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Robust performance of self-scheduled LPV control of doubly-fed induction generator in wind energy conversion systems

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Keywords
<<Robustness>>, <<Modelling>>, <<Simulation>>, <<Vector control>>, <<Wind energy>>.

Abstract
This paper describes the design of a self-scheduled current controller for doubly-fed induction generators in wind energy conversion systems (WECS). The design is based on viewing the mechanical angular speed as an uncertain yet online measurable parameter and on subsuming the problem into the framework of linear parameter-varying (LPV) controller synthesis. An LPV controller is then synthesized to guarantee a bound on the worst-case energy gain for all admissible trajectories of rotor speed in the operating range. Furthermore, this study investigates the robust performance of the LPV controller with respect to other bounded machine parameter variations and the impact of the stator voltage dips on the robustness of the control system. Two closed loop simulation models, one with a conventional control scheme and the other with an LPV control scheme, are developed for the control of the electrical torque and the power factor on the rotor side in order to compare the performance of the control systems. Some simulation results are given to demonstrate the performance and robustness of the control algorithm.

1 Introduction
Doubly fed induction machines (DFIMs) are recently considered to be an attractive solution for wind energy conversion systems (WECS) since they can be controlled efficiently in a wide speed-variable range. Suitable control strategies can be used to optimize the power converted from wind energy into electrical energy both from the stator and the rotor. Control actuation is performed at the rotor side through slip rings. This allows a reduction of the size of the power converter and, hence, of the cost of the overall system, especially at high-power levels.

In the regular configuration of variable speed wind turbines, the stator of DFIM is directly connected to the grid and the rotor is connected with two converters, one in the grid side, the so-called Grid Side Converter (GSC), and one in the rotor side, the so-called Rotor Side Converter (RSC), coupled by a DC-voltage link as shown in Figure 1.

1.1 Modelling and control of doubly-fed induction machine
In this paper, a dq reference frame that has the d axis coinciding with the grid voltage vector is adopted. In this reference frame, the DFIM equations can be written as

\[ \dot{x}_r = A_r(\omega)x_r + B_r v_s + B_r v_r \]  
\[ y_r = C_r x_r \]
where $x_r = (i_{rd} \ i_{rq} \ \Psi_{sd} \ \Psi_{sq})^T$; $v_s = (v_{sd} \ v_{sq})^T$; $v_r = (v_{rd} \ v_{rq})^T$; $y_r = (i_{rd} \ i_{rq})^T$;

$$A_{rc}(\omega) = \begin{pmatrix}
    -\left(\frac{a+1}{T_r} + \frac{a}{T_s}\right) & \omega_n - \omega & \frac{a}{L_m T_r} & -\frac{a}{L_m} \\
    -\left(\frac{a+1}{T_r} + \frac{a}{T_s}\right) & 0 & \frac{\sigma}{L_m} & \frac{a}{L_m T_s} \\
    \frac{\sigma}{L_m T_r} & 0 & 0 & \omega_s \\
    0 & 0 & -\frac{1}{T_s} & \omega_r
\end{pmatrix};$$

$$B_s = \begin{pmatrix}
    \frac{1-\sigma}{\sigma L_m} \\
    0 \\
    -\frac{1-\sigma}{\sigma L_m} \\
    1
\end{pmatrix}; \quad B_r = \begin{pmatrix}
    \frac{1-\sigma}{\sigma L_r} \\
    0 \\
    1 \\
    0
\end{pmatrix}; \quad C_{rc} = \begin{pmatrix}
    1 & 0 & 0 & 0 \\
    0 & 1 & 0 & 0
\end{pmatrix};$$

$v_{sd}, v_{sq}, v_{rd}, v_{rq}, i_{sd}, i_{sq}, i_{rd}, i_{rq}$ are voltage and current components of the stator and rotor respectively; $\Psi_{sd}, \Psi_{sq}$ are stator flux components; $L_s, L_r$ are stator and rotor inductances; $L_m$ is mutual inductance; $R_s, R_r$ are stator and rotor resistances; $\sigma = 1 - \frac{L_r^2}{L_s L_r}$ is the total linkage coefficient; $a = \frac{1-\sigma}{\sigma}$, $T_s = \frac{L_s}{R_s}$ and $T_r = \frac{L_r}{R_r}$ denote the time constants of stator and rotor; $\omega = \omega_s - \omega_r$ is the mechanical angular velocity of the rotor; $\omega_s$ is electrical angular velocity of stator (or grid); and $\omega_r$ is electrical angular velocity of rotor.

In the literature, the classical approach to DFIM vector control [1] allows one to achieve decoupled control of active and reactive power in both generator and motor operations. The control structure of DFIM including PI current controllers is described in [2, 3, 4, 5]. In some cases, the cross coupling term in the rotor equations that includes the mechanical angular speed is eliminated by adding a feed-forward term to the output of the $q$-axis controller [3, 6]. In these cases the difficulties of the nonlinear dynamics of the doubly-fed induction generator (DFIG) are not taken into account, i.e., the model of the machine is linearized and it is assumed that both the machine parameters required by the control algorithm and the grid voltage are precisely known. Clearly, such controller designs might result in a closed-loop behavior that is highly sensitive to a change in operating conditions and/or parameters.

In order to improve the system performance against changes in the machine parameters and exogenous inputs, an $H_{\infty}$ control approach for an induction generator in windmill power system is proposed in [7] and for induction motor control in [8]. Recently, the LPV current control approach, which takes the parameter variations into account directly in the control design, is applied for an induction motor in [9, 10]. In [10], the electrical angular rotor speed and the estimated magnetizing current are considered to be varying parameters. The control objective is to track references for the magnetizing current and the angular electrical rotor speed. A quasi-LPV approach is applied to the design of a stator current controller and a speed controller. In [9], the same method is employed for the inner current control loop, and the LPV controller synthesis is extended to a discrete time setting.

Our paper presents an alternative control strategy for DFIMs. The control objective is to track references for the electrical torque and the power factor. The mechanical angular speed $\omega$ in (1) is considered as a time-varying parameter. This particular choice is motivated by the fact that $\omega$, which causes the system to be nonlinear, can be measured online. Actually, its value varies by $\pm30\%$ around the synchronous
speed $\omega_s$. Therefore, with $-1 \leq \delta_\omega \leq 1$ and $p_\omega = 0.3$, the mechanical angular speed can be expressed as $\omega = \omega_s(1 + p_\omega \delta_\omega)$. Thus (1) now becomes affinely parameter dependent and can be rewritten as

$$\dot{x}_c = (A_{rs} + \delta_\omega A_{re0})x_r + B_s y_s + B_r v_r$$

(3)

where $A_{rs}, A_{re0}$ are time-invariant matrices defined by

$$A_{rs} = \begin{pmatrix} -\frac{a+1}{r_c} + \frac{a}{r_s} & 0 & \frac{a}{L_m r_c} & -\frac{a_0}{L_m} \\ 0 & -\frac{a+1}{r_c} + \frac{a}{r_s} & -\frac{1}{r_c} & \omega_s \\ \frac{L_m}{r_c} & 0 & \frac{a}{L_m} & -\omega_s - \frac{1}{r_c} \\
 0 & \frac{L_m}{r_c} & 0 & 0 \\
 \end{pmatrix}; \quad A_{re0} = \begin{pmatrix} 0 & -\omega_s p_\omega & 0 & -\frac{a_0 b_0}{L_m} \\ \omega_s p_\omega & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 \\
 \end{pmatrix}.$$
2 LPV controller synthesis for affinely parameter-dependent systems

2.1 $L_2$-gain performance

Let us consider an LPV system that is described as

\[
\begin{pmatrix}
\dot{x}(t) \\
x_p(t) \\
y(t)
\end{pmatrix} = \begin{pmatrix}
A(\delta(t)) & B_p(\delta(t)) & B(\delta(t)) \\
C_p(\delta(t)) & D_p(\delta(t)) & E(\delta(t)) \\
C(\delta(t)) & D(\delta(t)) & 0
\end{pmatrix} \begin{pmatrix}
x(t) \\
w_p(t) \\
u(t)
\end{pmatrix}
\]

(6)

where the matrices in (6) are affine functions of the parameter vector that varies in the polytope $\delta_c$ with vertices $\delta^1, ..., \delta^k$, that is

\[
\delta(t) \in \delta_c = \text{conv} \{\delta^1, ..., \delta^k\} \triangleq \left\{ \sum_{j=1}^k \lambda_j \delta^j, \quad \lambda_j \geq 0, \quad \sum_{j=1}^k \lambda_j = 1 \right\}
\]

(7)

The optimization problem is to search for an LPV controller that is defined with affine functions as

\[
\begin{pmatrix}
\dot{x}(t) \\
x_p(t) \\
y(t)
\end{pmatrix} = \begin{pmatrix}
A_c(\delta(t)) & B_c(\delta(t)) \\
C_c(\delta(t)) & D_c(\delta(t))
\end{pmatrix} \begin{pmatrix}
x_c(t) \\
y(t)
\end{pmatrix}
\]

(8)

such that the closed-loop system of (6) and (8)

\[
\begin{pmatrix}
\dot{z}(t) \\
z_p(t)
\end{pmatrix} = \begin{pmatrix}
A(\delta(t)) & B(\delta(t)) \\
C(\delta(t)) & D(\delta(t))
\end{pmatrix} \begin{pmatrix}
\dot{z}(t) \\
z_p(t)
\end{pmatrix}
\]

(9)

is internally stable and the $L_2$-norm of $w_p(t) \rightarrow z_p(t)$ is bounded by a given number $\gamma > 0$ for all possible parameter trajectories $\delta: [0, \infty) \rightarrow \delta_c$.

Note that the matrices $A(\cdot), B(\cdot), C(\cdot), D(\cdot)$ in (9) are given as

\[
\begin{pmatrix}
A(\delta(t)) & B(\delta(t)) \\
C(\delta(t)) & D(\delta(t))
\end{pmatrix} = \begin{pmatrix}
A(\delta(t)) + B(\delta(t))D_c(\delta(t)) & B(\delta(t))C_c(\delta(t)) + B(\delta(t))D_c(\delta(t))F(\delta(t)) \\
C_p(\delta(t)) + E(\delta(t))D_c(\delta(t)) & C(\delta(t)) + E(\delta(t))D_c(\delta(t))F(\delta(t))
\end{pmatrix}
\]

The characterization of robust stability and performance for the closed-loop system (9) is provided by the following theorem:

**Theorem 2.1** If there exists a constant matrix $X > 0$ for which

\[
\begin{pmatrix}
I & 0 \\
0 & I
\end{pmatrix}^T \begin{pmatrix}
0 & X & 0 \\
X & 0 & 0 \\
0 & 0 & -\gamma I
\end{pmatrix} \begin{pmatrix}
I & 0 \\
0 & I
\end{pmatrix} < 0
\]

holds for all $\delta \in \delta_c$,

(10)

then the system (9) is uniformly exponentially stable and the $L_2$ gain from $w_p$ to $z_p$ is bounded by $\gamma$. \( \Box \)

**Proof:** See [11].

Since the system (6) is affinely parameter-dependent with respect to the time-varying parameter $\delta(t)$ in (7), the state-space matrices of (6) range in the polytope defined as follows [12]:

\[
\begin{pmatrix}
A(\delta(t)) & B_p(\delta(t)) & B(\delta(t)) \\
C_p(\delta(t)) & D_p(\delta(t)) & E(\delta(t)) \\
C(\delta(t)) & D(\delta(t)) & 0
\end{pmatrix} \in \text{conv} \left\{ \begin{pmatrix}
A^j & B^j_p \\
C^j_p & D^j_p \\
C^j & D^j
\end{pmatrix} : j = 1, ..., k \right\}
\]

\[
\triangleq \left\{ \begin{pmatrix}
A(\delta^j) & B_p(\delta^j) & B(\delta^j) \\
C_p(\delta^j) & D_p(\delta^j) & E(\delta^j) \\
C(\delta^j) & D(\delta^j) & 0
\end{pmatrix} : j = 1, ..., k \right\}
\]
If \( \begin{bmatrix} B \\ E \end{bmatrix} \) and \((C \ F)\) are parameter independent, the describing matrices for the closed-loop system (9) are also affine in the parameter, i.e.

\[
\begin{pmatrix} A(\delta(t)) \\ C(\delta(t)) \\ D(\delta(t)) \end{pmatrix} \in \text{conv} \left\{ \begin{pmatrix} A^j \\ C^j \\ D^j \end{pmatrix} = \begin{pmatrix} \Delta \delta(t) \\ \Delta C(t) \\ \Delta D(t) \end{pmatrix}, j = 1, \ldots, k \right\}.
\]

This implies for the synthesis inequalities (10) that we can replace the search over the polytope \( \delta_\mathcal{E} \) without loss of generality by the search over the extreme points \( \delta^1, \ldots, \delta^k \) of this set. Consequently, condition (10) can be reduced to a finite set of Linear Matrix Inequalities (LMIs) since it is equivalent to

\[
\begin{pmatrix} I \\ \Delta A(\delta^i) \\ \Delta C(\delta^i) \\ \Delta D(\delta^i) \end{pmatrix} = \begin{pmatrix} I \\ X \end{pmatrix} \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & -\gamma l & 0 \\ 0 & 0 & 0 & \gamma \end{pmatrix} \begin{pmatrix} I \\ \Delta C(\delta^i) \\ \Delta D(\delta^i) \end{pmatrix} \prec 0 \quad \text{for all } j = 1, \ldots, k.
\]

### 2.2 LPV controller synthesis

Elimination of the controller parameters in (11) leads to the following LMI conditions that guarantee the existence of a polytopic LPV controller [12, 13, 14]:

\[
\Phi^T_x \begin{pmatrix} \gamma X A(\bar{\delta}^i) + A(\bar{\delta}^i)^T X & XB_p(\bar{\delta}^i) & C_p(\bar{\delta}^i)^T Y \\ B_p(\bar{\delta}^i)^T X & -\gamma l & D_p(\bar{\delta}^i)^T \\ C_p(\bar{\delta}^i) & D_p(\bar{\delta}^i) & -\gamma l \end{pmatrix} \Phi_x \prec 0
\]

\[
\Phi^T_y \begin{pmatrix} \gamma X A(\bar{\delta}^i) + A(\bar{\delta}^i)^T Y & B_p(\bar{\delta}^i) & Y C_p(\bar{\delta}^i)^T Y \\ B_p(\bar{\delta}^i)^T Y & -\gamma l & D_p(\bar{\delta}^i)^T Y \\ C_p(\bar{\delta}^i) & D_p(\bar{\delta}^i) & -\gamma l \end{pmatrix} \Phi_y \prec 0
\]

for \( j = 1, \ldots, k \), where \( \Phi_x \) and \( \Phi_y \) form bases for \( \text{ker} \begin{pmatrix} B^T & 0 \\ C^T & F \end{pmatrix} \) and \( \text{ker} \begin{pmatrix} C^T & F \end{pmatrix} \) respectively.

After obtaining \( X \) and \( Y \) over the constraint LMIs (12)-(14), the controller parameters at each \( \bar{\delta}^j \) can be reconstructed by using the projection lemma [11]. Then a vertex controller \( K_j \) is any solution that satisfies (11) for the corresponding index \( j \).

Finally, the controller is implemented as follows: at time \( t \) we determine coefficients \( \lambda_1(t), \ldots, \lambda_k(t) \) which represent \( \bar{\delta}(t) \) according to (7), and we use

\[
K(t) = \sum_{j=1}^{k} \lambda_j(t)K_j = \sum_{j=1}^{k} \lambda_j(t) \begin{pmatrix} A_{kj} \\ C_{kj} \\ D_{kj} \end{pmatrix}
\]

as the system matrix for simulation.

### 3 Gain-scheduling design for the rotor side current controller

#### 3.1 System representation when the rotor angular speed \( \omega \) is treated as an uncertainty

With \( \Delta_\omega = \begin{pmatrix} \delta_\omega & 0 \\ 0 & \delta_\omega \end{pmatrix} \), the system (3) can be rewritten as follows:

\[
\begin{pmatrix} \dot{x}_r \\ y_r \\ z_{\omega_0} \end{pmatrix} = G_{rc} \begin{pmatrix} \dot{x}_r \\ v_z \\ v_r \\ w_{\omega_0} \end{pmatrix}, \quad w_\omega = \Delta_\omega z_{\omega_0}
\]

where

\[
G_{rc} = \begin{pmatrix} A_{rs} & B_r & B_{r_0} \\ C_{rc} & 0 & 0 \\ C_{r_0} & 0 & 0 \end{pmatrix};
\]

\[
C_{r_0} = \begin{pmatrix} 0 & -\omega_p p_0 & 0 \\ 0 & \omega_p p_0 & -\frac{\omega_p p_0}{L_m} \end{pmatrix}; \quad B_{r_0} = \begin{pmatrix} 1 & 0 & 0 \end{pmatrix}^T; \quad y_r = (i_r \quad i_q)^T.
\]
3.2 $H_{\infty}$ control of the LPV system

A mixed sensitivity T/S loop shaping $H_{\infty}$ optimization is proposed for the rotor current control loop (see Figure 2). The external control inputs $w_{rc}$ consist of stator voltages and reference rotor currents $w_{rc} = \left( v_{sd} \ v_{sq} \ \text{ref} \ i_{rd}^\text{ref} \ i_{rq}^\text{ref} \right)^T$. The controller outputs are $v_r = \left( v_{rd} \ v_{rq} \right)^T$. The controller inputs or tracking errors are $e_r = \left( e_{rd} \ e_{rq} \right)^T = \left( i^\text{ref}_{rd} - i_{rd} \ i^\text{ref}_{rq} - i_{rq} \right)^T$. The measured outputs are $y_r = \left( i_{rd} \ i_{rq} \right)^T$. The sensitivity function is $S_{rc} = (I + G_{rc}K_{rc})^{-1}$ and the complementary sensitivity function is $T_{rc} = I - S_{rc}$.

The interconnection of the system is shown in Figure 3. The weighting function $W_{rs} = \begin{pmatrix} W_{rtd} & 0 \\ 0 & W_{rtq} \end{pmatrix}$ is a first-order low-pass filter used to shape the sensitivity for tracking. The weighting function $W_{rt} = \begin{pmatrix} W_{rtd} & 0 \\ 0 & W_{rtq} \end{pmatrix}$ is a first-order high-pass filter used to shape the complementary sensitivity function to guarantee the robustness against high frequency un-modelled dynamics.

![Figure 3: The interconnection of the system](image)

The standard $H_{\infty}$ control problem is to find a stabilizing LTI controller $K_{rc}(\omega)$ at fixed frozen values of $\omega$ such that the $H_{\infty}$-norm of the channel $w_{rc} \rightarrow z_{rc}$ is smaller than a given number $\gamma$.

3.3 Synthesis of gain-scheduled current controller

The gain-scheduled controller synthesis is similar to the classical $H_{\infty}$ synthesis, but both the plant and the controller are now LPV systems. The optimization problem is to find a stabilizing controller $K_{rc}(\omega)$ such that the $L_2$-gain of the channel $w_{rc} \rightarrow z_{rc}$ is smaller than $\gamma$ for all trajectories of $\omega(t) \in [\omega_{\min}, \omega_{\max}] = [(1 - p_\omega)\omega_s, (1 + p_\omega)\omega_s]$.

The synthesis LMIs (12)-(14) are solved by using the LMI Control Toolbox [15]. If a solution $(X,Y)$ is given, the vertex controllers $K_j$ are constructed as solutions of (11). Now we are ready to compute the polytopic LPV controller by measuring values of $\omega$ online and by getting a vertex decomposition as expressed in (7). Then the state-space matrices describing the LPV controller are also given online by the interpolation

$$
\begin{pmatrix}
A_{Krc}(t) & B_{Krc}(t) \\
C_{Krc}(t) & D_{Krc}(t)
\end{pmatrix} = \frac{\delta_{w}^{\max} - p(t)}{\delta_{w}^{\max} - \delta_{w}^{\min}} \begin{pmatrix}
A_{Krc1} & B_{Krc1} \\
C_{Krc1} & D_{Krc1}
\end{pmatrix} + \frac{p(t) - \delta_{w}^{\min}}{\delta_{w}^{\max} - \delta_{w}^{\min}} \begin{pmatrix}
A_{Krc2} & B_{Krc2} \\
C_{Krc2} & D_{Krc2}
\end{pmatrix}
$$

where $p(t) = \frac{\omega(t) - \omega_{s}}{\omega_{s} p_{w}}$. 


4 Simulations

4.1 Performance of the system with dynamic wind

When the wind speed is less than the rated wind speed, the objective is to maximize the captured energy by adjusting the rotor speed in order to operate the turbine along the maximum power curve or, in other words, to keep the tip-speed ratio optimal. As the wind speed increases, the generator is allowed to accelerate until it reaches its rated speed. The pitch angle values in the power optimization region are all found close to zero for the given wind turbine. When the wind speed is higher than the rated wind speed, the input power from the generator is maintained at its rated power by increasing the pitch angle to keep the electrical power output at rated power. Hence, the pitch angle control is mostly used above rated wind speed to prevent overload of the generator power. The performances of the controlled system with the LPV controller under wind speed changes is shown in figure 4.

![Graphs showing wind speed, rotor electrical angular speed, torque, power factor, and rotor currents for different times.]

Figure 4: The performance of the system with dynamic wind

4.2 Performance comparison of deadbeat and LPV controllers

In this study two complete simulation models, one based on a conventional control scheme that is called dead-beat control (similar to that in [16]) and the other based on the described LPV framework are developed for the control of the electrical torque and the power factor on the rotor side in order to compare the performance of the closed loop systems.

4.2.1 Performance of the system with step changes

Figure 5 shows the performance of the system tested with step changes of the set-point values of $i_{rd}$ and $i_{rg}$. From 0 to 0.05s the stator windings are open. The stator voltages are regulated so as to be equivalent to grid voltages in amplitude and phase. After that, the stator windings are connected to the grid and the transient period appears to be about 0.2s. At the beginning of the transient period, the rotor currents may fluctuate strongly due to sudden increase of stator flux and currents. The simulation results show that the variations of the rotor currents during the transient period with the LPV controller are much smaller than those for the deadbeat controller.
4.2.2 Robustness of the control system against other parameter changes

By varying the values of $L_s$ we see form Figure 6 that the system with dead-beat controller stays stable if the $L_s$ deviates by $[97.9\%, 101.12\%]$ of its nominal values, while with the self-scheduled controller this range is $[95.8\%, 150\%]$.

Similarly, by varying the values of $L_r$ we see from Figure 7 that the system with dead-beat controller stays stable if the $L_r$ deviates by $[97.84\%, 125\%]$ from its nominal value, while with the self-scheduled controller this range is $[95.68\%, 150\%]$.

4.2.3 The impact of stator voltage dips on the robustness of the control system

When the grid undergoes a fault, the sag in the grid voltage will result in an increase of the current in the stator windings of the DFIM. Because of the magnetic coupling between stator and rotor, this current will also flow into the rotor circuit and the power converter leading to the destruction of the converter if nothing is done to protect it. On the other hand, the study in [5] shows that the dynamics of the DFIM has
poorly damped poles in the transfer function of an LTI model of the machine. This will cause oscillations in the flux if the DFIM is affected by grid disturbances. After such disturbances, an increased rotor voltage will be needed to control the rotor currents. When this required voltage exceeds the voltage limit of the converter, it is not possible any longer to control the current as desired [17]. Therefore, the control system should maintain operation and reduce oscillations as much as possible during grid voltage faults. A comparison of the performance between the dead-beat control and the scheduled control systems in

the presence of grid voltage faults is presented in Figure 8. The grid voltage was dropped down to 25% of the normal voltage. This phenomenon occurred in a period of 300 msec before it recovered to its rated value. After that, the grid voltage once again was dropped to 50% of the rated voltage during 200 msec. The graphs show that the oscillations of torque and currents in the case of self-scheduled control were remarkably dampened at the grid voltage fault time.

5 Conclusion

The self-scheduled LPV control method has been applied to design the rotor side current controller for the DFIM in a variable speed wind turbine system, where the online measurable rotor mechanical
angular speed is considered as the time varying parameter. Hence the designed controller maintains the performance requirements for all trajectories of rotor speed over its variation range in a systematic way. A classical control approach, the so-called dead-beat controller, is also developed for the control of the electrical torque and the power factor on the rotor side. Under disadvantageous variations of machine parameters and for grid voltage dips, simulation results show that the designed LPV controller is far more robust than the dead-beat controller. Oscillations in the stator and rotor currents are considerably reduced during the grid voltage faults, and the closed loop system recovers from the faults much faster than in the conventional case. Hence, the new control scheme improves the performance of the closed-loop DFIM considerably.

Experiments on a real-time laboratory set up are currently performed, and a more thorough analysis of the performance improvements are being investigated.

References


