotor speed are available for measurement. Broadly speaking, the control law consists of a fast LPV inner loop used to track stator current references which are generated by a non-linear input-output linrearization state feedback gain. The flux vector is estimated by an observer which, as we will see, has also an LPV structure. Robust stability and robust performance of the flux observer and the current loop are easy to prove in the LPV context. The non-linear model of the motor and tracking requirements come from the benchmark given in (Ortega et al., 2000). Full non-linear simulations, including parametric variations, saturation limits, time delay and noisy measurements demonstrate that modern  $H_{\infty}$  techniques combined with input/output linearization offer a promising and effective way to design robust controllers for induction motors.

The paper is organized as follows. The model of the induction motor is given in section 2. The controller design is described in section 3. Section

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4 includes the simulations results and conclusions are given in section 5.

# 2. INDUCTION MOTOR MODEL

We consider a squirrel-cage induction motor whose nominal physical parameters are given in table 1.

	Description	Parameter	value	Units
ſ	Stator Inductance	$L_s$	0.47	H
	Rotor Inductance	$L_r$	0.47	H
	Mutual Inductance	$L_{sr}$	0.44	H
	leakage factor $\sigma_s = \sigma_r$	$\sigma$	0.12	
	Stator resistance	$R_s$	0.8	Ω
	Rotor resistance	$ar{R}_r$	3.6	Ω
	Moment of inertia	$D_m$	0.06	$kg.m^2$
	viscous damping constant	$R_m$	0.04	N.m.s
L	Number of pole pairs	$n_p$	2	

Table 1. Nominal physical parameters of the induction motor

Under the assumption of linear magnetic circuits, a 5th order non-linear model (stator-fixed frame) of the induction motor is given by

$$G: \begin{cases} \dot{x}_1 = a_1(x_2x_5 - x_3x_4) + a_2x_1 + a_3\tau_L \\ \dot{x}_2 = a_4x_2 - n_px_1x_3 + a_5x_4 \\ \dot{x}_3 = n_px_1x_2 + a_4x_3 + a_5x_5 \\ \dot{x}_4 = a_6x_2 + a_7x_1x_3 - \gamma x_4 + a_8u_1 \\ \dot{x}_5 = -a_7x_1x_2 + a_6x_3 - \gamma x_5 + a_8u_2 \\ y = [x_1, x_4, x_5]^T \end{cases}$$
 (1)

The state vector is  $x = [\omega, \phi_a, \phi_b, i_a, i_b]^T = [x_1, x_2, x_3, x_4, x_5]^T$ , where  $\omega$  is the rotor speed,  $\phi = [\phi_a, \phi_b]^T$  are the rotor fluxes,  $i_s = [i_a, i_b]^T$  are the stator currents and  $u_s = [u_1, u_2]^T$  represents the stator voltages. The measured output is  $y = [\omega, i_a, i_b]^T$ , the control input is  $u_s = [u_1, u_2]^T$  and  $\tau_L$  is the load torque disturbance.

The outputs to be controlled are

- Shaft rotor speed  $x_1 = \omega$
- Rotor flux norm  $\Phi = \|\phi\|$

The parameters of the induction motor model are defined as follows:  $a_1 = n_p L_{sr}/(D_m L_r)$ ,  $a_2 = -R_m/D_m$ ,  $a_3 = -1/D_m$ ,  $a_4 = -1/T_r$ ,  $a_5 = L_{sr}/T_r$ ,  $a_6 = L_{sr}/(T_r \sigma L_s L_r)$ ,  $a_7 = n_p L_{sr}/(\sigma L_s L_r)$  and  $a_8 = 1/(\sigma L_s)$  where

$$T_r = \frac{L_r}{R_r}, \quad \gamma = \frac{R_s}{L_s \sigma} + \frac{L_{sr}^2}{L_s \sigma L_r T_r}$$

$$\sigma = 1 - \frac{L_{sr}^2}{L_s L_r} \tag{2}$$

with nominal values given in table 2.

The rotor resistance  $R_r$ , which will change in the experiments, and the load torque disturbance  $\tau_L$ .  $R_r$  is assumed to vary in the range  $[0.7\bar{R}_r, 1.3\bar{R}_r]$ . The change in the rotor resistance affects linearly or affinely the values of parameters  $a_4$ ,  $a_5$ ,  $a_6$ 

Parameter	r nominal value	
$a_1$	31.21	
$a_2$	-0.667	
$a_3$	-16.67	
$\bar{a}_4$	-7.66	
$ar{a}_5$	3.37	
$\bar{a}_6$	127.14	
$a_7$	33.19	
$a_8$	17.73	
$\bar{\gamma}$	197.78	
$\bar{\tau}_L$	$ar{ au}_L$ 7(Nm)	

Table 2. Nominal parameter values of the induction motor model

and  $\gamma$ . The load torque disturbance is assumed to vary in the range  $[0.25\bar{\tau}_L,\bar{\tau}_L]$ . In table 2, the nominal values of parameters subject to variation are denoted by  $\bar{a}_4$ ,  $\bar{a}_5$  etc...; similarly,  $\bar{R}_r$  in table 1.

# 3. CONTROLLER STRUCTURE AND CONTROLLER DESIGN

The controller is made up of four elements which will be designed separately. Figure 1 shows the structure considered.  $K_{lpv}$  is a current feedback controller which tracks the current setpoint reference  $i_{sref}$ . The role of  $K_{lpv}$  is to ensure good tracking over the entire operating range (i.e. when  $\omega$  is varying). It will have an LPV form and be designed using quadratic  $H_{\infty}$  gain scheduling stabilisation (Apkarian and Gahinet, 1995), (Apkarian and Adams, 1998), (Gahinet and al, 1995).

The reference current vector  $i_{sref}$  is generated through a static input-output linearization state feedback  $F_{io}$ . The flux vector is estimated by  $G_{obs}$  which, as we will see has an LPV structure. Finally,  $K_{lin}$  is a simple LTI regulator based on the linearized map relating the input  $\nu$  with the speed and the flux.

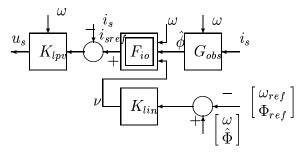


Fig. 1. Controller structure

# 3.1 Polytopic $H_{\infty}$ current feedback

Let us consider the second, third, fourth and fifth state space equations of the induction motor given in (1). These four equations constitute an affine parameter-dependent plant if the rotor speed  $x_1 =$ 

 $\omega$  is taken as the parameter. More precisely, let us define the subsystem  $G_1$  with state vector  $\xi = [x_2, x_3, x_4, x_5]^T$  having  $i_s = [x_4, x_5]^T$  as output vector and  $u_s = [u_1, u_2]^T$  as input vector. From (1)  $G_1$  can be written as follows:

$$G_1: \begin{cases} \dot{\xi} = (A_0 + \omega A_1)\xi + Bu \\ i_s = C\xi \end{cases}$$
 (3)

with

$$A_0 = \begin{pmatrix} a_4 & 0 & a_5 & 0\\ 0 & a_4 & 0 & a_5\\ a_6 & 0 & -\gamma & 0\\ 0 & a_6 & 0 & -\gamma \end{pmatrix}, \tag{4}$$

$$A_{1} = \begin{pmatrix} 0 & -n_{p} & 0 & 0\\ n_{p} & 0 & 0 & 0\\ 0 & a_{7} & 0 & 0\\ -a_{7} & 0 & 0 & 0 \end{pmatrix}$$
 (5)

$$B = \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ a_8 & 0 \\ 0 & a_8 \end{pmatrix}, \quad C = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} (6)$$

and with

$$\omega \in [\omega_{min}, \omega_{max}] = [-110, 110] \quad rad/s \quad (7)$$

Alternatively,  $G_1(t)$  admits the following polytopic state-space representation

$$G_1: \alpha(t)S_1 + (1 - \alpha(t))S_0$$
 (8)

with

$$\begin{cases} S_{0} = (\tilde{A}_{0}, B, C, 0_{2 \times 2}) \\ S_{1} = (\tilde{A}_{1}, B, C, 0_{2 \times 2}) \\ \tilde{A}_{0} = A_{0} + \omega_{min} A_{1} \\ \tilde{A}_{1} = A_{0} + \omega_{max} A_{1} \\ 0 \le \alpha(t) = \frac{\omega(t) - \omega_{min}}{\omega_{max} - \omega_{min}} \le 1 \end{cases}$$
(9)

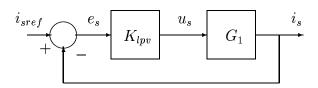


Fig. 2. Current feedback

The robust multivariable LPV controller  $K_{lpv}$  has to provide satisfactory performance over the entire operating range of the motor (i.e. when  $\omega$  varies). The design problem is tackled using the one-degree-of-freedom control structure depicted in figure 2. A mixed-sensitivity T/S loop shaping quadratic-gain  $H_{\infty}$  optimization is proposed for the design of the polytopic regulator  $K_{lpv}$ . The optimization problem is to find a stabilizing con-

trol law  $K_{lpv}$  to minimize, for all  $\omega$  in (7), in a quadratic  $H_{\infty}$  sense, the cost function

$$\left\| \begin{bmatrix} W_T T \\ W_S S \end{bmatrix} \right\|_{\infty} \tag{10}$$

 $W_S$  and  $W_T$  are used to shape the output sensitivity function  $S = (I + G_1 K_{lpv})^{-1}$  and the complementarity sensitivity function T := I - S.

$$W_S = diag(w_S, w_S), w_S = \frac{s + \omega_{BS} A_S}{1/M_S s + \omega_{BS}}$$
(11)

with  $A_S=1/500,\ M_S=2,\ \omega_{BS}=550.$  The following constant weight was chosen

$$W_T = diag(0.8, 0.8) \tag{12}$$

The polytopic regulator  $K_{lpv}$  was computed using the function hinfgs in the LMI control toolbox (Gahinet and al, 1995). The  $H_{\infty}$  performance,  $\gamma_1$ , was approximately 1.37. The closed-loop time-responses of the current controlled system for a 1 A step demand in  $i_a$  along the rotor speed trajectory

$$\omega(t) = 1000t \tag{13}$$

are given in figure 3. The decoupling is excellent. The settling time is about 60ms and the overshoot of 1.1 A is small. The overshoot on the control input is about 27 V which can be considered satisfactory.

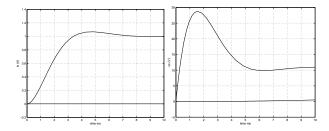


Fig. 3. Unit step closed-loop responses of the current gain scheduled feedback system along the rotor speed trajectory given in (13). Stator currents (left) and corresponding stator tensions (right)

# 3.2 Input-ouput linearization

Let  $Y=[\omega,\Phi]$  with  $\Phi=\sqrt{\phi_a^2+\phi_b^2}$ . For the induction motor, the relative degree is well defined and the zero dynamics are stable; thus the input-output linearization can be performed. Let us consider the three first equations of (1). Assuming that the rotor speed and the flux vector  $\phi$  are available for measurement, our aim is to construct a non-linear relation to produce an ideal reference stator current vector  $i_{sref}$  to make the closed-loop

input-output map  $T_{\nu,Y}$  linear.

Let  $\nu_2$  be the first derivative of the flux. From (1) and using the definition of  $\Phi$ , we get

$$\dot{\Phi} = \frac{\phi_a(\bar{a}_4\phi_a + \bar{a}_5i_{aref}) + \phi_b(\bar{a}_4\phi_b + \bar{a}_5i_{bref})}{\Phi}$$

$$:= \nu_2 \tag{14}$$

Let  $\nu_1$  be the first derivative of the rotor speed  $\omega = x_1$ .  $\nu_1$  is given by

$$\dot{\omega} = a_1(\phi_a i_{bref} - \phi_b i_{aref}) + a_2 \omega + a_3 \bar{\tau}_L$$

$$:= \nu_1 \tag{15}$$

(14) and (15) can be rewritten as

$$\begin{bmatrix} \phi_a \bar{a}_5 & \phi_b \bar{a}_5 \\ -a_1 \phi_b & a_1 \phi_a \end{bmatrix} \begin{bmatrix} i_{aref} \\ i_{bref} \end{bmatrix} = \begin{bmatrix} \Phi(\nu_2 - \bar{a}_4 \Phi) \\ \nu_1 - a_3 \bar{\tau}_L - a_2 \omega \end{bmatrix}$$
(16)

Thus,

$$F_{io}(\omega, \phi, \nu) := \begin{bmatrix} i_{aref} \\ i_{bref} \end{bmatrix}$$

$$= \frac{1}{a_1 \bar{a}_5 \Phi^2} \begin{bmatrix} a_1 \phi_a & -\phi_b \bar{a}_5 \\ a_1 \phi_b & \phi_a \bar{a}_5 \end{bmatrix} \begin{bmatrix} \Phi(\nu_2 - \bar{a}_4 \Phi) \\ \nu_1 - a_3 \bar{\tau}_L - a_2 \omega \end{bmatrix}$$
(17)

The nonlinear state feedback control law  $i_{sref} = [i_{aref}, i_{bref}]^T$  given in (17) is well defined whenever  $\Phi \neq 0$  and results in

$$G_2: \begin{cases} \dot{\omega} = \nu_1 \\ \dot{\Phi} = \nu_2 \\ Y = [\omega, \Phi] \end{cases}$$
 (18)

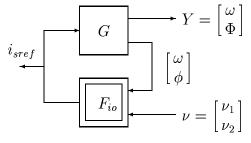


Fig. 4. Input-Ouput linearization

# 3.3 LPV flux observer design

An LPV flux estimator can be designed using both tension and current measurements. The design is based on the plant  $G_3$  which is equal to  $G_1$  but with  $C = I_4$ , where  $a_4$  is the uncertain parameter which is linearly dependent on the rotor resistance. The aim is to design an observer able to maintain a good flux estimation at any speed despite uncertainty in the rotor resistance. We know that  $G_1$  admits a polytopic representation  $(S_0, S_1)$  where  $a_4$  enters linearly in  $S_0$  and  $S_1$ . Therefore,

we can extract  $a_4$  from the polytopic representation using the concept of Linear Fractional Transformations (see e.g (Zhou et al., 1995)). Our aim is to take explicitly into account the uncertainty due to the rotor resistance variation in the design of the flux observer. Let us consider the block diagram of figure 5. The matrix  $\delta I_2$  represents the rotor resistance uncertainty. It is normalized in such a way that  $\|\delta\|_{\infty} \leq 1$ .

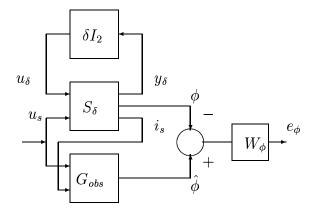


Fig. 5. Flux observer interconnection structure

More precisely, the uncertain parameter  $a_4$  can be rewritten as  $a_4 = \bar{a}_4(1 + \delta_a)$ ,  $|\delta_a| < \delta_{max}$ ,  $\delta_{max} = \frac{a_{4max} - a_{4min}}{2\bar{a}_4}$ . Let us consider, the following LTI system which corresponds to  $S_0$  (the vertex for  $\omega = \omega_{min}$ ).

$$S_{0\delta}: \begin{cases} \dot{x} = (\bar{a}_4 A_2 + \omega_{min} A_1) x + P_{12} u_{\delta} / \delta_{max} + Bu \\ y_{\delta} = P_{21} x \\ y = x \end{cases}$$
(19)

with

$$A_{2} = \begin{pmatrix} 1 & 0 & -L_{sr} & 0\\ 0 & 1 & 0 & -L_{sr}\\ -\frac{L_{sr}}{\sigma L_{s}L_{r}} & 0 & \frac{L_{sr}^{2}}{\sigma L_{s}L_{r}} & 0\\ 0 & -\frac{L_{sr}}{\sigma L_{s}L_{r}} & 0 & \frac{L_{sr}^{2}}{\sigma L_{s}L_{r}} \end{pmatrix} (20)$$

$$P_{12} = \begin{pmatrix} 1 & 0 & -\frac{L_{sr}}{\sigma L_{s}L_{r}} & 0\\ 0 & 1 & 0 & -\frac{L_{sr}}{\sigma L_{s}L_{r}} \end{pmatrix}^{T} (21)$$

$$P_{21} = \begin{pmatrix} 1 & 0 & -L_{sr} & 0\\ 0 & 1 & 0 & -L_{sr} \end{pmatrix} (22)$$

and where  $A_1$  and B are given in (5) and (6) respectively.

It is easy to show that

$$S_0 = F_l(\delta I_2, S_{0\delta}) \tag{23}$$

and

$$S_1 = F_l(\delta I_2, S_{1\delta}) \tag{24}$$

where  $S_{1\delta}$  is the same as  $S_{0\delta}$  if one replaces  $\omega_{min}$  by  $\omega_{max}$  in (19) and where  $\delta := \delta_a/\delta_{max}$ .

The observer design is based on the interconnection of figure 5. The regulated output is the weighted flux estimation error  $e_{\phi}$ . The measured outputs are the current vector and the supply voltage. Set

$$w = \begin{bmatrix} u_{\delta} \\ u_{s} \end{bmatrix} \quad z = \begin{bmatrix} y_{\delta} \\ e_{\phi} \end{bmatrix} \quad u = \hat{\phi} \quad y = \begin{bmatrix} u_{s} \\ i_{s} \end{bmatrix} \quad (25)$$

The observer design consists of finding  $u = G_{obs}y$  to minimize, for all admissible speed trajectories, the closed-loop quadratic  $H_{\infty}$  performance from w to z ( $\gamma_3$ ). According to the small gain theorem, robust stability and robust performance will be achieved whenever  $\gamma_3 < 1$ . As is customary in  $\mu$ -synthesis e.g. (Zhou et al., 1995), (Balas and al, 1993),(Boyd et al., 1994) we can enforce a better minimization of  $T_{e_{\phi},u_s}$ , by scaling  $u_{\delta}$  by  $d_s = diag(d_1, d_2)$  with  $d_1 > 0$ ,  $d_2 > 0$  and  $y_{\delta}$  by  $d_s^{-1}$  without affecting performance and stability.

 $W_{\phi}$  was chosen as the identity matrix and after some trial and error  $d_s$  was selected as

$$d_s = diag(3.7, 3.7) (26)$$

The  $H_{\infty}$  performance,  $\gamma_3$ , was approximately 1.17. It is easy to check that  $\|T_{y_{\delta},u_{\delta}}\|_{\infty} < 1$ . Therefore the observer is robustly, quadratically stable (i.e. stable for any variation of the rotor resistance and for any speed trajectory). The frequency responses of  $T_{e_{\phi},u_s}$  for various values  $\omega$  and  $a_4$  as given in figure 6.

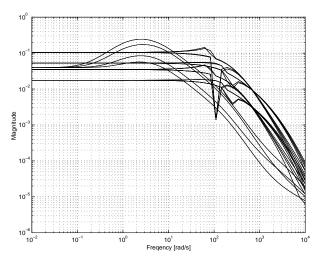


Fig. 6. Singular value plot of  $T_{e_{\phi},u_s}$  for 5 and 4 equally spaced values of  $\omega$  and  $a_4$ 

#### 3.4 Control of the speed and the flux

To control the speed and the flux, one has to consider the plant  $G_2$  given in (18). It is a chain of 2 integrators, representing the nominal closed-loop input-output map  $T_{Y,\nu}$  when the condition

 $i_s = i_{sref}$  holds. The regulator can be designed using any classical linear method. In this example, a simple diagonal gain was found to provide good tracking and sufficient robustness. Trial and error led to  $K_{lin} = -diag(343, 286)$ , for which the closed-loop bandwidth is approximately 300rad/s.

#### 4. SIMULATION RESULTS

Full non-linear simulations were carried out for the speed and flux step demand profiles and for parameter variations shown in figure 7. Figure 7 represents the worst case rotor resistance changes which may correspond to breaks in the rotor bars. Saturation limits on the stator tensions and on the the reference currents were included to prevent peaks especially at initialization. Zero-order holds with a sampling frequency of  $F_s=4kHz$  and a unit time delay  $T_s=1/F_s$  were added in the feedback loop to simulate a digital implementation. The saturation limits on the reference current components were taken as

$$|i_{srefi}| < 7A, \quad i = 1, 2$$
 (27)

and for the control signal

$$|u_i| < 210V, \quad i = 1, 2 \tag{28}$$

The flux and speed demands were filtered through the following second order filters

$$F_{ref} := \frac{\omega_n^2}{s^2 + 2\xi\omega_n s + \omega_n^2} \tag{29}$$

with  $\omega_n = 8, \, \xi = 0.8$ .

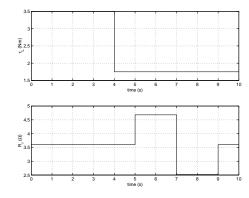


Fig. 7. Load torque and rotor resistance variations

#### 4.1 Controller using the LVP observer

The estimation of the flux is provided by the LPV observer. The response of the rotor speed, given in figure 8. Figure 9 shows that the modulus of the control signal  $(u_s)$  stays within the limits

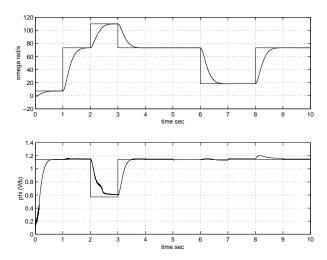


Fig. 8. Speed and flux tracking with the LPV observer  $G_{obs}$ 

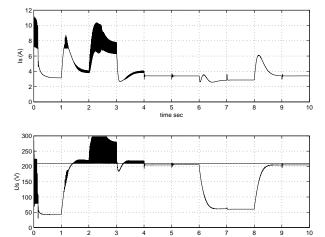


Fig. 9. Stator voltage modulus  $||u_s||$  and stator current modulus  $||i_s||$  with limits (LPV observer)

except between t=2s and t=3s when the speed demand is 110rad/s. The stator current modulus remains below the limit of 12A.

# 5. CONCLUSIONS

This paper has presented a non-linear controller design for an induction motor. It was assumed that only the stator currents and the rotor speed were available for measurement. The controller has a structure made up of four different subsystems: an LPV feedback current loop, a non-linear linearizing state feedback, an LPV observer and an LTI regulator. An LMI-based approach, has been proposed to design a quadratically stable flux observer and an output feedback regulator to track the stator currents. In both cases we have obtained a scheduled time varying system (LPV) which ensures a finite  $L_2$  attenuation for a given closed-loop transfer function which represents the design requirements (tracking robustness for the

LPV current regulator and flux estimation error for the LPV observer). The main advantage of using LPV methods is that they provide a systematic way of designing an  $H_{\infty}$  flux observer for the induction motor assuming that the rotor speed is available. Stability of the flux estimator was demonstrated using small-gain based analysis. The results for the benchmark were found to be satisfactory. Due to the very low order of the its components ( $K_{lpv}$  is a 2 by 2 speed dependent regulator with only 6 states, the speed dependent flux observer has only 4 states, 4 inputs and 2 outputs) the proposed control law is easy to implement and can work with relatively slow DSP cards. A drawback of the proposed controller is that it requires a measure of the rotor speed to update the LPV parts of the control law.

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