Neuro-Second Order Sliding Mode Control of a DFIG Supplied by a Two-Level NSVM Inverter for Wind Turbine System

H. Benbouhenni*(C.A.), Z. Boudjema** and A. Belaidi*

Abstract: This paper applied second order sliding mode control (SOSMC) strategy using artificial neural network (ANN) on the rotor side converter of a 1.5 MW doubly fed induction generator (DFIG) integrated in a wind turbine system. In this work, the converter is controlled by a neural space vector modulation (NSVM) technique in order to reduce powers ripples and total harmonic distortion (THD) of stator current. The validity of the proposed control technique applied on the DFIG is verified by Matlab/Simulink. The active power, reactive power, torque and stator current are determined and compared with conventional control method. Simulation results presented in this paper shown that the proposed control scheme reduces the THD value and powers ripples compared to traditional control under various operating conditions.

Keywords: Doubly Fed Induction Generator, Artificial Neural Network, Space Vector Modulation, Neural Space Vector Modulation, Total Harmonic Distortion, Second Order Sliding Mode Control.

1 Introduction

WIND energy is becoming one of the most important renewable power sources used nowadays. Recently, power converter command has mostly been studied and developed for wind power conversion system (WECS) integration in the electrical grid. The use of energy electronic converters allows variable speed operation of the WECS to extracts maximum power from the turbine [1].

The DFIG has been widely used for large scale wind generation systems [2]. However, the stator winding directly connected to the grid and the rotor winding connected to the grid through a variable frequency converter. On the other hand, various control methods has been proposed in the literature, for studying the behavior of the DFIG based wind conversion system during normal operation. The most important are: vector control (VC) [3-5], direct torque control (DTC) [6-8], direct power control (DPC) [9-11], intelligent control (IC) [12, 13], robust control (RC) [14, 15], backstepping control (BC) [16]. In literature [17], VC command is the most popular technique used in the DFIG based wind energy conversion system. However, this control scheme gives more powers ripples and big THD of stator current.

For robust and high performance VC command, a sliding mode controller (SMC) was studied in the literature [18-20]. However, the SMC technique was proposed by Utkin in 1977 [21]. This control method based on the theory of variable structure systems has been extensively employed for nonlinear systems [22]. Like every control techniques have some advantages and disadvantages, SMC control has too. Some of the advantages are presented in [23]. The basic disadvantages of the SMC method using sat or sign function are the chattering phenomenon. Many papers have been proposed to overcome this problem. In [24], second order sliding mode controller was designed to command the active and reactive powers. SMC and fuzzy controller are combined to command the
In [26], a hybrid control based on fuzzy logic controller and a second order sliding mode was designed to command the DFIG based WECS. In this paper, we propose a new second order sliding mode control (SOSMC) based on ANN. However, in the aim to augment the performances of the DFIG control we propose in this paper a hybrid method based on SOSMC and ANN, the results control scheme is known as (NSOSMC).

Traditionally, the space vector modulation (SVM) technique is widely used for control of multilevel inverters of AC machines. The principle of SVM method is detailed in [27-29]. On the other hand, this technique gives minimum THD and powers ripples compared with pulse width modulation (PWM) technique. In this work, we propose a new SVM technique based on ANN controller named (NSVM). This technique gives more and more minimum of THD value, simple scheme and easy to implement compared with conventional SVM method.

In this paper, we apply the NSOSMC technique on the DFIG based wind power conversion system using NSVM strategy compared to the conventional SVM inverter.

2 NSVM Technique

In this section, we propose an SVM technique based on ANN controller. However, the ANN consists of several cascaded layers of neurons with sigmoid activation functions [30]. On the other hand, The ANN controller has many models. The ANN controller contains three layers: output layers, hidden layers, and input layers. Each layer is composed of several neurons [31].

Since SVM method based on the principles of space vectors and need to calculate of sector and angle [32, 33]. On the other hand, we propose a new SVM technique of two-level inverter based on calculation of maximum and minimum of three-phase voltages ($V_a, V_b, V_c$). The advantages of the proposed SVM method is not needed to calculate the sector and angle, easy to implement and gives a strong performance for the real-time feedback command compared with PWM technique. Fig. 1 shows the principle of the SVM technique of two-level inverter.

The principle of the NSVM is similar to traditional SVM. However, the hysteresis controllers are replaced by neural controllers and this method based on neural classification has the advantage of simplicity and easy implementation. On the other hand, the NSVM gives more and more minimum THD of stator current compared to traditional SVM technique. The structure of SVM based on ANN controller is shown in Fig. 2.

The convergence of the network in summer obtained by using the value of the parameters grouped in Table 1. The structure of the ANN controller used to perform the SVM technique is an ANN with one linear input node, 12 neurons in the hidden layer, and one neuron in the output layer. As shown in Fig. 3 the ANN controller is composed of two layers, Layer 1 (Fig. 4) and Layer 2 (Fig. 5).

3 Model of Turbine

The wind turbine input energy usually is:

$$P_i = 0.5 \rho S_n V_w^3$$  (1)

The output mechanical power of wind turbine is:

$$P_m = C_p P_i = 0.5 \rho C_p S_n V_w^3$$  (2)
### Table 1 Parameters of the LM for hysteresis controllers.

<table>
<thead>
<tr>
<th>Parameters of the LM</th>
<th>Values</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of hidden layer</td>
<td>12</td>
</tr>
<tr>
<td>TrainParam.Lr</td>
<td>0.002</td>
</tr>
<tr>
<td>TrainParam.show</td>
<td>50</td>
</tr>
<tr>
<td>TrainParam.eposh</td>
<td>1000</td>
</tr>
<tr>
<td>Coeff of acceleration of convergence (mc)</td>
<td>0.9</td>
</tr>
<tr>
<td>TrainParam.goal</td>
<td>0</td>
</tr>
<tr>
<td>TrainParam.mu</td>
<td>0.9</td>
</tr>
<tr>
<td>Functions of activation</td>
<td>Tensing, Purling, gensim</td>
</tr>
</tbody>
</table>

### Fig. 3 Structure of ANN hysteresis comparators.

### Fig. 4 Layer 1.

### Fig. 5 Layer 2.

The ratio of the tip speed of the turbine blades to wind speed is:

\[ \lambda = \frac{R \Omega}{v} \]  

(3)

\[ C_p \] can be described as [34, 35]:

\[ C_p(\beta, \lambda) = C_1 \frac{(\frac{C_3}{\lambda_1} - C_2 \beta - C_4)}{} \exp\left(-\frac{C_5}{\lambda_1}\right) + C_6 \lambda \]  

(4)

\[ 1 = \frac{1}{\lambda} + 0.08 \beta - 0.035 \beta^3 + 1 \]  

(5)

The torque produced by the turbine is expressed in the following way:

\[ T_r = \frac{P}{\Omega} = 0.5 \rho \pi R V^2 \psi_r C_r \]  

(6)

\[ C_r = \frac{C_p}{\lambda} \]  

(7)

where: \( C_p \) represents the wind turbine power conversion efficiency, \( \rho \) is air density, \( S_w \) is wind turbine blades swept area in the wind, \( V \) is wind speed, \( R \) is blade radius, \( \Omega \) is angular speed of the turbine, \( C_i \) is the torque coefficient. \( C_1 = 0.5176, \) \( C_2 = 116, \) \( C_3 = 0.4, \) \( C_4 = 5, \) \( C_5 = 21, \) \( C_6 = 0.0068. \)

### 4 Model of DFIG

The traditional electrical equations of the DFIG in the Park frame are written as follows [36, 37]:

\[ V_{dr} = R_d I_{dr} + \frac{d}{dt} \psi_{qr} - \alpha_3 \psi_{qr} \]  

(8)

\[ V_{qr} = R_q I_{qr} + \frac{d}{dt} \psi_{dr} \]  

\[ V_{dr} = R_d I_{dr} + \frac{d}{dt} \psi_{dr} - \alpha_1 \psi_{dr} \]  

\[ V_{qr} = R_q I_{qr} + \frac{d}{dt} \psi_{qr} + \alpha_2 \psi_{qr} \]  

The rotor and stator flux can be expressed as:

\[ \psi_{dr} = L_r I_{dr} + M I_{qr} \]  

(9)

\[ \psi_{qr} = L_q I_{qr} + M I_{dr} \]  

\[ \psi_{dr} = L_r I_{dr} + M I_{qr} \]  

\[ \psi_{qr} = L_q I_{qr} + M I_{dr} \]  

The reactive and active powers at the stator can be expressed as:

\[ P = \frac{3}{2} \left( V_{dr} I_{dr} + V_{qr} I_{qr} \right) \]  

(10)

\[ Q = \frac{3}{2} \left( V_{dr} I_{dr} \right) \]  

The electromagnetic torque is expressed as:

\[ T_e = pM \left( I_{dr} I_{qr} - I_{qr} I_{dr} \right) \]  

(11)

\[ T_e = T_e + J \frac{d}{dt} \Omega + f \Omega \]  

(12)

where \( V_{dr}, V_{qr}, V_{ds}, \) and \( V_{ds}, \) are the two-phase rotor and stator voltages, \( I_{dr}, I_{qr}, I_{ds}, \) and \( I_{ds}, \) are the two-phase rotor and stator currents, \( \psi_{dr}, \psi_{qr}, \psi_{ds}, \) and \( \psi_{ds}, \) are the two-
phase rotor and stator fluxes, $L_r$, $L_s$ and $M$ are respectively the inductance own rotor, stator, and the mutual inductance between two coils, $R$, and $R_s$ are respectively the resistances of the stator and rotor windings, $\omega_r$ is the electrical pulsation of the stator and $\omega_c$ is the rotor one. $T_L$ is the load torque, $T_e$ is the electromagnetic torque, $\Omega$ is the mechanical rotor speed, $J$ is the inertia, and $f$ is the viscous friction coefficient and $p$ is the number of pole pairs. $P_s$ is the stator active power, $Q_s$ is the stator reactive power.

5 Robust Control Strategies of DFIG

In this section, comparison of DFIG performances using two nonlinear controllers: second order sliding mode, neuro-second order sliding mode.

5.1 Second Order Sliding Mode Control

Sliding mode control is an effective robust control technique for unmatched perturbations [38]. The principle of this technique is detailed in [39]. The major disadvantage of the SMC controller is the chattering phenomenon. To eliminate the chattering phenomenon, a second order sliding mode is proposed in [40, 41], and widely applied to various systems [42-44]. In [45], second order sliding mode controller is developed to control DFIM machine. In [46], second order sliding mode control is developed to control reactive and active powers of DFIG.

The second order sliding mode control (SOSMC) does not need accurate mathematical models like classical controllers. On the other hand, we choose the error between the reference stator energies and measured as second order sliding mode surfaces, so we can write the following expression:

$$S_p = \left[ \frac{P_{ref} - P_s}{Q_{ref} - Q_s} \right]$$

We derived the above errors, we obtain

$$\dot{S_p} = \left[ \frac{\dot{P}_{ref} - \dot{P}_s}{\dot{Q}_{ref} - \dot{Q}_s} \right]$$

Then we will have

$$\dot{S_p} = \left[ \frac{\dot{P}_{ref} - \frac{\alpha}{\sigma L_c} (V_{ao} - R, I_{ao} - g, \sigma L, I_{ao} - g, \frac{MV_s}{L_c})}{\dot{Q}_{ref} - \frac{\alpha}{\sigma L_c} (V_{ao} - R, I_{ao} + g, \sigma L, I_{ao})} \right]$$

where $\alpha = -V_o M/L_o$. If we define the $A_1$ and $A_2$ functions as follows.

$$A_1 = \frac{P_{ref} - \frac{\alpha}{\sigma L_c} (V_{ao} - R, I_{ao} - g, \sigma L, I_{ao} - g, \frac{MV_s}{L_c})}{\dot{Q}_{ref} - \frac{\alpha}{\sigma L_c} (V_{ao} - R, I_{ao} + g, \sigma L, I_{ao})}$$

Thus we have

$$\dot{S_p} = \left[ \frac{\alpha}{\sigma L_c} V_{ao} + A_1 \right]$$

On deriving the relationship of Eq. (17) yields:

$$\dot{S_q} = \left[ \frac{\alpha}{\sigma L_c} V_{ao} + A_2 \right]$$

The SOSMC proposed based on the super twisting algorithm known (ST) which is introduced by Levant.

$$V_{ao} = u_1 + u_2$$

Then it follows that

$$\begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} -\lambda_1 \text{sgn}(S_q) \\ -\delta_2 \text{sgn}^0.5(S_q) \end{bmatrix}$$

And

$$V_{ao} = w_1 + w_2$$

Including

$$\begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} -\lambda_2 \text{sgn}(S_p) \\ -\delta_2 \text{sgn}^0.5(S_p) \end{bmatrix}$$

To ensure the convergence of regulators in the infinity of time constants and are chosen to satisfy the following inequality

$$\begin{bmatrix} \lambda_1 & \mu_1 \\ \delta_2 & \sigma \end{bmatrix} \geq \begin{bmatrix} 4\mu_1 (\lambda_1 + \mu_1) \\ (\lambda_1 - \mu_1)(L_c, \sigma) \end{bmatrix}$$

$$|A_1| < \mu_1 \quad \forall i = 1, 2$$

Fig. 6 shows the block diagram of second order sliding mode control strategy.
5.2 Neuro-Second Order Sliding Mode Control

The application of artificial neural networks attracts the attention of many scientists from all over the world [47]. This intelligent technique have many advantages, it is simple architecture, inexact input data, the possibility of approximating non-linear function, insensitivity to the distortion of the network, easy of training and generalization [48]. The neuro-second order sliding mode control (NSOSMC) is similar to a traditional SOSMC. However, the switching controller term sign ($S(x)$), has been replaced by ANN controller as given by Fig. 7. On the other hand, the NSOSMC give more and more minimum powers ripples compared with traditional SOSMC method.

The structure of the proposed neural controllers was a network with one linear input node, 8 neurons in the hidden layer, and one neuron in the output layers.

Fig. 8 shows the neural network training performance of ANN controllers for active and reactive powers. Fig. 9 shows the internal structure of ANN controller for active and reactive powers. Fig. 10 shows the block diagram of the internal structure of hidden layer.

6 Simulation Results

In this section, simulations are carried out with a 1.5MW DFIG machine attached to a 398V/50Hz grid, using the Matlab/Simulink. Parameters of the DFIG are given in Table 2. Two control strategies, SOSMC-SVM and NSOSMC-NSVM, are simulated and compared regarding reference tracking, powers ripples, stator current harmonics distortion, and robustness against DFIG parameter variations.

6.1 Reference Tracking Test

Figs. 11-12 shows the harmonic spectrums of one phase stator current of the DFIG obtained using Fast Fourier Transform (FFT) method for SOSMC-SVM and NSOSMC-NSVM one respectively. Table 3 shows the comparative analysis of THD value. It can be clear observed that the THD is reduced for NSOSMC-NSVM control method (THD = 0.16%) when compared to SOSMC-SVM (THD = 0.40%).
Figs. 13-15 show the obtained simulation results. For the proposed command strategies, the stator reactive and active power tracks almost perfectly their references values. Moreover, the NSOSMC-NSVM control strategy reduced the powers ripples and torque ripple compared to the SOSMC-SVM control scheme (See Figs. 16-18).

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Rated Value</th>
<th>Unity</th>
</tr>
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<tbody>
<tr>
<td>Nominal power</td>
<td>1.5</td>
<td>MW</td>
</tr>
<tr>
<td>Stator voltage</td>
<td>398/690</td>
<td>V</td>
</tr>
<tr>
<td>Stator frequency</td>
<td>50</td>
<td>Hz</td>
</tr>
<tr>
<td>Number of pairs poles</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>Stator resistance</td>
<td>0.012</td>
<td>Ω</td>
</tr>
<tr>
<td>Rotor resistance</td>
<td>0.021</td>
<td>Ω</td>
</tr>
<tr>
<td>Stator inductance</td>
<td>0.0137</td>
<td>H</td>
</tr>
<tr>
<td>Rotor inductance</td>
<td>0.0136</td>
<td>H</td>
</tr>
<tr>
<td>Mutual inductance</td>
<td>0.0135</td>
<td>H</td>
</tr>
<tr>
<td>Inertia</td>
<td>1000</td>
<td>Kg.m²</td>
</tr>
<tr>
<td>Viscous friction</td>
<td>0.0024</td>
<td>Nm/s</td>
</tr>
</tbody>
</table>

Table 2 The DFIG parameters.

Table 3 Comparative analysis of THD value (tracking test).

<table>
<thead>
<tr>
<th>Stator current</th>
<th>SOSMC-SVM</th>
<th>NSOSMC-NSVM</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.40</td>
<td>0.16</td>
<td></td>
</tr>
</tbody>
</table>

Fig. 11 THD of one phase stator current for SOSMC-SVM control (reference tracking test).

Fig. 12 THD of one phase stator current for NSOSMC-NSVM control (reference tracking test).

Fig. 13 Active power (reference tracking test).

Fig. 14 Reactive power (reference tracking test).

Fig. 15 Torque (reference tracking test).

Fig. 16 Zoom in the active power (reference tracking test).
6.2 Robustness Test

In order to examine the robustness of the proposed controls schemes, the nominal value of the $R_s$ and $R_r$ is multiplied by 2, the values of inductances $L_s$, $M_r$, and $L_r$ are multiplied by 0.5. Simulation results are presented in Figs. 19-23. As it is shown by these Figures, these variations present a clear effect on the active, reactive powers, and electromagnetic torque curves and that the effect appears more important for the SOSMC-SVM control scheme compared to NSOSMC-NSVM control (see Figs. 24-26). On the other hand, the THD value of stator current in the NSOSMC-NSVM has been reduced significantly. Table 4 shows the comparative analysis of THD value. Thus it can be concluded that the NSOSMC-NSVM control scheme is more robust than the SOSMC-SVM control.

**Table 4 Comparative analysis of THD value (robustness test).**

<table>
<thead>
<tr>
<th>Stator current</th>
<th>SOSMC-SVM</th>
<th>NSOSMC-NSVM</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>0.84</td>
<td>0.31</td>
</tr>
</tbody>
</table>

**Fig. 17** Zoom in the reactive power (reference tracking test).

**Fig. 18** Zoom in the torque (reference tracking test).

**Fig. 19** THD of one phase stator current for SOSMC-SVM control (robustness test).

**Fig. 20** THD of one phase stator current for NSOSMC-NSVM (robustness test).

**Fig. 21** Active power (robustness test).

**Fig. 22** Reactive power (robustness test).
7 Conclusion

The simulation and the control of a WECS based on a DFIG connected directly to the grid by the stator and fed by a two-level converter on the rotor side has been presented in this article. Our objective was the implementation of a NSOSMC technique of stator reactive and active powers generated by the stator side of the DFIG in order to make the system insensitive with the external disturbance and the parametric variation. On the other hand, we proposed a NSVM technique to control the two-level inverter. This new modulation gives a minimum THD value and powers ripples compared with conventional modulation technique. However, the various results obtained in simulation show the proposed command (NSOSMC-NSVM) robustness to the system and load parameters disturbances. The results obtained with this control are very interesting compared to the conventional SOSMC-SVM control in eliminating of the active and reactive powers ripples, chattering phenomenon and THD value of stator current. The NSOSMC-NSVM control has the advantage of being easily implemented by a program command and simple scheme.

Appendix

A. NSOSMC Control Block with NSVM Inverter

The NSOSMC control of a DFIG based on NSVM inverter is shown in Fig. 27.
B. Block Diagram of SVM Inverter

The proposed SVM technique of a DFIG based wind turbine system is shown in Fig. 28.

C. Block Diagram of Wind Turbine

The block diagram of wind turbine system is shown in Fig. 29.

The proposed SVM technique based on following steps:
1) Calculates the minimum voltages (min ($V_a$, $V_b$, $V_c$)).
2) Calculates the maximum voltages (max ($V_a$, $V_b$, $V_c$)).
3) Add the maximum and minimum voltages (max ($V_a$, $V_b$, $V_c$) + min ($V_a$, $V_b$, $V_c$)).
4) The last step is to compare step-3 waveforms with $V_{\text{triangle}}$ and generates the pulses for that switch presents in the three phase voltage source converter circuit.

References


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