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Increase Stability and Efficiency in PV-Battery-Grid Systems Using PSO Algorithm

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ABSTRACT

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Keywords:

particle swarm optimization, fuzzy control, photovoltaic cells, microgrids, bidirectional power flow In this article, the meta-heuristic algorithm PSO with PI fuzzy logic controller is proposed to develop a new control strategy of bidirectional converter (CSBC), in order to improve the stability and increase the efficiency of energy flow exchange in grid-connected photovoltaic generator (PVG) Battery hybrid system. The proposed command aims to satisfy the DC Motor load demand, manages the power flows from different parts, injects surplus energy into the grid as and when need, ensure charge-discharge battery operation even under the fluctuating condition of power generation, and stabilize the DC bus voltage. This new control has been placed in the network topology, which consists of a PVG Photovoltaic Generator, grid-connected DC/AC converter, storage battery and DC motor load. A bidirectional buck-boost converter is used for optimum exploit of power from PVG along with battery charging/discharging control, and for feeding DC motor load. The effectiveness of the proposed control scheme is demonstrated by using MATLAB / Simulink program. The results obtained, show that the proposed control provided the system with the stability of $V_{\rm dc}$ voltage, and contributed to improving the speed produced by the DC motor load. The results have been compared with the conventional control.

1. INTRODUCTION

In recent years, the integration of Photovoltaic energy sources has been advanced rapidly to fulfill increasing energy demand and treat with environmental issues. To achieve these goals, great importance has been given to the development of a control system to improve and manage energy in hybrid PVG-Battery-Grid system. Following probabilistic nature of solar energy, the existence of battery ensures continuity of energy flow and increases the reliability of the power supply, when microgrid is enabled with solar energy source [1, 2].

In PVG-battery grid, connected systems a portion of the power is utilized by system load and charge battery and excess power can yield revenue by selling it to the grid [3]. To enhance the stability and power quality provided to both AC and DC micro-grids sides, and make the PVG work in MPPT, several control methods have been discussed in the literature like ref. [4] which has grouped the power converter controllers and supervision module which monitors the parameters of the hybrid system in a single power management and control system (CAPMS), a new control strategy based on five operating modes which are defined by the number of possible cases of switching between the components of the hybrid PVG / battery stand-alone system [5], a control strategy uses adaptive droop control for PVG/battery hybrid unit in an islanded microgrid is proposed, most of them deal with energy management between Photovoltaic battery system and the grid-connected [6]. Nonetheless, this subject still attracts the attention of researchers in order to develop a hybrid power system and improve the quality of the energy produced and exchanged and increase in stability.

This research aims to achieve two goals Firstly extracting

the maximum possible energy (MPPT) and secondly, managing the flow of energy in the network with an increase in efficiency and stability.

The proposed control strategy in this work (CSBC) is used to act both as DC/DC boost converter controller and as DC/Dc buck-boost bidirectional converter controller. The fuzzy regulator optimized by the meta-heuristic algorithm PSO (Particle swarm optimization) has integrated in the structure of CSBC in order to improving stability and efficiency of the system.

2. SYSTEM DESCRIPTION

The proposed model shown in Figure 1, is composed of PVG ($P_{max}=252.4$ kw), Battery energy storage, AC load supplied by a DC/AC converter, and DC motor connected between PVG and the Battery through bidirectional converter.



Figure 1. The proposed structure of the hybrid (PV-Battery-Grid) energy generation system

The bidirectional converter is an important one for energy storage, it is not only enhancing the efficiency, but it also protects the source from under change or over change. it is a dual functionality can be used in (boost) charge mode or (buck) discharge mode [7, 8]. Additionally, the DC-AC power converter is used to transfers and converts the available power at the DC side of a PVG into AC power at the grid side [9]. While, the CSBC Controller is in charge of maintains balance and stability by equalizing power between PVG source, battery storage and DC machine loads in grid-connected mode.

2.1 PV generator (PVG)

The character for non-linearity between current and voltage in a PV panel is deduced from Eq. (1) [10, 11]. Figure 2 shows the (I-V) and (P-V) characteristics of the PVG used in the simulation of the proposed model which consists of solar panels of the brand SPR-E19-410-COM modules.



Figure 2. The (P-V) and (I-V) characteristic curves for different irradiation values, and different temperature values

$$I = N_{p}I_{PV} - N_{p}I_{0}(e^{(q(V+R_{s}\frac{IN_{s}}{N_{p}})/A \ k \ TN_{s})} - 1) - \frac{V + (R_{s}IN_{s} / N_{p})}{R_{p} + (N_{s} / N_{p})}$$
(1)

2.2 Bidirectional DC-DC converter (BDC)

Figure 3 shows the circuit diagram of the basic non-isolated bidirectional DC-DC converter (BDC). It is a device for the purpose of step-up or step-down the voltage level with the capability of flow power in either forward directions or in backward direction [12].

The BDC provides the required bidirectional power flow for battery charging and discharging using two bidirectional switches like MOSFET's or IGBT's [13]. Table 1 below presents the possible operating cases of bidirectional converter, that they are summarized as follows [14, 15].



Figure 3. The circuit diagram of the BDC

Table 1. The possible operating cases of BDC

Case	Switch Diode	State	Power flowing direction	Voltage comparison	Bidirectional Mode	State of battery
	Q_1	ON		\mathbf{V}_{in}		
1	Q_2	OFF	Forward	=V ₁ (high-voltage side)>V _{out}	Buck	Charging
	D2	Active		=V ₂ (low voltage-side)		
	Q_1	OFF		\mathbf{V}_{in}		
2	Q_2	ON	Backward	=V ₂ (low-voltage side) <v<sub>out</v<sub>	Boost	Discharging
	D_1	Active		=V ₂ (high-voltage side)		

Relationships between outputs and inputs (voltage-current) of BDC are determined as follows [16]:

Buck mode

$$\frac{V_1}{V_2} = \frac{1}{D} \tag{2}$$

Boost mode

$$\frac{V_1}{V_2} = \frac{1}{1 - D}$$
(3)

where, D is the duty cycle equal to $D=T_{on}/T$, and switching period is $D=T_{on}+T_{off}$:

2.3 Storage system

Batteries store and release the excess energy produced electrochemically, which qualifies it as a guarantee a continuous power supply for loads. The equivalent circuit of the (Ni-MH) battery model, shown in Figure 4 used in [14] is simulated in this paper.



Figure 4. Battery of storage system

The Eq. (4) determines the relation between the parameters, which characterizes a battery:

$$A.h_{used} = A.h \times (1 - SoC_{(0)}) + \int \frac{((V_{oc} - \sqrt{4R_{iv} - P_{el}})/2R_{iv})}{3600} dt$$
(4)

where, $A.h_{used}$ is the nominal capacity used of battery, and P_{el} is the output electrical power of battery.

2.4 Three-phase inverter DC/AC

The presence of this inverter is important as it separates between PVG and grid; it consists of 03 arms of power switching devices. Each arm consists of 04 switching devices along with their anti-parallel diodes and two neutral clamping diodes as shown in the Figure 5 below [17, 18].



Figure 5. Three-phase inverter DC/AC

2.5 DC motor load

In the wound-field DC machine model, an RL circuit as shown in Figure 6 represents the field circuit, according to the following relationship [19]:

$$\begin{cases} v_{DCch} = e + R_a i_{ch} + L_a \frac{di_{ch}}{dt} = e + R_a i_{ch} \\ e = K_b \omega_m \\ P_{el-DC-ch} = v_{DCch} i_{ch} = e i_{ch} + R_a i_{ch}^2 \end{cases}$$
(5)

where, e inducted electromagnetic force (emf); ω_m speed and K_b constant machine.



Figure 6. Equivalent circuit of a dc motor

3. CONTROL OF SYSTEM

The control system is dividing by two parts as follows: (1) An Inverter control (IC); (2) A control strategy for bidirectional converter (CSBC).

3.1 Inverter control (IC)

To control the electrical parameters of a DC-AC inverter, balancing of grid, and extract Maximum Power of PVG. we use the inverter control which consists of PLL & Measurements, DC Voltage Regulator, Current Regulator, MPPT (using fuzzy logic controller), and PWM Modulator as shown in Figure 7 [20, 21].



Figure 7. Inverter control (IC)



Figure 8. Membership functions

3.2 Maximum power point tracking (FLC)

Figure 8 shows the membership functions of the fuzzy logic controller (FLC) used to follow the MPPT.

3.3 Control strategy for bidirectional converter (CSBC)

In this work the connexion of a battery energy storage system with management Control Strategy optimized by PSO (particle swarm algorithm) For Bidirectional Converter, is required to maintain the stability, supply continuity and viability of the grid.



Figure 9. Intelligent control structure of BDC

The Figure 9 shows the structure adopted for control of buck-boost converter that allows a bidirectional flow of the

electrical energy during charge/ discharge process of Battery. The reference value $I_{bat-ref}$ that produced by dividing the power reference value P_{bat_ref} resulting from the power control unit by the Battery output voltage V_{bat} to generate the reference current [14]. Next, an FLC-PSO controller (PI type FLC tuning by PSO algorithm) regulate the current I-bat-mes to follow the reference signal in order to follow the reference signal. Ultimately, we get a command signal $\Delta d(k)$ used to produce the appropriate switching pulses for the BDC.

The control-to-current transfer function can be obtained in (6). Eq. (6) tells us that this is a third-order power stage model [14].

$$F_{id} = \frac{\hat{i}}{\hat{d}} = \frac{(s + \frac{1}{C_1 R_a}) \cdot (s + \frac{1}{C_2 R_{var_a Bat}}) \cdot V_{pv} - \frac{D}{C_1} \cdot \frac{I_L}{L} (s + \frac{1}{C_2 R_{var_a Bat}})}{s \cdot (s + \frac{1}{C_1 R_a}) \cdot (s + \frac{1}{C_2 R_{var_a Bat}}) + \frac{D^2 (s + \frac{1}{C_2 R_{var_a Bat}})}{LC_1} + \frac{(s + \frac{1}{C_1 R_a})}{LC_2}}$$
(6)

3.4 PIFLC-PSO controller

This tuning method was proposed in Ref. [22]. It's based on a PI controller that expressed in terms of fuzzy rules, in which a PSO algorithm is used to find for the optimal parameters in the fuzzy tuning rules. The Figure 10 shows a PIFLC-PSO controller topology.



Figure 10. Block diagram for the PIFLC-PSO controller

The relation that can be defined as the equation of the PIFLC it's the derivative on both sides of classical PI equation (7):

$$\begin{cases} u = k_p e + k_i \int_{0}^{t} e dt \\ \Delta u = k_p \Delta e + k_i e \end{cases}$$
(7)

In comparison with conventional PI controller, K_p and K_I can be assumed as fuzzy gains or scale factors of FLC [22].

In Figure 10, as shows Block diagram where fuzzy variables before applying to the fuzzy block and fuzzy output variables are multiplied by the scaling factors (G_0 , G_1 , and G_2) in Eq. (8), with tuning of output scaling factor G_2 using PSO algorithm [22, 23].

$$\begin{cases} E(k) = G_0 e(k) \\ \Delta E(k) = G_1 (\frac{e(k) - e(k-1)}{T}) \\ d(k) = d(k-1) + G_2 \Delta d(k) \end{cases}$$
(8)

In this work, Mamdani fuzzy inference system with two inputs and one output is proposed as showed in Plot surface input/output in the Figure 11. Then, scaling the output factor (G_2) was performed using Particle Swarm Optimization (PSO) algorithm that estimates the optimal variable through iterations by minimizing of a cost function. Eq. (9) and (10) presents performance index equation in continuous-time (J_C) and discrete-time (J_D), respectively [22, 23].

$$\begin{cases} J_c = k \int |e(t)| dt \\ e(t) = I_{bat_ref} - I_{bat_dc}(t) \\ t \in [0, t_1] \end{cases}$$
(9)

$$\begin{cases} J_{D} = k \sum_{i=1}^{k} |e(i)| \\ e[k] = I_{bat_ref} - I_{bat_dc}[k] \\ k = 1, 2, ..., iter \max \end{cases}$$
(10)

3.5 Using PSO algorithm for scaling G₂

PSO algorithm is an evolutionary computation technique, based on the swarm behaviour of the birds in nature. He has been applied to almost in every area in optimization and computational intelligence. The algorithm guides the optimization search through cooperation and competition between particles producing swarm intelligence [21, 22]. Individuals called particles, being each particle a possible solution of the problem, form the population. Eq. (11) govern the next position and velocity term of the particle [24, 25]. Equations the PSO algorithm [21-23]:

$$\begin{cases} x_{i}^{(k+1)} = x_{i}^{(k)} + v_{i}^{(k+1)} \\ v_{i}^{(k+1)} = w(k) \cdot v_{i}^{(k)} + c_{1} \cdot r \cdot (x_{pbesti}^{k} - x_{i}^{(k)}) + c_{2} \cdot r \cdot (x_{gbesti}^{k} - x_{i}^{(k)}) \\ w(k) = w_{max} - (\frac{w_{max} - w_{min}}{k_{max}}) \cdot k \\ F(k) = \sum_{i=1}^{N} |e(i)| \\ k = 1, 2, 3, \dots, n \end{cases}$$
(11)

where, x_i^k : Position of i_{th} particle; v_i^k : Velocity of i_{th} particle; P_{best} : Individual best position; k: Number of iterations; G_{best} : Best position reached by the particles of the swarm; w: Inertia term is calculated by Eq. (11); F(k): Fitness function; r: Random generated number with a range of [0, 1].

The main structure of algorithm is: Set the initial random particles, p_{best} , g_{best} , x_{pbest} and x_{gbest} For each interaction k; Update inertia weights by Eq. (11) For each particle i of the population Update velocities and position of all particle at iteration kusing Eq. (11) Evaluate the fitness of all particles. If fitness (x (i)) $\leq p_{best}$ (i) then update xp_{best} and p_{best} $x_{pbest} = x(i);$ p_{best} fitness (x (i)) If $p_{best} < g_{best}$ then update x_{gbest} and g_{best} x_{gbest}=xp_{best}; g_{best}=p_{best}; Next particle Next iteration x_{gbest} is the solution

In this work, the parameters for PSO algorithm using in simulation are show in Table 2.

Table 2. The parameters of PSO algorithm

Parameter	Symbol	Value
Swarm Coefficients	<i>c</i> ₁ , <i>c</i> ₂	2.05
Min weighting coefficient	W_{min}	0.4
Max weighting coefficient	Wmax	0.9
Iteration	k	50
Swarm size	n	25
Lower bound of variables	$\Delta d(k)_{min}$	[-0.15 -0.148]
Upper bound of variable	$\Delta d(k)_{max}$	[0.15 0.151]



Figure 11. Plot surface input/output FLC [E(k), $\Delta E(k)$] and $\Delta d(k)$

4. SIMULATION RESULTS

As shown in the Figure 12, the whole work implemented under Mat-lab/ Simulink software. Here, the performance of PV generator, DC-AC converter, Bidirectional buck-boost converter, battery, Grid connection, DC Motor load, are taken and also our performance is compared by using two techniques of control PI and PIFLC-PSO.



Figure 12. The model simulation



Figure 13. Solar irradiance

The above diagram (Figure 13) represents the solar radiation applied to the PV panels, which changes gradually between 200 and 800 w/m².



Figure 14. Output voltage and current DC-AC converter

Figure 14 shown an output voltage and current of DC/AC Boost converter that connects the PV generator with the grid. During the change of Irradiance, the output AC current also change.



Figure 15. Grid voltage and current source

Figure 16 shows that the delivered photovoltaic power is more stable when not using a smart controller (PI-FLC-PSO) comparing by the classic regulation (PI), this can be seen by zooming the intervals [1.56-1.7] and [2.42-2.52]

The below graph (Figure 17) indicates the comparative analysis of the state of charge Battery (Soc %). The plot is compared in both cases when using the PI controller and PI-FLC PSO controller. This result shows that the minimum value of Soc in the PI-FLC PSO control increases by 0.015% over the other case, taking the same conditions for the proposed system. This slight improvement in the state of charge (Soc%) is due to the rapid search for optimal solution (value of the output-scaling factor G₂) obtained after the calculation of the PSO algorithm this is what accelerates the obtaining the optimal switching angles which act on the bidirectional converter.



Figure 16. Output PV power



Figure 17. State of charge Battery (SOS %)

The Figure 18 represents the voltage DC Bus Vdc, it shows that the PI-FLCPSO has successfully stabilize the DC-link voltage at the reference value (500 V), compared to the other controller, despite the change in solar radiation shown in Figure 13.



Figure 18. Voltage DC Bus (V_{dc})

The plot below (Figure 19) indicates the comparative analysis of the speed DC motor in both cases of control (PI &PIFLC PSO). This comparison shows that the response is rapid and the speed is maintained stable when using a PI FLC controller tuning with PSO algorithm, the delay time between them is equal 0.05 sec.



Figure 19. Speed DC Motor load

5. CONCLUSIONS

In this paper, an intelligent control strategy for a bidirectional DC-DC converter based on the FLC tuned by PSO are proposed in the management the energy flows in the grid-connected photovoltaic Battery storage system. this strategy is implemented and evaluated in the Matlab/Simulink environment.

The optimal value of the output-scaling factor for the PIFLC was calculated applying PSO algorithm, this allows ensuring adaptive regulation of the output currents of the DC motor load and stabilizes the DC link voltage of DC/AC converter and DC/DC bidirectional converter. This is what we observed

through the results obtained, we say that the proposed control provided the system with the stability of V_{dc} voltage, and also contributed to improving the speed of response in the consumption required by the DC motor load.

In perspective of this work, this control Simulink model proposed to better manage the circulation of power flow in hybrid systems (PV/Battery-grid) must be executed on dSPACE platforms to experimentally verify its effectiveness, in very variable weather conditions.

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