Dynamic Modeling and Controller Design of Distribution Static Compensator in a Microgrid Based on Combination of Fuzzy Set and Galaxy-based Search Algorithm

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Abstract

This paper presents a nonlinear controller for a Distribution Static Compensator (DSTATCOM) of a microgrid incorporating the Distributed Generation (DG) units. The nonlinear controller has been designed based on partial feedback linearization theory and Proportional-Integral-Derivative (PID) controllers try to adjust the voltage and trace the output. This paper has proposed a combination of a fuzzy system and Galaxy-based Search Algorithm (GbSA) to optimize the parameters of the PID controllers. The results confirm that the characteristics of the response of the proposed controller (i.e. settling and rise times, the maximum overshoot and the steady-state error of the voltage step response of the DSTATCOM) is significantly improved by finding a high-quality solution. The proposed hybrid tuning method for the Partial Feedback Linearizing (PFL) controller concluded a better DC voltage regulation for the capacitor within the DSTATCOM. Furthermore, in the event of fault the proposed controller tuned by the fuzzy-GbSA method has shown a better performance in comparison with the conventional controller or controllers tuned by Genetic Algorithm (GA) or Particle Swarm Optimization (PSO) methods on both fault duration and after clearing times.


1. Introduction

Distribution networks are enhanced through the upgrade of the recent achievements in distributed generations and Flexible AC Transmission Systems (FACTS) [1, 2]. FACTS devices includes the STATCOM, the Unified Power Flow Controller (UPFC), and the Dynamic Voltage Restorer (DVR) [3]. For optimal design of a controlling strategy the use of the DG and the FACTS units, are important tasks. Controlling the reactive power of the DG or the FACTS units is necessary to keep the voltage profile in the distribution network’s buses within the desirable limit [4]. Gao and Irvani presented a voltage control plan for an interfaced DG unit. They used a Voltage Source Converter (VSC) as the interface medium [5]. Their strategy has the following advantages: 1) enables operation of a DG unit in both grid-connected and islanded (autonomous) modes, 2) caters current limit ability for the VSC during fault, 3) and enables smooth transmission capability between grid connected and islanded modes.

Kumar and Mishra presented an algorithm for a DSTATCOM operating in voltage control mode in literature [6]. The proposed scheme provides Unity Power Factor (UPF) at the load terminal during nominal operation, which is not possible in the conventional technique. Also, the DSTATCOM injects lower currents and, reduces losses in the converter and feeder. Despite these advantages, this model does not consider nonlinear and dynamic model of DSTATCOM. Singh et al. used an optimization control algorithm using an adaptive fuzzy logic controller to regulate the DC bus voltage of the DSTATCOM [7]. The proposed method can control the DC bus voltage of the VSC of the DSTATCOM to modify the response and to reduce only the overshoot and undershoot of the traditional PI controller. Other response features, such as the settling...
and rise times or the steady state error have not been investigated.

Mahmud et al. proposed a nonlinear controller model for the DSTATCOM in a distribution network with the DG units [8]. The controller can regulate the bus voltage at various operating conditions and has a better performance compared to the conventional PI controller. The controller was designed using the PFL theory. After linearization, two PI controllers were employed to track the reference output. The PI controllers have a static relation with the reference value, thus there is no guarantee to get to an optimal set of parameters for the PI controller.

The main objective of this paper is to design a PFL controller for the DSTATCOM considering the nonlinear and dynamic modeling of the DSTATCOM along with tuning the parameters of the designed controller using the combination of fuzzy set and galaxy based search algorithm in order to improve the controller response of the voltage profile at the Point of Common Coupling (PCC).

The rest of the paper is organized as follows: the DSTATCOM modelling and controlling basis is presented in Section 2. Section 3 explains the fuzzy-galaxy-based search algorithm and optimal tuning of the DSTATCOM controller parameters based on this algorithm. The simulation results are presented in Section 4 and the paper has been concluded in Section 5.

2. Nonlinear Modelling and Control of the DSTATCOM

2.1. The DSTATCOM

A DSTATCOM connected to a distribution network [8] is illustrated in Figure 1. Equation (1) can be concluded from Figure 1 as follows:

\[
\begin{align*}
i_a &= -\frac{R}{L} i_a + \frac{1}{L} (v_a - \epsilon_a) \\
i_b &= -\frac{R}{L} i_b + \frac{1}{L} (v_b - \epsilon_b) \\
i_c &= -\frac{R}{L} i_c + \frac{1}{L} (v_c - \epsilon_c)
\end{align*}
\]

where: \(i_{a,b,c}\): The AC currents of DSTATCOM, \(L\): The filter and transformer inductances, \(R\): The inverter and transformer resistances, \(\epsilon_{a,b,c}\): The voltages of the line, \(\epsilon_{a,b,c}\): The inverter output voltages.

If the angular velocity of the AC voltage and current vectors are \(\omega\) and considering the direct-quadrature (dq) reference system is rotating in a similar speed, Equation (1) can be rewritten in the dq reference frame as follows:

\[
\begin{align*}
i_d &= -\frac{R}{L} i_d + \omega q_d + \frac{1}{L} (v_d - \epsilon_d) \\
i_q &= -\frac{R}{L} i_q - \omega d_q + \frac{1}{L} (v_q - \epsilon_q)
\end{align*}
\]

The voltage equations in the dq reference frame can be presented as follows:

\[
\begin{align*}
e_d &= m v_{dc} \sin \phi \\
e_q &= m v_{dc} \cos \phi
\end{align*}
\]

where: \(m\): The modulation indicator, and \(\phi\): The firing angle of the Pulse Width Modulation (PWM) used in the DSTATCOM.

According to equations (2) and (3), the DSTATCOM equation in the dq reference form has been extracted as follows:

\[
\begin{align*}
i_d &= -\frac{R}{L} i_d + \omega q_d + \frac{1}{L} (v_d - m v_{dc} \cos \phi) \\
i_q &= -\frac{R}{L} i_q - \omega d_q + \frac{1}{L} (v_q - m v_{dc} \sin \phi) \\
v_{dc} &= \frac{1}{C} [m \sin \phi q_d + m \cos \phi d_q]
\end{align*}
\]

where \(v_{dc}\) is the voltage of the capacitor and \(C = \frac{3}{2} C_{dc}\) which \(C_{dc}\) is the capacitor’s capacitance.

In a three phases-balanced circuit, the quadrature part of the voltage is equal to zero [9]. Therefore, \(v_q = 0\) and \(Q = \frac{3}{2} V d i_d\) and by controlling the current \(i_q\), the reactive power has been controlled. Also, a suitable compensation of reactive power can be obtained by regulating the \(v_{dc}\) properly.

2.2. The Controller Design

The State-Space (SS) model of the DSTATCOM in Equation (4) can be formulated as follows:

\[
\begin{align*}
z &= \begin{bmatrix} i_q \\ v_{dc} \end{bmatrix} \\
z &= m(z) + n(u) = \begin{bmatrix} R L i_d + \omega q_d + \frac{v_d}{L} \\ -\frac{R}{L} i_q - \omega d_q \end{bmatrix} + \begin{bmatrix} \frac{v_d}{L} \\ 0 \end{bmatrix} \\
y &= \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} i_q \\ v_{dc} \end{bmatrix} \\
u &= \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} m \cos \phi \\ m \sin \phi \end{bmatrix}
\end{align*}
\]
This equation describes a nonlinear system with two inputs and outputs.

2. 2. 1. Linearization of the DSTATCOM Equations

In this section, the PFL theory is implemented using Equation (5) to linearize the nonlinear system.

Feedback linearization theory is a multi-variable control technique based on a mathematical model of the procedure to be controlled. This method converts the nonlinear system into a linear system through varying the variables and selecting a suitable control input. If the following form is presented for a system in the steady state:

\[ \dot{z} = m(z) + n(z)u \]

\[ y = h(z) \]

(6)

where \( z \), \( m(z) \), \( n(z) \), \( u \) and \( y (h(z)) \) are the state variables, state matrix, inputs coefficients, inputs variables and the outputs vectors, respectively, then the main purpose of using the feedback linearization theory is to design a controllable input which it can offer a linear relation between the new controller input and the old system input as follows:

\[ u = a(z) + b(z) v \]

(7)

This equation represents a feedback scheme between the new controller input \( v \) and old system input \( u \). This theory is used for linearizing the system of Equation (5). By applying the PFL theory, the DSTATCOM system is transformed into two linear equations and all the nonlinearities would be removed. Both linear equations have the stable dynamics [8]. If the nonlinear DSTATCOM system in Equation (5) with vector \( z \) transforms into the system with the new state variable \( m \) in Equation (8) [8, 10]:

\[ \dot{m} = Cm + Dv \]

(8)

(where \( C \) is the state matrix and \( D \) is the input vector of the system), then the new DSTATCOM equations are as follows:

\[ \dot{m} = -a \cdot \frac{R}{L} \cdot \frac{v_{di}}{L} \cdot u_2 \]

\[ \dot{m} = \frac{1}{C} \cdot l_1 \cdot u_1 + \frac{1}{C} \cdot l_2 \cdot u_2 \]

(9)

The system in above equation can be linearized as follows:

\[ \dot{m}_1 = i_1 \]

\[ \dot{m}_2 = i_2 \]

(10)

where, \( i_1 \) and \( i_2 \) are the new linear inputs as described using Equation (11):

\[ i_1 = -a \cdot \frac{R}{L} \cdot \frac{v_{di}}{L} \cdot u_2 \]

\[ i_2 = \frac{1}{C} \cdot l_1 \cdot u_1 + \frac{1}{C} \cdot l_2 \cdot u_2 \]

Thus, based on the above equation and Equation (7), the control law can be delivered as:

\[ u_1 = \frac{-C}{i_d} \left[ i_d + i_q \right] \cdot \frac{L}{i_d \cdot v_{di}} \cdot i_d + \frac{L}{v_{di}} \cdot \omega + \frac{R}{i_d \cdot v_{di}} \]

\[ u_2 = \frac{L}{v_{di}} \left[ i_1 + a \cdot \frac{R}{L} \cdot \frac{v_{di}}{L} \right] \]

Two new controllers \( i_1 \) and \( i_2 \) must be properly designed for precise tracking of reference outputs \( (i_{d,c}, \) and \( v_{dc,c}) \). For this purpose, in this paper, the two PID controllers are proposed as follows:

\[ i_1 = k_1 e_1 + k_2 \int_{0}^{t} e_1 dt + k_3 \frac{de_1}{dt} \]

\[ i_2 = k_2 e_2 + k_3 \int_{0}^{t} e_2 dt + k_4 \frac{de_2}{dt} \]

(13)

where, \( e_1 = i_{d,c} - i_q \) and \( e_2 = v_{dc,c} - v_{dc} \) are the pursuit error values of the \( i_q \) and \( v_{dc} \) respectively.

3. PROPOSED METHOD FOR TUNING THE CONTROLLER PARAMETERS

This section describes the Galaxy-based Search Algorithm (GbSA) followed by a description of the tuning of DSTATCOM by fuzzy-GbSA-PFL-PID method.

3. 1. Galaxy-based Search Algorithm (GbSA)

The Galaxy-based Search Algorithm (GbSA) was first proposed by Hamed Shah-Hosseini [11, 12]. It is an optimization technique that tries to mimic the arms of spiral-based galaxies moving to avoid the local optima. The flowchart of the GbSA is given in Figure 2. In general the GbSA is comprised of two main components, namely the Spiral Chaotic Move and Local Search.

In normal working state, the Spiral Chaotic Move is iterated for MaxRep number of times. For searching better solution, the Spiral Chaotic Move uses a spiral movement enhanced by a chaotic variable generated by Next Chaos.

The chaotic sequence is generated by the logistic map:

\[ x_{n+1} = \lambda x_n (1 - x_n) \]

\[ n = 0, 1, 2, \ldots \]

(14)

In this paper, \( \lambda = 4 \) and \( x_0 = 0.2 \).

The Local Search is used to search other possible solutions within the current solution denoted as (SG). Once found, Flag will be given a True value and the local search function is activated. The local search is used to search possible optimum solutions within the updated current solution.
Once a better solution compared to the current solution is obtained, the function is terminated and the control function is transferred to the local search.

3.2. Fuzzy-GbSA-PFL-PID Controller

The gains of the suggested PID controllers are chosen in such a way to minimize the observed pursuit errors (e1 and e2). Equation (13) shows the gains of the PID controllers are affected by the reference values of the current and voltage, which in turn these values depend on the amount of the predicted reactive power to improve the voltage stability. The prediction technique for the reactive power has been discussed in reference [13]. According to the predicted reactive power (\( Q_{ij} \)), the reference current (\( i_{q_{ij}} \)) is evaluated as follows [8]:

\[
i_{q_{ij}} = \frac{2Q_{ij}}{\gamma d}
\]

Based on the obtained \( i_{q_{ij}} \), the DC voltage reference (\( V_{dc_{ij}} \)) is evaluated by Equation (16).

\[
v_{dc_{ij}} = v_d + R_i q_{ij} + \omega L_i q_{ij}
\]

By calculating the reference values of \( i_{q_{ij}} \) and \( V_{dc_{ij}} \), the gains of the controllers should be chosen in such a way to minimize the observed pursuit errors. In this paper, a galaxy-based search algorithm has been employed for tuning the PID controllers gains in such a way that the step response characteristics including the maximum overshoot (Mp), the SS error (Ess), the settling time (\( t_s \)) and the rising time (\( t_r \)) to be improved. Improvement of both rise and settling times increase the speed of the system response, but given that by reducing the rise time, initial speed response of the controller will improve and by reducing the settling time, the final part of the response experiences the steady state in a shorter time, so in order to enhance the speed of the controller’s response (from beginning to end), both rise time and settling time have been considered in the objective function. According to above explanations, a multi objective function is defined for the optimization problem as follows:

\[
F = a_1 M_p + a_2 E_{ss} + a_3 t_r + a_4 t_s
\]

where, \( a_1, a_2, a_3 \) and \( a_4 \) are the weighting coefficients of the objectives.

When there are various objectives to be considered simultaneously, a comparison is needed to get the best answer. Since various options of the objective function are in different ranges, all the objectives are normalized in the same range to prevent the convergence problem [3]. A fuzzy system is used for homogenization the objectives. Since the different parts of the objective function are in different rates, all the values should be normalized in the similar range [14-16]. Each objective has a membership function (\( \mu \)) that indicates the efficacy of its objective as Equation (18):

\[
\mu_{ij}(X) = \begin{cases} 
1, & f_{ij}(X) \leq f_{ij}^{\text{max}} \\
\frac{f_{ij}^{\text{max}} - f_{ij}(X)}{f_{ij}^{\text{max}} - f_{ij}^{\min}}, & f_{ij}^{\min} < f_{ij}(X) < f_{ij}^{\text{max}} \\
0, & f_{ij}^{\max} \leq f_{ij}(X)
\end{cases}
\]

where \( f_{ij} \) is the obtained value for the \( i \)th part of the objective function in the \( j \)th answer, \( f_{ij}^{\text{min}} \) is the best answer of in the single objective optimization for the \( i \)th objective function and \( f_{ij}^{\text{max}} \) is the worst answer which of the single objective optimization for the \( i \)th objective function.

The fuzzified objective functions are gathered into a fuzzy objective function as follows:

\[
F(X) = \sum_{i=1}^{4} a_i \mu_{ij}(X)
\]

where \( a_i \) is the weighting factor for the objective \( i \).

(17)

(The weighting coefficients of above equation are same as Equation (17)). To enhance or to degrade the effectiveness of each objective, different values can be considered for the coefficients. In this paper, the weighting factors are considered \( a_1=0.25, a_2=0.25, a_3=0.25, a_4=0.25 \).
which the four objectives are considered to have equal importance.

The combination of fuzzy systems and GbSA that were described in previous sections is used to find an optimal tuning of the controller in the following steps:

**Step 1:** Read all required data of distribution network; set the GbSA parameters. Produce an initial solution randomly.

**Step 2:** For each GbSA iteration process, run the power flow program [3], compute the fuzzy objective degree (i.e. $Mp$, $ts$, $tr$, and $Ess$). Compute the fitness value of the fuzzified objective function and store the current solution as the best answer.

**Step 3:** Update the GbSA parameters and consider it as a new solution. For the new solution, run the process in Step 2.

**Step 4:** If the fitness value of the new solution is better than the best solution, then replace it.

**Step 5:** If the stopping criteria is reached, go to Step 6, otherwise go to Step 3 and repeat the process.

**Step 6:** Defuzzify the best solution to obtain the optimum PID gains for the controller.

**Step 7:** Terminate the algorithm.

### 4. SIMULATION RESULTS AND DISCUSSION

The performance of the proposed controller has been investigated on a modified test power system with the proposed microgrid [17]. This power distribution system operates at 25 kV and 50 Hz and includes a different of DG Resources units (DG based on fossil and renewable fuels) and variety types of loads. The test system presented in literature [17] has been modified by adding three new buses with a Photovoltaic (PV) array at bus 14, 7 kilometer additional transmission line between buses 13 and 15. Also a new load (2 MW + 0.5 Mvar) was assumed at bus 15 and at bus 12, the capacity of DSTATCOM is considered as 2.5 Mvar. This test system has been shown in Figure 3. As shown in this figure, bus 4 is point of common coupling (PCC). For this test system with microgrid, three types of the DG units including wind turbine, PV and micro-turbine are connected to bus 8, bus 14 and bus 6 respectively, to supply the loads of buses 9, 15 and 11 and the over plus power was injected to the Distribution System (DS).

Two different cases have been assumed to analysis the step response characteristics of the voltage at bus 12 (PCC) as follows:

**Case (a):** The voltage response at bus 12 to a sudden voltage step of 2% in $v_{dc}$ while the DSTATCOM has been equipped with the PFL-PID controller without optimization by the fuzzy-GbSA technique.

In this case, the gains should be selected such that the output follows the reference values to minimize the error.

In this study, the gains are selected as follows: $k_{p1} = 2iu_{r}$, $k_{u} = iu_{r}$, $k_{d1} = 2/3iu_{r}$, $k_{d2} = 2v_{dc}$.

As it is seen, the gains of the controllers depend on the reference values (i.e., $v_{dc}$, and $i_{r}$/). These reference values are calculated by Equations (15) and (16).

**Case (b):** The voltage response at bus 12 to a sudden voltage step of 2% in $v_{dc}$ while the DSTATCOM uses the PFL-PID controller tuned by the fuzzy-GbSA technique.

The optimized gains of two PID controllers ($k_{p1}$, $k_{u}$, $k_{d1}$, $k_{d2}$) have been given in Table 1 for two cases (a) and (b). The calculated $ts$, $tr$, $Mp$ and $Ess$ values have been presented in Table 2.

For case (a), from Table 1 it can be realized that by using the controller without applying optimization, the first PID gains are computed as $k_{p1} = 0.7375$, $k_{u} = 0.3846$ and $k_{d1} = 0.0963$ and the second PID gains are calculated as $k_{d2} = 0.4629$, $k_{s2} = 0.5126$ and $k_{d2} = 0.1975$. Also for this case from Table 2, $ts$, $tr$, $Mp$ and $Ess$ values are obtained as 0.4128 (s), 2.0316 (s), 0.0386 %, and $2.5\times10^{-2}$ respectively.

For case (b), as shown in Table 1, the first PID gains are obtained as $k_{p1} = 0.6835$, $k_{u} = 0.2194$ and $k_{d1} = 0.1509$ and the second PID gains are calculated as $k_{d2} = 0.8453$, $k_{s2} = 0.3138$ and $k_{d2} = 0.1871$. For this case, from Table 2, $ts$, $tr$, $Mp$ and $Ess$ values are obtained as 0.4094 (s), 0.6390 (s) 0.0136 % and $2.03\times10^{-4}$ respectively.

Figure 4 shows a graphical comparison between the step response parameters for the both cases (a) and (b).

![Figure 3. Single line diagram of the modified test power system with the proposed microgrid](image-url)
From this figure and Table II it can be understood that the controller parameters tuned by the proposed method significantly improve the four step response characteristics i.e. $M_p$, $t_s$, $t_r$ and $E_{ss}$ values.

The convergence behavior of the fitness function is presented in Figure 5 i) for both case (a) and case (b). As seen in this figure, the final fitness value is 0.1319 and 0.1210 for case (a) and case (b) respectively. Also, from this figure, it can be comprehended that case (b) is converged to a better optimal fitness value compared to case (a).

The GbSA parameters have been heuristically chosen through a trial and error process as follows:

- $\phi_0 = -\pi$, $\gamma_0 = 1$, $\Delta \theta = 0.001$, $r = 0.001$, $L = 70$,
- $k_{MAX} = 120$ and MaxRep=120. In this case, the total CPU time (in seconds) spent for running the proposed algorithm in Matlab (2013b) software was 32.19 s.

It should be mentioned that the experiments are performed on a laptop with a Pentium IV CPU running Microsoft Windows 7 operating system.

Case (b) is simulated using GA [18] and PSO [19] methods, to compare with the results obtained by proposed fuzzy-GbSA technique.

Table 3 represents a comparison between the obtained step response characteristics using GA, PSO and proposed fuzzy-GbSA for case (b). From this table, it is observed that the performance of the fuzzy-GbSA is better compared to GA and PSO, in improving all terms of the settling time, the maximum overshoot, the rise time and the steady-state error of the step response.

The convergence behavior of the fitness function using GA, PSO and proposed method (fuzzy-GbSA) has been compared in Figure 5ii).

As shown in this figure, for GA technique the final fitness value is 0.1275 and for the PSO method it was equal to 0.1252 while the corresponding obtained value for the proposed method is equal to 0.1210. Therefore, it can be concluded that proposed technique has a better optimal fitness value compared to the GA and PSO methods.

Furthermore, the effectiveness of the proposed controller can be investigated at the time of occurrence of fault within the distribution system. If the fault remains for a long time, it may lead to voltage instability within the network. In this condition, it is vital to preserve the voltage profile of the PCC to arrest voltage instability. This task can be achieved by interchange adequate energy between the DSTATCOM and distribution system whenever needed. To investigate the effectiveness of the proposed controller in the time of occurrence of fault within the distribution system, it was assumed a fault is happened at bus 2 at $t=1$ sec and will be disappeared at $t=1.2$ s.

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**TABLE 1.** The computed gains for the PID controllers based on cases (a) and (b)

<table>
<thead>
<tr>
<th></th>
<th>Case (a)</th>
<th>Case (b)</th>
</tr>
</thead>
<tbody>
<tr>
<td>$k_{1p} * k_{2p}$</td>
<td>0.7375, 0.4629</td>
<td>0.6835, 0.8453</td>
</tr>
<tr>
<td>$k_{1i} * k_{2i}$</td>
<td>0.3846, 0.5126</td>
<td>0.2194, 0.3138</td>
</tr>
<tr>
<td>$k_{1d} * k_{2d}$</td>
<td>0.0963, 0.1975</td>
<td>0.1509, 0.1871</td>
</tr>
</tbody>
</table>

**TABLE 2.** The step response parameters ($M_p$, $t_s$, $t_r$ and $E_{ss}$ values) for cases (a) and (b)

<table>
<thead>
<tr>
<th></th>
<th>Case (a)</th>
<th>Case (b)</th>
</tr>
</thead>
<tbody>
<tr>
<td>$M_p$</td>
<td>0.0386</td>
<td>0.0136</td>
</tr>
<tr>
<td>$t_r$</td>
<td>0.4128</td>
<td>0.4094</td>
</tr>
<tr>
<td>$t_s$</td>
<td>2.0316</td>
<td>0.6390</td>
</tr>
<tr>
<td>$E_{ss}$</td>
<td>$2.5 \times 10^{-2}$</td>
<td>$2.03 \times 10^{-4}$</td>
</tr>
</tbody>
</table>

**TABLE 3.** Comparison of the obtained step response characteristics using GA, PSO and proposed fuzzy-GbSA methods for case (b)

<table>
<thead>
<tr>
<th>Item</th>
<th>GA, PSO and Fuzzy-GbSA results</th>
</tr>
</thead>
<tbody>
<tr>
<td>$M_p$</td>
<td>0.0259, 0.0179, 0.0136</td>
</tr>
<tr>
<td>$t_r$</td>
<td>0.4293, 0.41521, 0.4094</td>
</tr>
<tr>
<td>$t_s$</td>
<td>0.6453, 0.6407, 0.6390</td>
</tr>
<tr>
<td>$E_{ss}$</td>
<td>$2.03 \times 10^{-4}$, $3.7 \times 10^{-4}$, $4.2 \times 10^{-4}$</td>
</tr>
</tbody>
</table>
Three various cases are proposed to evaluate the performance of the proposed tuned controller in the event of the fault as follows:

**Case (I):** using a conventional controller for controlling the DSTATCOM [12].

**Case (II):** using the PFL-PID controller for controlling the DSTATCOM without optimization by the fuzzy-GbSA.

**Case (III):** using the PFL-PID controller tuned by proposed fuzzy-GbSA method for controlling the DSTATCOM.

For the three cases, the voltage profile of the PCC has been shown in Figure 6. As can be seen in Figure 6, for case (I), the distribution system cannot return to the pre-fault condition after disappearance of fault.

For case (II), the network observes a better voltage profile. Although the voltage profile is returned to the pre-fault condition but this process takes about a few tenths of a second.

For case (III), the distribution system experiences a better voltage profile compared to case 2 at the fault duration time and the voltage profile can return to the pre-fault condition without delay. Therefore, on both fault and clearing times, the PFL-PID controller tuned by the fuzzy-GbSA technique has a better performance compared to the conventional or the PFL-PID controller without optimization by the fuzzy-GbSA method.

5. CONCLUSION

This paper presents a nonlinear PFL-PID controller for the DSTATCOM tuned by a combination of fuzzy system and GbSA approach. Two various cases were assumed to analysis the step response characteristics of the voltage at PCC and three different cases were considered to evaluate the performance of the proposed controller in the event of the fault within the network. The performance of the proposed controller was evaluated by testing it on a modified test power system with the proposed microgrid. Analyzing the step response of the voltage of the DSTATCOM indicates that tuning the PFL-PID controller parameters using the fuzzy-GbSA method, can significantly improve the rise time, maximum overshoot, settling time and steady state error values compared to the controller without optimization with the fuzzy-GbSA. Without tuning the parameters of the controller, the obtained fitness value was 0.1319 while the fitness value 0.1210 achieved when the parameters of the controller were tuned using the proposed technique.

It can be concluded the proposed method with the fitness value equal to 0.1210 has a better performance compared with the GA and the PSO algorithms with the fitness equal to 0.1275 and 0.1252 respectively.

Furthermore, the controller was evaluated in the event of fault within the system. In this condition, the PFL-PID controller tuned by fuzzy-GbSA method provides better results compared to the conventional controller or the PFL-PID controller without optimization by the fuzzy-GbSA method on both fault duration and after clearing times.

7. REFERENCES


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Keywords: Distribution System, Distribution Static Compensator, Galaxy-based Search Algorithm (GbSA), Fuzzy Sets, Optimization

Abstract: In this paper, a new nonlinear controller is designed for a DSTATCOM in a microgrid based on a parallel hybrid system of the Fuzzy Set (FS) and Galaxy-based Search Algorithm (GbSA). The controller is designed using the feedback linearization (FLT) method based on the DC grid voltage. The controller is designed to handle the nonlinearities and uncertainties of the system. The simulation results show that the proposed controller can effectively control the DSTATCOM and improve the power quality in the microgrid.