Robust DTC Control of Doubly-Fed Induction Machines Based on Input-Output Feedback Linearization Using Recurrent Neural Networks

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Abstract

This paper describes a novel Direct Torque Control (DTC) method for adjustable speed Doubly-Fed Induction Machine (DFIM) drives which is supplied by a two-level Space Vector Modulation (SVM) voltage source inverter (DTC-SVM) in the rotor circuit. The inverter reference voltage vector is obtained by using input-output feedback linearization control and a DFIM model in the stator a-b axes reference frame with stator currents and rotor fluxes as state variables. Moreover, to make this nonlinear controller stable and robust to most varying electrical parameter uncertainties, a two layer recurrent Artificial Neural Network (ANN) is used to estimate a certain function which shows the machine lumped uncertainty. The overall system stability is proved by the Lyapunov theorem. It is shown that the torque and flux tracking errors as well as the updated weights of the ANN are uniformly ultimately bounded. Finally, effectiveness of the proposed control approach is shown by computer simulation results.

Key Words: Artificial Neural Network, Direct Torque Control, Doubly Fed Induction Machine, Feedback linearization

I. INTRODUCTION

Doubly-fed induction generator (DFIG) wind turbines with converters rated at about 25–30% of the generator rating are becoming increasingly popular. DFIG-based wind turbines offer variable speed constant frequency (VSCF) operation, four-quadrant and decoupled active and reactive power capabilities, lower converter cost and reduced power loss when compared to other wind turbines. In a DFIG system, the stator is usually connected to a phase grid directly and the rotor is fed by two back-to-back pulse width modulation (PWM) or space vector modulation (SVM) inverters (Fig.1). This arrangement provides flexibility of operation at sub-synchronous and super-synchronous speeds in both generating and motoring modes. The power inverter only needs to be rated for a fraction of the total output power; the fraction depends on the permissible sub- and super-synchronous speed range [1].

So far both vector control (VC) and direct torque control (DTC) methods have been widely applied to squirrel cage induction machine drives [2]. Although field oriented vector methods have been applied to DFIM drives [3], [4], little attention has been paid to DTC and flux control of these types of drives. In field the oriented methods applied to DFIM drives, in order to design the rotor current controllers, it is necessary to neglect the voltage drop across the stator leakage impedance [3]–[5]. Therefore, the injected active and reactive powers to the stator can be independently controlled. This assumption leads to a steady-state error in both the motoring and generating modes of operation. In [6] the conventional bang-bang DTC method is combined with the direct rotor flux field oriented control method and applied to an adjustable speed DFIM drive. The DTC controller is designed based on neglecting the voltage drop across the rotor resistance. In [7] and [8], a back stepping tracking controller has been introduced for a DFIM drive. However, only unity power factor operation, measured on the stator supply voltage side, has been proposed. In [9], a stator power factor adjustable DTC method was proposed to control wound rotor double-fed induction machines. The implementation of the rotor flux based method and the stator power factor adjustable method...
requires information from both the stator side and the rotor side. Consequently the complexity of the control system is increased.

In [10], a DTC strategy was proposed for the DFIG used in variable speed wind power generation systems. Instead of the rotor flux or the stator power factor, the power factor of the rotor is controlled and kept at unity. This results in a control system, simplicity and a reduction of the inverter power rating. In order to reduce the torque and the stator current ripple, an ANN based hysteresis torque and flux controller was proposed in [11] for DFIMs. Simulation results show the superiority of this method over conventional bang-bang DTC control methods. However, there is no proof of the stability of the overall system. Moreover, the robustness of the system to machine parameter variations was not investigated.

Instead of using one converter in the rotor circuit, two converters are used in each side of a DFIM, i.e. the rotor and the stator [12]. One converter controls the stator flux while the other controls the rotor flux separately. This method, known as the Dual-DTC strategy, uses two switching tables for the stator and rotor converters. Although interesting features and more degree of freedom arise by using the Dual-DTC approach, higher complexity and an increase in the cost of the system are inevitable.

The nonlinear control techniques reported so far for DFIM drives are mostly parameter dependent and seriously affected by machine parameter deviations.

Our contribution in this paper is to introduce a DTC-SVM control scheme for DFIMs based on the input-output linearization technique, using a motor fifth order model in the stationary two axis reference frame with the stator currents and rotor fluxes as state variables. It is well known that input-output feedback control method needs to know the exact model of a system with known parameters [13]. As a result, such a controller cannot guarantee system robustness against parameter uncertainties. In the last decade, some researchers have tried to solve this problem by using the adaptive input-output feedback control method or the adaptive back stepping control approach [14], [15]. These two methods can make a drive system robust and stable to unknown constant parameters. To overcome this problem, this paper uses a two layer recurrent ANN to estimate the lumped uncertainty function of the system. Having estimated this function, it will be shown that the drive system is robust to most electrical parameter variations.

ANN weights are on-line tuned based on an estimation law which is obtained by the Lyapunov stability theorem. As a result, in this method, there is no need to know the actual value of the function which is supposed to be estimated. In addition, by this method the boundedness of the tracking errors and the updated ANN weights can be guaranteed.

II. Doubly-Fed Induction Machine Model

For a linear magnetic circuit and balanced operating conditions, the equivalent two-phase model of a symmetrical DFIM, represented in the fixed stator a-b reference frame with its stator directly connected to the grid is:

\[
\frac{d i_{as}}{dt} = - \left( \frac{R_s}{L_s} + \frac{R_{l_m}}{L_s L_r} \right) i_{as} + \frac{R_{l_m}}{L_s L_r} \psi_{ar} + \frac{\omega_r}{L_s} \psi_{br} + \frac{u_{as}}{L_s} + \frac{L_m}{L_s} u_{sr} \\
\frac{d i_{bs}}{dt} = - \left( \frac{R_s}{L_s} + \frac{R_{l_m}}{L_s L_r} \right) i_{bs} + \frac{R_{l_m}}{L_s L_r} \psi_{ar} + \frac{\omega_r}{L_s} \psi_{br} + \frac{u_{bs}}{L_s} + \frac{L_m}{L_s} u_{sr} \\
\frac{d \psi_{ar}}{dt} = \frac{R_{l_m}}{L_r} i_{as} - \frac{R_r}{L_r} \psi_{ar} - \omega_r \psi_{br} + u_{ar} \\
\frac{d \psi_{br}}{dt} = \frac{R_{l_m}}{L_r} i_{bs} - \frac{R_r}{L_r} \psi_{br} + \omega_r \psi_{ar} + u_{br}
\]

where \(i_s, \psi_r, u_s, R, L\) denote the stator currents, the rotor flux linkage, the stator terminal voltage, the rotor terminal voltage, the resistance and the inductance, respectively. The subscript s and r stand for the stator and rotor, while the subscripts a and b stand for the vector component with respect to a fixed stator reference frame. \(\omega_r\) denotes the rotor electrical speed and \(L_m\) is the mutual inductance. \(L_s = L_r(1 - (\frac{L_s}{L_s + L_r}))\) is the redefined leakage inductance.

The generated torque of a DFIM in terms of the stator current and the rotor flux linkage components is as follows:

\[
T_e = \mu (\psi_{ar} i_{bs} - \psi_{br} i_{as})
\]

where \(\mu = \frac{3P L_m}{2 T_r}\) and \(P\) is the number of poles. The mechanical dynamic equation is given by:

\[
J \frac{d \omega_m}{dt} + B \omega_m + T_L = T_e
\]

where \(J\) and \(B\) denote the moment of inertia of the motor and the viscous friction coefficient, respectively. \(T_L\) is the external load torque and \(\omega_m\) is the rotor mechanical speed \((\omega_r = (\frac{P}{2}) \omega_m)\).

\[
x = \begin{bmatrix} i_{as} & i_{bs} & \psi_{ar} & \psi_{br} \end{bmatrix}^T
\]

Let be the state vector and let the generated torque \(T_e\) be the output \(y\) of the dynamic system (1), that is:

\[
y = T_e = \frac{3P L_m}{2 T_r} (\psi_{ar} i_{bs} - \psi_{br} i_{as})
\]

It is well-known that torque control is very important for high-performance motion control [16]. However, from (1) and (5) the generated torque \(T_e\) of a DFIM can be viewed as a nonlinear function of the state variables \(x\) of the dynamic model (1), \(i_{as}, i_{bs}, \psi_{ar}\) and \(\psi_{br}\). Therefore, it is difficult to evaluate the torque response from (5) by the control inputs \(u_{ar}\) and \(u_{br}\) designed for the model (1). Therefore, based on (1), it is a task for the torque control of induction machines in industrial and practical applications.

III. Input-Output Feedback Control

For the proposed nonlinear controller, the state coordinate transformation is applied. Therefore, the state-coordinates transformed model from (1) can be rewritten in the following compact form:

\[
\dot{x} = f(x) + g_1 u_{ar} + g_2 u_{br}
\]
where \( x \) is defined in (4) and:

\[
f(x) = \begin{bmatrix}
\frac{-R_s}{L_r} + \frac{R_L^2}{L_r^2} i_{as} + \frac{R_s}{L_r} L_m \frac{\omega _L}{L_r} \psi _{br} + \frac{u_{as}}{L_r} \\
\frac{-R_s}{L_r} + \frac{R_L^2}{L_r^2} i_{bs} + \frac{R_s}{L_r} L_m \frac{\omega _L}{L_r} \psi _{ar} + \frac{u_{bs}}{L_r} \\
\frac{R_s}{L_r} \frac{L_m}{L_s} i_{as} - \frac{R_s}{L_r} \psi _{br} - \omega _L \psi _{br} \\
\frac{R_s}{L_r} \frac{L_m}{L_s} i_{bs} - \frac{R_s}{L_r} \psi _{ar} + \omega _L \psi _{ar}
\end{bmatrix}
\]  

(7)

\[
g_1 = \begin{bmatrix}
-\frac{L_m}{L_r L_s} & 0 & 1 & 0
\end{bmatrix}^T
\]

(8)

\[
g_2 = \begin{bmatrix}
0 & -\frac{L_m}{L_r L_s} & 0 & 1
\end{bmatrix}^T.
\]

(9)

At this stage, the developed torque \( T_c \) and the squared modules of the rotor flux linkage, \( |\psi _r|_2 = |\psi _{ar}|^2 + |\psi _{br}|^2 \), are required to be the controlled outputs. Therefore, by considering:

\[
h_1(x) = \frac{3P L_m}{2} \psi _{ar} i_{bs} - \psi _{br} i_{as}
\]

(10)

\[
h_2(x) = |\psi _{ar}|^2 + |\psi _{br}|^2
\]

The controller output errors \( e_1 \) and \( e_2 \) are defined as follows:

\[
e_1 = T_c - T_c^e
\]

\[
e_2 = (|\psi _{ar}|^2 + |\psi _{br}|^2) - |\psi _r|^2 = |\psi _r|^2 - |\psi _r|^2.
\]

(11)

Then:

\[
\dot{e} = F + Gv
\]

(12)

where:

\[
F = \begin{bmatrix}
F_{i1} \\
F_{i2}
\end{bmatrix}
\]

\[
F_{i1} = \frac{3P L_m}{2} \left(-\frac{R_s}{L_r} + \frac{R_L^2}{L_r^2} + \frac{R_s}{L_r} \frac{L_m}{L_r} \frac{\omega _L}{L_r} (i_{bs} \psi _{ar} - i_{as} \psi _{br}) - \frac{\omega _L L_m}{L_r L_s} (|\psi _{ar}|^2 + |\psi _{br}|^2) - \frac{u_{as} \psi _{br} - u_{bs} \psi _{ar}}{L_r} - \frac{i_{as} \psi _br - i_{bs} \psi _{ar}}{L_s}
\right)
\]

\[
F_{i2} = \frac{2R_s}{L_r} \left(i_{as} \psi _{br} - i_{bs} \psi _{ar}ight) - \frac{R_s}{L_r} \left(|\psi _{ar}|^2 + |\psi _{br}|^2\right)
\]

\[
G = \begin{bmatrix}
3P \frac{L_m}{2} (i_{bs} \psi _{ar} - i_{as} \psi _{br}) - 3P \frac{L_m}{2} (i_{as} \psi _{br} + i_{bs} \psi _{ar}) \\
\frac{3P L_m}{2} \psi _{ar} - \frac{3P L_m}{2} \psi _{br}
\end{bmatrix}
\]

(13)

\[
v = \begin{bmatrix}
v_{as} \\
v_{bs}
\end{bmatrix}
\]

Based on the input-output feedback linearization method, the following control inputs are introduced for the system:

\[
v = G^{-1}[-F - Ke], \quad K > 0.
\]

(14)

\[
\dot{v} = G^{-1}[-\dot{F} - Ke], \quad \dot{K} > 0.
\]

IV. ANN BASICS

Recent developments in the field of Neuro-computing present ANN as a potential controller for slip-energy recovery drive systems. ANNs are valuable on several counts. For example, they are capable of performing massive parallel processing and can provide a good degree of fault tolerance. Furthermore, ANNs are adaptive, in that they can learn and infer solutions from the data presented to them by learning the underlying relationships even if these relationships are difficult to find and describe. ANNs can precisely process data which may be different from the data used in the training stage. Moreover, ANNs may be able to handle imperfect or noisy data. Finally, ANNs can approximate a wide range of nonlinear functions to any desired degree of accuracy under certain conditions [17]. In the application of a drive system, ANNs can be trained to emulate the nonlinear dynamics of that system by learning a suitable sets of input/output patterns for it [18].

The structure of a two-layer ANN has been shown in Fig. 2. \( \phi \) is defined as a collection of NN weights and \( \phi \) as basis functions. Then the net output is:

\[
y = W^T \phi (x)
\]

(16)

Let \( S \) be a compact simply connected set of \( \mathbb{R}^n \). With the map \( f : S \to \mathbb{R}^n \). Define \( C^m(S) \) as the functional space such that \( f \) is continuous [19]. A general nonlinear function \( f(x) \in C^m(S), x(t) \in S \), can be approximated by an ANN as:

\[
f(x) = W^T \phi (x) + \epsilon (x)
\]

(17)

where \( \epsilon (x) \) is the bounded NN functional reconstruction error vector and \( \phi \) is the sigmoid activation function [20].

V. ANN CONTROLLER DESIGN

In order to make the drive control system robust against machine parameter uncertainties, the ANN controller is designed as:

\[
v = G^{-1}[-\hat{F} - Ke], \quad K > 0
\]

(18)

where \( \hat{F} \) is the online estimate of the nonlinear function \( F \), which is the output of the two layer ANN (with the general structure shown in Fig. 2) through (17) as:

\[
F = W^T \phi + \epsilon
\]

(19)

And:

\[
\dot{\hat{F}} = \dot{\hat{W}}^T \phi.
\]

(20)

Note that in (19) \( W \), as a collection of NN weights, is constant and bounded by a known positive value. Also, the reconstruction error is bounded by a known value.

Equations (11), (18), (19) and (20) lead to:

\[
\dot{e} = \hat{W}^T \phi - Ke + \epsilon.
\]

(21)
It is noted that the actual inputs to the chosen ANN are the rotor fluxes, the stator currents, the rotor speed and the torque reference. Using the Lyapunov theory and considering (18) and (21), the ANN weights may be estimated on-line by the following equation:

$$\dot{\hat{w}}^T = \Gamma \Phi e^T - k_\alpha \Gamma \| e \| \dot{\hat{w}}. \quad (22)$$

By estimating the function $\hat{F}$ and using the Lyapunov analysis, it can be seen that the torque and stator flux error signals as well as the ANN updated weights are all UUB [20].

VI. STATOR ACTIVE AND REACTIVE POWER CONTROL

For the generating mode of operation of DFIMs, it is required to regulate the stator active-reactive power, whose references are $P_s^*$ and $Q_s^*$, respectively. In [21], a synchronous $d$ and $q$ axis rotating reference frame with $d$ axes coinciding with the space voltage vector for the main ac supply has been considered. Stator active-reactive power control has been expressed in terms of the stator current as:

$$i_d^* = \frac{3}{2} \frac{Q_s^*}{U} \quad i_q^* = \frac{3}{2} \frac{P_s^*}{U} \quad (23)$$

It is assumed that a prime mover maintains the rotor speed at constant value during the generating mode. For the motoring mode of operation, the torque and stator reactive power control objectives have been converted to stator current control [21]. In addition, the rotor flux references can be obtained as:

$$\psi_d^* = -\frac{1}{\beta \omega_0} \left( -\frac{R_s}{\sigma} i_d^* - \alpha_0 i_d^* \right)$$
$$\psi_q^* = -\frac{1}{\beta \omega_0} \left( \frac{R_s}{\sigma} i_q^* - \alpha_0 i_q^* - \frac{1}{\sigma} U \right) \quad (24)$$

Therefore, the rotor flux reference and the torque reference are calculated as:

$$T_r^* = \mu \left( \psi_d^* i_q^* - \psi_q^* i_d^* \right). \quad (25)$$

VII. STABILIZATION OF ROTOR DC LINK VOLTAGE

It is necessary to maintain the rotor dc-link voltage constant during drive system operation. This can be achieved using a supply side three-phase PWM converter as shown in Fig. 3. With proper control of this converter, the rotor dc link voltage can be maintained constant regardless of the magnitude and the direction of the rotor power. The vector control of this inverter, with a reference frame oriented along the stator (or supply) voltage vector position, enables independent control of the active and reactive powers flowing between the supply and the supply-side converter [22], [23]. At this end, a capacitor in the dc-link and supply-side converter is used to mitigate variations of the capacitor voltage due to variations of the rotor power.

Using the control method described in [22], the $d$ and $q$ axis equations corresponding to the main ac power supply in a particular synchronous rotating reference frame with a $d$ axis coinciding with the voltage space vector are given by:

$$\frac{d}{dt} i_d = \frac{1}{L} (v_d - R_i i_d + \omega_L i_q - v_d)$$
$$\frac{d}{dt} i_q = \frac{1}{L} (-R_i - \omega_L i_d - v_q). \quad (26)$$

One may note that in this reference frame, the $d$ axis being coincident with the main space voltage vector means that:

$$v_d = v_s, \quad v_q = 0.$$  

By considering $i_d^*$ and $i_q^*$ as reference currents, the current errors are:

$$\epsilon_1 = i_d - i_d^* \quad \epsilon_2 = i_q - i_q^* \quad (27)$$

Consequently, the system error dynamics are:

$$\dot{\epsilon}_1 = \frac{v}{L} - \frac{R}{L} i_d + \omega_L i_q - \frac{v_d}{L} - i_d^*$$
$$\dot{\epsilon}_2 = -\frac{R}{L} i_q - \omega_L i_d - v_q - i_q^*. \quad (28)$$

Define the AC side inverter reference voltages as:

$$v_d = L \left( \frac{v_d}{L} - \frac{R}{L} i_d + \omega_L i_q - i_d^* + k \epsilon_1 \right)$$
$$v_q = L \left( -\frac{R}{L} i_q - \omega_L i_d - i_q^* + k \epsilon_2 \right). \quad (29)$$

Linking (27) and (28), gives:

$$\dot{\epsilon}_1 = -k \epsilon_1$$
$$\dot{\epsilon}_2 = -k \epsilon_2. \quad (30)$$

Considering the Lyapunov function as:

$$V = \frac{1}{2} \epsilon_1^2 + \frac{1}{2} \epsilon_2^2. \quad (31)$$

The derivative of this function becomes:

$$\dot{V} = \epsilon_1 \epsilon_1 + \epsilon_2 \epsilon_2 = -k (\epsilon_1^2 + \epsilon_2^2) < 0. \quad (32)$$

Neglecting the losses in the supply side converter, the injected active power to the rotor is:

$$P_r = E_{i_{tor}}. \quad (33)$$

Also, from Fig. 3, one can obtain:

$$\frac{3}{2} v_d i_d = E_{i_{os}} \quad (34)$$
$$C \frac{dE}{dt} = i_{os} - i_{or}. \quad (35)$$
By substituting $i_d$ and $i_q$ from equations (33) and (34) into (34), the rotor dc link voltage variation is obtained as:

$$dE = \frac{1}{C} \left( \frac{3}{2} \frac{v_d i_d}{E} - P_r \right) dt. \tag{36}$$

Based on the proposed theory, the block diagram shown in Fig. 4 is proposed to keep the rotor dc link voltage constant.

VIII. SYSTEM SIMULATION

An overall block diagram of the proposed control approach is shown in Fig. 5. The proposed approach has been applied to a three-phase, 5 kW, 380 V, 6-pole, 50 Hz DFIM drive and the simulation results are shown in Fig. 6-11. The simulation results shown in Fig. 6 are for the motoring mode of operation, as well as the sub and super synchronous speeds, with $R_r = 3R_{rn}$, $R_s = 3R_{sn}$.

The motor load torque varies between 10 Nm and 30 Nm for each half second. $R_{rn}, R_{sn}$ are the nominal rotor and stator resistances, respectively. From these results, it can be seen that the proposed torque and flux control scheme can quickly track the reference commands and is robust against parameter uncertainties. In this case, based on the proposed torque and flux control, a PI controller is used as the speed controller.

Fig. 7 and Fig. 8 show the simulation results obtained for the sub and super synchronous speed generator operating conditions with $\omega_r = 290 \text{ rad/sec}$ and $\omega_r = 375 \text{ rad/sec}$, respectively. It is obvious from these results that the proposed controller has high performance. In Fig. 6 the rotor active power is positive due to increases in the stator and rotor resistances. It is concluded from these Figs that ANN can estimate the system uncertainties and make the system robust against rotor and stator resistance variations.

Fig. 9 shows the drive system performance in generating mode above the synchronous speed. These results are obtained for the nominal condition, and the active and reactive power references as well as the torque and flux references are also shown in Fig. 9. The DC link voltage is represented by E.

From Fig. 9, it is clear that in addition to active and reactive power control objective, which leads to the torque and flux control strategy, the DC link voltage is successfully stabilized by the input-output feedback linearization method.

For the sake of comparison, the method in this paper is compared with the one in [24], where a direct active and reactive power control strategy for DFIGs has been used. This method was first used in three-phase PWM rectifiers.
where the converter switching states were selected from an optimal switching table based on the instantaneous errors between the reference and estimated values of the active and reactive powers, and the angular position of the estimated converter terminal voltage vector. In [24] two three-level hysteresis comparators were used to generate the respective active and reactive power states. Based on these states and also the position of the stator flux in the rotor reference frame, the optimal rotor voltage vector is selected. First the stator flux is estimated in the stationary reference frame using the following equation:

$$\psi_s = \int (V_s^* - R_s I_s^*) dt.$$  \hspace{1cm} (37)

Since the amplitude and frequency of stator voltage are relatively fixed, an accurate estimation of the stator flux is provided using the above relation. It is then transformed to the rotor reference frame using the rotor angular position.

The presence of the rotor resistance is a critical problem in this method, since the impact of the rotor resistance on the rotor flux (as in the following relation) has been neglected:

$$\frac{d\psi_r}{dt} = V_r^* - R_r I_r^*.$$  \hspace{1cm} (38)

Therefore, the same problems in stator flux estimation at low speeds can arise in rotor flux estimation near synchronous speed and for high rotor resistance machines.

The method in [24] has been simulated and the results are depicted in figures 10 and 11. It is obvious that the proposed controller is superior in terms of transient response and robustness to parameter variations.

REFERENCES


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