Intelligent maximum power point tracker enhanced by sliding mode control

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ABSTRACT

In solar photovoltaic systems applications, the maximum power point tracker has been involved for different purposes to support their performance. The maximum power point tracking (MPPT) works on growing the obtained electricity from the solar photovoltaic energy and consequently increases the quantity of the delivered electrical power from the photovoltaic (PV) system. Relying on this point, this paper introduces an intelligent tracker to guarantee the MPP working condition for a small size 150 W stand-alone PV system. In this study, an intelligent algorithm is proposed to have a fast and accurate tracker. Moreover, a robust sliding mode controller is inserted for improving the performance of a direct current (DC-DC) boost converter. The converter is working in a continuous conduction mode operation to enhances the MPP tracker. Simulink of MATLAB is adopted to implement the system. The results of the simulated tracker are evaluated comparatively based on the artificial neural network (ANN) algorithm with and without inserting the sliding mode (SM) controller for different light intensity trends and levels. Simulation results analyzed and confirmed the effectiveness of the proposed tracker.

Keywords:
ANN algorithm
DC-DC boost converter
MPPT
Photovoltaic PV cell
Sliding mode control SMC

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1. INTRODUCTION

Solar energy advantages which are being clean, inexpensive, and ubiquitous make solar energy a very important source of renewable energy. Photovoltaic (PV) cell is the fundamental part of the PV module, which converts the solar photovoltaic energy to an electricity as a direct current direct current (DC) voltage source. The power quantity is positively proportional to the instantaneous level of the incident light whereas it is negatively proportional with instantaneous level of PV panel temperature. The instantaneous harvested energy is converted to electrical power fluctuating based on the levels of light intensity and ambient temperature [1]–[3]. In the solar PV applications, to gain higher electrical power by solar energy conversion, many maximum power point tracking (MPPT) algorithms are explained in [4], whereas different algorithms and methods have been proposed and analyzed in [5]–[19]. Incremental conductance (IC), short circuit current, constant voltage (CV), and perturb & observe (P&O) algorithm have offered in [5]–[9] to maximize the power point tracking conditions during the work of the PV panel.

The demerits of the above MPPT algorithms are represented by low accuracy, low robustness, slow and oscillated response. The researchers are focusing on mitigating the effects of these demerits by proposing
higher performance algorithms and methods. Fuzzy logic control has presented in [10]–[16] for improving MPP tracking process in terms of better robustness and higher response. Fuzzy logic controller (FLC) with hill-climbing algorithm offered in [10] to have enhanced MPPT performance. A precise and quick response of fuzzy logic controller has been proposed through sensorless strategy in [11]. The effectiveness of the FLC controller is analyzed in [12]–[16] through studying membership effect on the stability behaviour of the MPPT tracker for different purposes PV system.

Intelligent neural network-based algorithms have presented in [17]–[26] for multidisciplinary systems. Artificial neural network (ANN) algorithm guarantees the effective, fast, and accurate MPPT response. ANN-based controllers were proposed in [17]–[23], to select a suitable duty cycle of DC-DC converter. Neural network algorithms were proposed in [24]–[26] for other applications in terms of power factor correction.

On the other side, the researchers focused on analyzing the converter performance effect on PV system effectiveness. Different types of converters, have been studies, and analyzed in [27]–[30]. To have a stable and robust response from the converter, sliding mode controller (SMC) have been proposed in many studies [31]–[35] to guarantee the robustness and accuracy in the converter performance and to have constant load voltage during the source or load variation.

This paper presents a stand-alone PV system of power size 150 W controlled by an intelligent MPP tracker supported by a robust SMC. The inserted SM controller works on having a smooth, accurate, and robust response from the involved boost converter. This paper focuses on the intelligent tracker performance after inserting the SM controller, whereas the difficulties of SM controller design, such as the chattering drawback, and will be discussed in the future study. The remaining is being as: the selected PV panel behaviour is shown in section 2. The design of ANN structure shown in section 3. Section 4 shows the design steps of the SM controller and demonstrates the proposed system diagram. Section 5 describes the system simulation and analyze the collected simulation results. Section 6 shows the conclusion points.

2. PHOTOVOLTAIC MODULE

This study selects the PV module type ELDORA 150P [36] of main specifications as shown in Table 1. The level of delivered power from the ELDORA 150P panel is positively proportional to the level of incident light, whereas the delivered power is negatively proportional to ambient temperature of the panel. The output power behaviour of the PV cell is nonlinear with the output voltage variation. The electrical representation of the smaller PV unit is shown in Figure 1.

<table>
<thead>
<tr>
<th>Table 1. Electrical parameters of ELDORA 150P panel</th>
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<tbody>
<tr>
<td>PV Panel Parameter</td>
</tr>
<tr>
<td>Maximum Power, ( P_{\text{max}} ) (W)</td>
</tr>
<tr>
<td>Maximum Voltage, ( V_{\text{max}} ) (V)</td>
</tr>
<tr>
<td>Maximum Current, ( I_{\text{max}} ) (A)</td>
</tr>
<tr>
<td>Open Circuit Voltage, ( V_{\text{oc}} ) (V)</td>
</tr>
<tr>
<td>Short Circuit Current, ( I_{\text{sc}} ) (A)</td>
</tr>
<tr>
<td>Panel Efficiency (%)</td>
</tr>
<tr>
<td>Temperature Coefficients (( T_c )) of Power (%/°C)</td>
</tr>
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</table>

Figure 1. Electrical representation of a PV unit or cell

In (1) shows \( I_{pv} \) which is the output PV current that equals the generated current \( I_{sc} \) minus diode current and minus shunt resistor \( R_{sh} \) current \( I_{sh} \). From (2), the induced current \( I_{dc} \) can be calculated. The current value is related to the solar cell area \( A \), the generation rate \( G_s \), and the electron and hole diffusion lengths \( L_{ne} \) and \( L_{np} \) respectively. From (3), diode current \( I_d \) can be calculated, and from (4), the parallel current \( I_{sh} \) of the parallel resistor \( R_{sh} \) can be calculated. \( V_{pv} \) represents the output voltage of PV cell which is calculated from (5) by considering the drop voltage across the series resistor \( R_s \). The total voltage out from the panel \( V_{pv,\text{Module}} \) and the current out from the panel \( I_{pv,\text{Module}} \) can be calculated from (6), and (7), respectively. The total output
The voltage can be determined by considering the total serially connected solar units $N_s$, while the total current can be determined by considering the number of parallel-connected branches $N_{sh}$ [9]–[13], [19]–[22]. The current and power delivered by the PV panel type ELDORA 150P are shown in Figure 2 including current curves in Figure 2(a) and power curves in Figure 2(b) at different levels of incident light; 200, 400, 600, 800, and 1000 W/m$^2$ at the normal temperature of the room 25°C. Figure 3 shows the PV panel current curves in Figure 3(a) and power curves in Figure 3(b) at many ambient temperatures 15, 25, 35, 45, and 55°C during a constant level of light intensity 1000 W/m$^2$.

\[
I_{pv} = I_{sc} - I_d - I_{sh}
\]  
\[
I_{sc} = qAG(T_n + L_p)
\]  
\[
I_d = I_0 \left( V_d e^{V_T} - 1 \right)
\]  
\[
I_p = \frac{V_d}{R_{sh}}
\]  
\[
V_{PV} = V_d - R_s I_{PV}
\]  
\[
V_{PV, Module} = N_s x V_{PV}
\]  
\[
I_{PV, Module} = N_{sh} x I_{PV}
\]  

Figure 2. The behaviour of PV module type ELDORA 150P during different levels of light intensity and constant ambient temperature 25°C (a) curves of output current and (b) curves of output power.
3. ALGORITHM OF ARTIFICIAL NEURAL NETWORK

The characteristics of maximum power point tracker can be more accurate and fast in response using feed-forward ANN algorithm [17]–[26]. The neural network algorithm estimates the accurate value of the instantaneous MPP reference voltage by evaluating the instantaneous levels of the incident light and panel ambient temperature. A generated reference voltage enforces the involved boost converter to work in a MPP condition to harvest maximum power from the incident solar energy. Figure 4(a) explains the presented algorithm design of neural network. The ANN algorithm includes one, and two input layer, hidden layers respectively, and four neurons in each hidden layer. The output layer represents the last layer of the proposed ANN. The input layer receives the instantaneous values of light intensity and ambient temperature, whereas the output layer produces the instantaneous value of reference voltage after the processing of the hidden layer. Figure 4(b) shows the neuron structure, in which there are weights of each input to the neuron: $X_{in1}$, $X_{in2}$, $X_{in3}$, and $X_{in4}$. After weighting all neuron inputs, all together add to Bias ($B$) to produce the internal result of $Z$. One of activation functions (linear, sigmoidal, or hyperbolic transfer function) to produce the output value of $y_n$ as shown in (8)–(11) respectively will manipulate the instantaneous produced value $Z$:

$$Z = \sum_{i=1}^{N} W_{ni} X_{ni} + \text{Bias}$$  \hspace{1cm} (8)

Linear bipolar: $y_n = f(x)$  \hspace{1cm} (9)

Sigmoidal: $f(x) = \frac{1}{1+\exp^{-x}}$  \hspace{1cm} (10)
Hyperbolic tan: \[ f(z) = \frac{1 + e^{-2z}}{1 - e^{-2z}} \] (11)

The mean square error (MSE) is considered for evaluating the effectiveness of the presented NN algorithm. This parameter is demonstrated in (12), which indicates the difference between the target and predicted values. A smaller value of MSE indicates the accurate performance of the designed ANN algorithm.

\[ MSE = \frac{1}{Q} \sum_{k=1}^{Q} err(k)^2 \] (12)

Where \( Q \) is the input vectors number, and \( err(k) \) is the error between the target and the estimated values.

\[ S(x) = x_1 + x_2 \] (13)

So \( S(x) = -\frac{1}{C}i_L + \left(1 + \frac{1}{RC}\right)V_o + V_{ref} \) (14)

And \( \dot{S} = 0 = -\frac{1}{C}i_L + \left(1 + \frac{1}{RC}\right)V_o + \left(1 - \frac{1}{RC}\right)\left(1 - \frac{1}{C}i_L - \frac{V_o}{RC}\right) \) (15)

A nonlinear component \( u_n \), and an equivalent component \( u_{eq} \) are involved in \( u \) [31]–[34]:

\[ u = u_{eq} + u_n \] (16)

The formula of \( u_{eq} \) can be determined by equaling \( \dot{S} \) to zero.
\[ u_{eq} = \frac{a_1 l_L + a_2 v_o - a_3 v_s}{a_1 l_L + a_3 v_o} \]  
\[ (17) \]

where \( a_1 = \frac{1 - R C}{R C^2} \)  
\[ (18) \]

\[ a_2 = \frac{R^2 C + L R C - L}{L R^2 C^2} \]  
\[ (19) \]

\[ a_3 = \frac{1}{L C} \]  
\[ (20) \]

Whereas the nonlinear component \( u_n \) is:

\[ u_n = \text{sign}(S) \]  
\[ (21) \]

The presented intelligent MPP tracker supported by SMC is demonstrated as a block diagram in Figure 6. The selected PV panel ELDORA 150P is connected to the boost converter. The converter is controlled by the instantaneous value of \( u \) through the pulse width modulation (PWM) pulses generator. The ANN algorithm evaluates the variables of light intensity and ambient temperature for accurately producing a reference voltage. Enforcing the converter to be driven by reference voltage guarantees the MPP position to maximize harvesting the electricity through the PV panel. The designed sliding mode controller produces a suitable switching state \( u \) by monitoring the instantaneous reference voltage, output voltage and inductor current of the converter.

5. SIMULATION RESULTS

The boost converter parameters are designed by considering the study in [4], the selected switching frequency is 40 kHz, as shown in Table 2, in which the converter parameters are also shown. The study ANN involves input layer of two neurons, two hidden layers of four neurons each, and output layer of one neuron. Figure 7 shows the algorithm structure in Figure 7(a) and performance in terms of MSE in Figure 7(b) which reflects an accurate ANN performance of MSE (6.0421e-5) at epoch 7.
Table 2. Designed parameters of boost converter

<table>
<thead>
<tr>
<th>Converter Parameter</th>
<th>Parameter Value</th>
</tr>
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<tbody>
<tr>
<td>Inductor (mH)</td>
<td>10</td>
</tr>
<tr>
<td>Capacitor (µF)</td>
<td>2200</td>
</tr>
<tr>
<td>Connected load (Ω)</td>
<td>15</td>
</tr>
<tr>
<td>Switch frequency f_s (kHz)</td>
<td>40</td>
</tr>
<tr>
<td>Sampling time (µsec)</td>
<td>0.5</td>
</tr>
</tbody>
</table>

Figure 7 ANN algorithm (a) structure and (b) performance MSE

Figure 8 represents the full simulation program, which includes two simulations. The left part is for simulation of MPP tracker using ANN only, whereas the right position is for MPP tracker using ANN supported by the presented SM controller. The simulation results are collected during 1 second of four equally divisions 0.25 second. The simulation is done in a comparative way using ANN algorithm with and without the SMC inserting to evaluate the effectiveness of the SM controller in smoothing the mitigating the overshoot in the load power. Simulation results are collected during ambient room temperature 25 ºC and different light intensity levels; 600, 800, 1000, and 700 W/m² respectively. Figure 9(a) shows the MPP tracker performance based on ANN only. The figure demonstrates the voltage, current, and power of the connected load. Overshoots are clearly noticed in the voltage, current, and power of the load. On the other hand, the positive effect of SM controller in smoothing the shape of load power is noticed in Figure 9(b), which shows the MPP tracker performance when it is supported by sliding mode controller, and all the overshoots are avoided. Figure 10 shows the tracker response in a comparative way before and after inserting the SMC and how the overshoots (red colour) can be smoothed (blue colour) by the inserted SM controller.
6. CONCLUSION

In this study, an intelligent MPP tracker supported by a robust sliding mode controller has been presented. The designed SM controller is suitable for boost converter. An intelligent tracker is evaluated comparatively with and without inserting SM controller. ANN algorithm has been adopted in this study to guarantee the MPP tracking working conditions through fast predicting an accurate reference voltage, and then this voltage has processed by the SM controller to have a smooth response from the tracker and to avoid the overshot in the load power. Simulink of MATLAB is adopted to simulate the presented intelligent MPP tracker, firstly, and the simulation results are collected without inserting the SM controller into the tracker system.
After that, the designed SM controller has inserted to support and enhance the tracker performance. The results indicate the effectiveness of the presented MPP tracker after inserting the SM controller and promise a high-performance prototype as a future step.

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