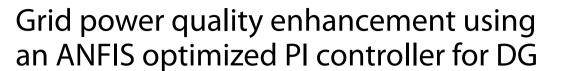
ORIGINAL RESEARCH

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Srishail K. Bilgundi¹, R. Sachin², H. Pradeepa^{1*}, H. B. Nagesh² and M. V. Likith Kumar¹

Abstract

Grid frequency variation causes phase angle deviation in current with respect to voltage. This is sensed at the phase-locked loop in the controller. In past studies the effect of grid frequency variation is neglected while designing the controller for power quality restoration. When modern grids are connected to large numbers of non-linear loads and various types of distributed generation (DG), it results in continuous variation in grid frequency. Thus it is necessary to consider the grid frequency variation for effective power quality restoration. However, tuning of conventional Pl controller gains considering frequency variation is very difficult. Thus it is necessary to develop an adaptive intelligent nonlinear controller to tackle the effects of frequency variation, voltage distortion and non-linear load simultaneously. This paper presents the importance of considering the effects of the frequency variation, grid voltage distortion and non-linear load, while designing and deploying a controller for power quality restoration. The proposed controller supplies power to local load as well as transferring surplus power to the grid from DG along with the additional benefit of improving grid power quality. A DG with an ANFIS optimized Pl current controller for power quality enhancement is proposed. The method is economical as it requires no additional hardware. Results are compared with Pl, Pl-RC and fuzzy current controllers to validate the effectiveness of the proposed controller.

Keywords: Adaptive neuro fuzzy inference system (ANFIS), Fuzzy logic, PI controller, Distributed generation (DG), Proton exchange membrane fuel cells (PEMFC), Total harmonic distortion (THD)

1 Introduction

The growing global energy crisis, environmental pollution and depletion of fossil fuels have led to greatly increased use of alternate renewable energy sources. As a result, renewable energy sources in the form of DGs are integrated in large numbers into power systems [1]. There are many advantages of DG systems over traditional power generation using fossil fuels, such as clean electric power generation, low cost, high efficiency, and small size. In a grid synchronized mode, a DG system is typically operated to dispatch the extracted power from renewable energy sources to the utility grid. To fully utilize the advantages of a DG system, it can supply power

to the local loads and the available power surplus is then transferred to the grid [2]. In both cases without and with the local load configurations, the DG system needs to supply high-quality grid current with a THD less than 5%, as recommended in the IEEE519 Standard [3].

Various strategies for current control have been introduced to produce high-quality grid current, such as PI, predictive, hysteresis, proportional-resonant (PR) and artificial intelligence-based controllers. Hysteresis control offers rapid response and is simple, though it creates disturbances with variable and high frequency switching, which results in complex design of the output filter and high current ripple [4]. Another viable solution for grid-connected DG current regulation is predictive control. However, a strong controller function depends on system parameters [5], and thus quality of the grid current is affected by the system uncertainties. The effective solution used to achieve high-quality grid current is

¹ Department of Electrical and Electronics Engineering, The National Institute of Engineering, Mysuru, Karnataka, India Full list of author information is available at the end of the article



^{*}Correspondence: pradeep3080@nie.ac.in

a PI controller in the $\alpha\beta$ and dq frames. However, such a current controller is effective only when the voltage of the grid is sinusoidal and balanced. However, in modern power systems, most loads are controlled by AC/DC converters acting as nonlinear loads. Thus the grid voltage at the PCC is usually distorted or unbalanced, and consequently the performance of the DG controller in supplying high-quality grid current to the power system can deteriorate [6].

In [6], a novel compensation approach is introduced for THD reduction by controlling DG current under grid voltage distortion, where the grid voltage harmonics are extracted using the Cauchy–Schwarz theory of inequality. However, the quality of grid current heavily depends on the accuracy of harmonic analysis of the grid voltage, so less accurate harmonic analysis makes it difficult to maintain the quality of grid current. Moreover, this method operates only off-line, as the required calculation time is very large.

In [7] and [8], a resonant controller using several selective harmonic compensators is developed in which multiple resonant controllers are used to eliminate the 5th and 7th components of the grid voltage harmonics by tuning the controller at the multiple 6th harmonics. Improved quality of grid current can then be achieved by the multiple resonant controllers. However, because only one specific harmonic component is regulated by a single resonant controller, many resonant controllers would be required to compensate high order harmonics. This increases the control system complexity. A control scheme is adopted with the simplified technique of repetitive control, where a bank of resonant controllers, called repetitive controllers (RC), compensate higher order harmonics with simple delays to improve the grid current quality [9].

The presence of nonlinear local load and grid voltage distortion has a major impact on the quality of supply DG current [10]. To address this problem, a feed-forward loop with measurement of local load current is adopted in [10] and [11]. However, its main drawback is that it requires additional current sensors for measuring the local load current. Furthermore, none of the methods simultaneously tackle the impact of the local nonlinear load, grid voltage distortion, and grid frequency variation.

Under dynamically changing operating conditions and continuous variations in system parameters, it is difficult to obtain satisfactory results by use of a conventional PI controller. An option for dynamic response improvement in any system is self-tuning, although it is time-consuming and not easy to realize a controller with self-tuning PI gains. The use of intelligent techniques such as neural networks, fuzzy logic and the ANFIS system are some of

the alternatives. For vague and non-linear systems, the simple rule-based fuzzy logic control (FLC) can be used, while FLC can be made adaptive by the ANFIS which is a combination of artificial neural networks (ANN) and FLC [12]. An advanced space vector pulse width modulation (SVPWM) based predictive current (PC) controller for switching pulse control in an inverter is introduced in [5]. A comparative study of different DG current controllers for power quality improvement is proposed. Applications of ANN and Fuzzy in various research areas have been carried out in [13]-[14] where increased accuracy and functionality are achieved.

In a traditional PI controller, the gains are calculated assuming the grid frequency is constant, whereas in practice this is not true. Thus, the traditional PI controller fails to compensate for power quality issues when frequency variation exists along with grid voltage distortion and non-linear loads. The aforementioned limitations are addressed in this paper, and the novelty of the proposed work is to simultaneously tackle the impact of the local nonlinear load, grid voltage distortion and grid frequency variation, and to compensate for power quality issues by optimally tuning the PI controller with the help of intelligent techniques like Fuzzy, ANFIS, etc. Different case studies are carried out with traditional PI controllers and the proposed intelligent PI controllers, and the results are compared to show the effectiveness of adopting intelligent techniques for tuning PI controller parameters.

2 Analysis of distributed generation-fuel cell

In a fuel cell, chemical energy from a hydrogen and oxygen electro-chemical reaction is converted directly into electrical energy. The by-products are water and a small amount of carbon dioxide (only when CO and $\mathrm{CH_4}$ are used other than pure hydrogen). The fuel cell is considered as a promising technology to meet energy demands, and is well suited for distributed generation because of its modularity, scalability, high efficiency and no harmful emission. The most commonly used fuel cell is a PEM fuel cell, because it can operate at a low temperature of $20-100~\mathrm{^{\circ}C}$. Figure 1 shows the schematic diagram of a PEM fuel cell, and the equations representing the hydrogen and oxygen electro-chemical reaction are [15]:

$$H_2 = 2H^+ + 2e^- \tag{1}$$

$$\frac{1}{2}O_2 + 2H^+ + 2e^- = H_2O$$
 (2)

$$H_2 + \frac{1}{2}O_2 = H_2O \tag{3}$$

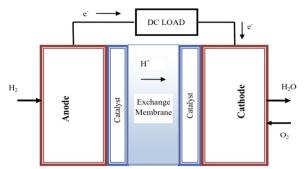


Fig. 1 Schematic diagram of PEMFC

Figure 2 shows the model of a fuel-cell, which is a modified version of the proposed approach in [16]. The cell output voltage V is given by [17]:

$$V = E_{OC} - V_{act} - V_r \tag{4}$$

where

$$V_{act} = A \ln \left(\frac{i_{fc}}{i_o} \right) \cdot \frac{1}{\frac{sT_d}{3} + 1}$$
 (5)

$$A = \frac{RT}{Z\alpha F}$$

$$V_r = r_{ohm} i_{fc} \tag{6}$$

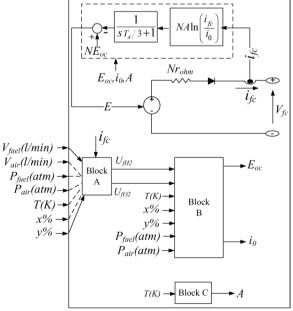


Fig. 2 Typical fuel cell stack model

and the output voltage of multiple cells in series is given by:

$$V_{fc} = N \cdot V \tag{7}$$

where E_{oc} is the generated voltage, V_{act} is the activation losses (initial reaction kinetics losses), and V_r is the diffusion and resistive losses (losses due to charge transport). T_d is the settling time delay of the cell, i_{fc} is the fuel cell current output, and i_0 is the exchange current. A is the Tafel slope, R=8.3145 J/(molK), F=96,485 As/mol, z is the number of moving electrons, α is the coefficient of charge transfer, T is the operating temperature and N is the number of cells in series.

3 Effect of grid voltage distortion, nonlinear load and grid frequency variation on current

Figure 3 shows a typical DG connected to a distribution system. The system consists of a 3-phase VSI with dispatchable DG source (i_{DG}) and an LC output filter. These supply local load (i_L) and the available power surplus is transferred to the utility grid (i_g) . The DG current transferred to the local load and to the grid must be sinusoidal, balanced, and THD at PCC must be less than 5%. However, satisfying these requirements is not easy because of nonlinear local loads and distortions in the grid voltage.

3.1 Evaluating effect of grid voltage distortion

Figure 4 shows the Thevenin equivalent of the DG system connected to the grid. To assess the impact of grid voltage distortion and local nonlinear load on the performance of DG, the DG-VSI configuration is modeled as V_i , while R_f and L_f are the equivalent resistance and inductance, respectively. To simplify the evaluation of the effect of grid voltage distortion, the local nonlinear load is not considered here. From Fig. 4a, there is:

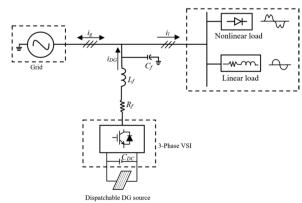


Fig. 3 Typical DG connected to distribution system

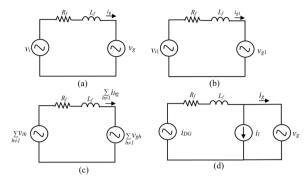


Fig. 4 Model of grid-connected DG system under distorted grid voltage condition and with nonlinear local load

$$v_i - v_g - L_f \frac{di_g}{dt} - R_f i_g = 0 \tag{8}$$

The system model shown in Fig. 4a can be decomposed into Fig. 4b, c, where the grid voltage and inverter voltage in Fig. 4b only have fundamental components whereas in Fig. 4c, the grid voltage and inverter voltage contain both fundamental and harmonic components. The inverter and grid voltages (v_i and v_g) with fundamental and harmonic components are represented respectively as:

$$\left\{ \begin{array}{l} v_i = v_{i1} + \sum_{h \neq 1} v_{ih} \\ v_g = v_{g1} + \sum_{h \neq 1} v_{gh} \end{array} \right\}$$
(9)

Thus, (8) can be rewritten by considering the fundamental and harmonic components separately as:

$$v_{i1} - v_{g1} - L_f \frac{di_{g1}}{dt} - R_f i_{g1} = 0$$
(10)

$$\sum_{h \neq 1} \nu_{ih} - \sum_{h \neq 1} \nu_{gh} - L_f \frac{d\left(\sum_{h \neq 1} i_{gh}\right)}{dt} - R_f \sum_{h \neq 1} i_{gh} = 0$$
(11)

Because of the grid harmonic voltage components $\sum_{h \neq 1} \nu_{gh}$ and inverter harmonic voltage components $\sum_{h \neq 1} \nu_{ih}$, current harmonics $\sum_{h \neq 1} i_{gh}$ are induced into the grid current. If DG cannot compensate for inverter voltage harmonics $\sum_{h \neq 1} \nu_{ih}$ and grid voltage harmonics $\sum_{h \neq 1} \nu_{gh}$, it will result in non-sinusoidal grid current i_g . Thus, the current controller needs to handle inverter voltage harmonics $\sum_{h \neq 1} \nu_{gh}$, and grid voltage harmonics $\sum_{h \neq 1} \nu_{gh}$.

3.2 Evaluating effect of nonlinear load

To evaluate the effect of local load, the DG grid with a local load is shown in Fig. 4d, where i_l is represented as the consumed local load current, i_{DG} as the generated current and i_g as the grid current. From Fig. 4d it is seen that:

$$i_{DG} = i_l + i_g \tag{12}$$

Considering nonlinear load, the load current contains the fundamental component i_{L1} and harmonic components i_{Lh} , as:

$$i_l = i_{L1} + \sum_{h \neq 1} i_{Lh} \tag{13}$$

Substituting (13) in (12) yields:

$$i_g = i_{DG} - \left(i_{L1} + \sum_{h \neq 1} i_{Lh}\right)$$
 (14)

From (14), it can be seen that to supply sinusoidal grid current i_g , DG current i_{DG} must compensate $\sum_{h\neq 1} i_{Lh}$ for the load current harmonics. Traditional current controllers such as PI or PR are ineffective, because of lack of capability for selective harmonic component regulation. Therefore, to generate the specific harmonic components, design of a low-cost and effective current controller is necessary.

3.3 Effect of frequency variation

Since the frequency in the grid varies continuously, any change in frequency reference will significantly affect the controller performance and output current could be out of phase. System impedance is decreased because of the decrease in system frequency, and this results in increased current distortion.

4 Proposed control scheme

The proposed advanced current control strategy simultaneously addresses the effects of local nonlinear load, grid voltage distortion and frequency variation. The DG current control strategy is developed in the dq reference frame as shown in Fig. 5. The three main parts of the control scheme proposed include a DG current controller, current reference generation and a phase-locked loop (PLL). As shown in Fig. 5, the control strategy operates without grid voltage harmonic analysis and local load current measurement, and thus it does not require additional hardware.

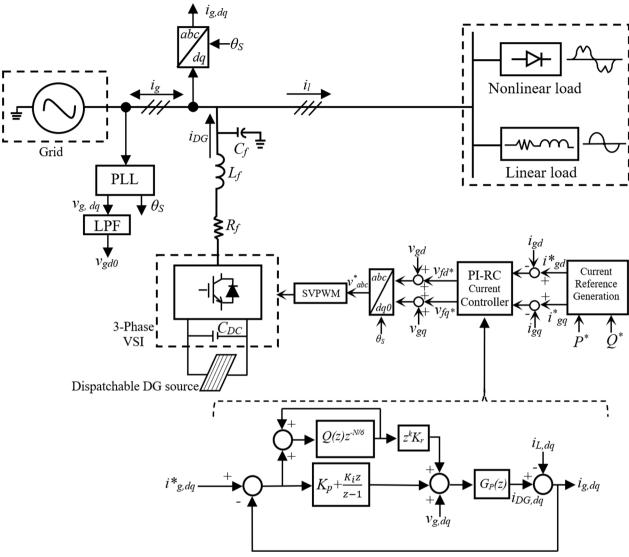


Fig. 5 Model application of PI-RC in mitigation of power quality problems

4.1 Current reference generation

As shown in Fig. 4, the current references for the current controller are generated as [11]:

$$\begin{cases}
i_{gd}^* = \frac{2}{3} \frac{P^*}{v_{gd}} \\
i_{gq}^* = \frac{2}{3} \frac{Q^*}{v_{gd}}
\end{cases}$$
(15)

where i_{gd}^* and i_{gq}^* are the d- and q-axis components of the reference grid current, respectively. v_{gd} represents the instantaneous d-axis grid voltage component. P^* and Q^* are the reference active and reactive power, respectively. Under ideal conditions the grid voltage is purely sinusoidal so v_{gd} has a constant value. However, the magnitude of v_{gd} varies if the grid voltage is distorted. This can result in variation of reference current i_{gd}^* and i_{gq}^* . To overcome

this problem, the average value of v_{gd} is estimated by a low-pass filter, and (15) is modified as:

$$\begin{cases}
i_{gd}^* = \frac{2}{3} \frac{P^*}{\nu_{gd0}} \\
i_{gq}^* = \frac{2}{3} \frac{Q^*}{\nu_{gd0}}
\end{cases}$$
(16)

where v_{gd0} is the average value of v_{gd} obtained through the LPF.

4.2 Current controller

4.2.1 PI-RC current controller

The proposed DG current controller strategy developed in the dq reference frame includes PI and RC controllers. The simplified control strategy can be achieved as large numbers of harmonic components can be compensated by one single RC with a simple delay function. With the PI and RC controller combination, the current controller dynamic response is greatly enhanced over that of the traditional RC. Therefore, with reduced numbers of sensors, the control method proposed can be adopted easily without extra hardware installation into the existing DG control system.

In the discrete time domain, PI and RC controller open-loop transfer functions are given respectively as:

$$G_{PI}(z) = K_p + \frac{K_i z}{z - 1} \tag{17}$$

$$G_{RC}(z) = \frac{K_r z^k z^{-N/6}}{1 - O(z)z^{-N/6}}$$
(18)

where K_p is the proportional gain and K_i is the integral gain of the PI controller. Q(z) is a filter transfer function, z^k is the phase lead term, K_r is the RC gain, while $z^{-N/6}$ is the time delay unit.

N/6 is the number of sample delay given in (18) of the RC, and $N=f_{sample}/f_s$, where f_{sample} and f_s are the sample frequency and fundamental system frequency, respectively. However, traditional RC has severe drawbacks as it suffers from a long delay of N samples resulting in a very slow dynamic response. The delay problem of traditional RC can be removed by considering only $(6n\pm1)$ th harmonics because they are the dominant components in three-phase systems, where (n=1, 2, 3, ...). Thus, compared to the traditional RC, time delay is reduced by a factor of six to N/6.

4.2.2 Fuzzy current controller

Figure 6 shows the schematic diagram of the fuzzy controller for DG current control. FLC makes decisions using data sets of different types for reasoning. FLC is chosen over conventional PID and PI controllers as system parameter variation is very insensitive in FLC while FLC execution is very smooth and simple.

In the case of the Mamdani fuzzy model, an essential requirement is expert knowledge. It is more human-like with intuitive skill, although it causes significant computational burden and requires an interface procedure. The Sugeno technique has successful computation with versatile design, appealing to various control issues of nonlinear dynamic systems. The FLC scheme is divided into 3 main blocks, i.e., functional decision-making, fuzzification and defuzzification.

The fuzzification purpose is to control change in the error ΔE , convert the input–output signal crisply and error E into fuzzified signals recognized by fuzzy sets with different levels of membership functions. The linguistic variables of error rate are shown in Fig. 6a,

linguistic variables of error are in Fig. 6b, and linguistic variables of output in Fig. 6c, with a triangular membership function as positive small (*PS*), positive big (*PB*), positive medium (*PM*), negative small (*NS*), negative big (*NB*), negative medium (*NM*) and zero (*ZE*). The membership functions of the decision-making output are either linear or constant in the Sugeno strategy.

Defuzzification represents the output of linguistic labels which are used to control the output to what it needs to be. A crisp solution is changed from fuzzification. The most widely recognized strategy is Sugeno's weaver

In the decision-making process, there is a rule-based link between input (error and rate of change in error of current) and output signals. Table 1 shows the rule base used in the FL controller.

In the proposed method, the network voltage of each phase is tracked by the PLL and a reference signal with magnitude unity locked to grid frequency is generated for each phase. The proposed FLC has two inputs measured in real time at each sampling time, i.e., the error rate and error. The error and error rate for phase A is defined as:

$$e_A = V_{PLL_A} - V_{S_A} \tag{19}$$

$$\Delta e_A = err_A(n) - err_A(n-1) \tag{20}$$

The fuzzified signal input is represented by membership functions in notations of a fuzzy set. The linguistic variables produced by rules defined by 'if ... then ...' and control signals are obtained by defuzzification to generate gating pulses by the SVPWM. The process of defuzzification solution results in the form:

$$U = \frac{\sum_{i} y_{i} \mu(y_{i})}{\sum_{i} \mu(y_{i})} \tag{21}$$

where the crisp value of the output is denoted by U, $\mu(y_i)$ is the membership associated grade, and y_i represents the controller normalized output for the ith interval.

4.2.3 ANFIS optimized PI current controller

Figure 7 shows the ANFIS optimized PI current controller for DG current control. From the PI current controller initial data, the ANFIS optimized PI current controller architecture is generated using the ANFIS editor. This architecture has 1:3:3:3:3:1 structure, so contains one input and one output, and by an error propagation algorithm it is further tuned online. The square node is an adaptive node and its parameters are changed during training, while the parameters presented in the circle nodes are fixed. Node functions of each layer in the ANFIS structure are described as follows.

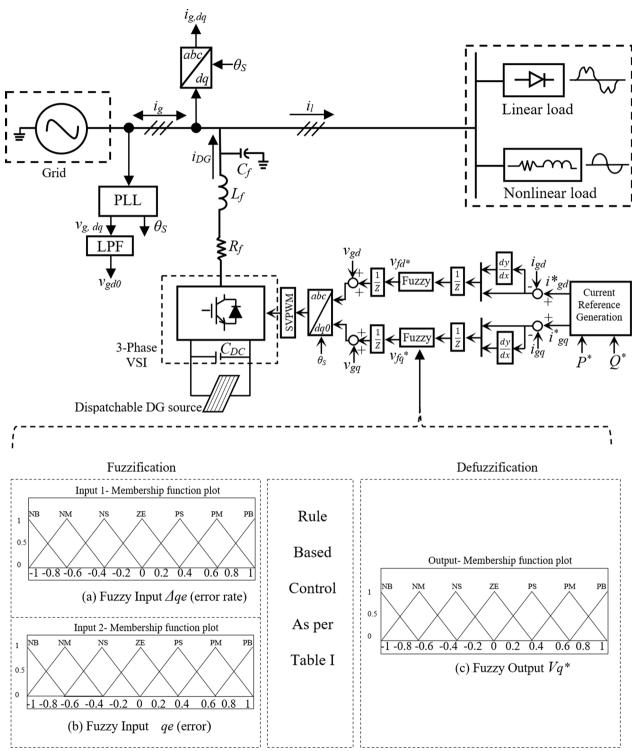


Fig. 6 Application of fuzzy in mitigation of power quality problems

 Layer 1: Parameters in this layer are called precondition or premise parameters, represented by a square, and are used for fuzzification. Corresponding node equations are:

$$O_i^1 = \mu A_i(ve_1) \tag{22}$$

$$\mu A_{i}(ve_{1}) = \begin{cases} 0 & ve_{1} \leq a_{i} \\ \frac{ve_{1} - a_{i}}{b_{i} - a_{i}} & a_{i} \leq ve_{1} \leq b_{i} \\ \frac{c_{i} - ve_{1}}{c_{i} - b} & b_{i} \leq ve_{1} \leq c \\ 0 & c_{i} \leq ve_{1} \end{cases}$$
(23)

where i = 1,2,3, O_i^1 is the output of the ith node in layer-1, and a_i , b_i , c_i are the parameters of the triangular membership function.

• Layer-2: Parameters in this layer are labeled as π_i and represented by a circle. They are used for the firing system by multiplying the incoming signals and forwarding them to the next layer. Corresponding node equations are:

$$w_j = \mu A_1(ve_1) \times \mu A_2(ve_1) \times \mu A_3(ve_1)$$
 (24)

where j = 1,2,3.

• *Layer-3*: Parameters in this layer are labeled N and represented by a circle. For each rule, the normalized firing strength is calculated in this layer as:

$$\overline{w_j} = \frac{w_j}{\sum_{k=1}^3 w_k} \tag{25}$$

where j = 1,2,3.

Layer-4: Parameters in this layer are called consequent parameters represented by a square. They give the output of each node. Corresponding node equations are:

$$O_j^4 = \overline{w_j} f_j = \overline{w_j} (p_j v e_1 + t_j)$$
 (26)

- where $\overline{w_j}$ is the layer-3 output, O_j^4 is the jth node layer-4 output, and p_j and t_j are determined during training. They are sets of consequent parameters.
- Layer-5: Parameters in this layer are labeled ∑ represented by a circle. They are used for summation. This layer sums up all incoming signals to calculates the output y. Corresponding node equations are:

$$y = \sum_{i=1}^{3} \overline{w_i} f_i = \sum_{i=1}^{3} \left[\left(\overline{w_j} \nu e_1 \right) r_j + \left(\overline{w_j} \right) t_j \right]$$
 (27)

Similarly, node functions of each layer of the ANFIS structure are also obtained for the q-component.

5 Simulation results

There are some assumptions made for fuel cell operation, i.e., properties of fed hydrogen and oxygen into the PEMFC stack are ideal, PEMFC anode and cathode temperature is stable and equal to the temperature of the stack, PEMFC pressure drop is negligible and cell resistance is constant.

In Fig. 8, the voltage and current characteristics of the fuel cell output are shown. A nominal rated current of 133.3A and voltage of 45 V are generated from the combination of 65 PEMFC stack models resulting in a power output of 133.3*45 \approx 6 kW. The output voltage of the PEMFC is given to the DC-DC boost converter to match the required voltage of the DC bus.

The harmonics contained in the distorted grid voltage in the simulation are: 5th=3%, 7th=3%, 11th=1%, 13th=1%, resulting in a THD of 4.82%. Thus the distorted grid voltage THD complies with the IEEE519 limits. The optimized gains of different controllers are calculated as explained in previous sections and are tabulated in Table 2.

To analyze the effectiveness of the proposed ANFIS-PI optimized DG current controller, it is compared with PI, PIRC and Fuzzy DG current controllers for compensating

Table 1 Rule based representation

E	ΔΕ									
	NB	NM	NS	ZE	PS	PM	PB			
NB	PB	PB	PB	PM	PM	PS	ZE			
NM	PB	PB	PM	PM	PS	ZE	ZE			
NS	PB	PM	PS	PS	ZE	NM	NB			
ZE	PB	PM	PS	ZE	NS	NM	NB			
PS	PM	PS	ZE	NS	NM	NB	NB			
PM	PS	ZE	NS	NM	NM	NB	NB			
PB	ZE	NS	NM	NM	NB	NB	NB			

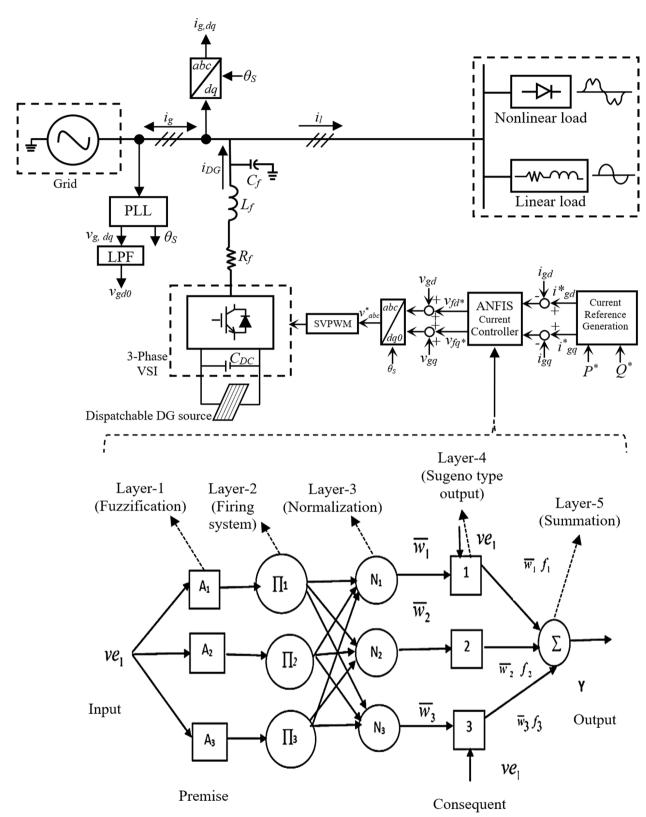


Fig. 7 ANFIS optimized PI controller for DG control

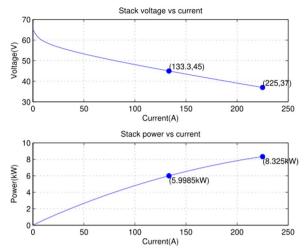


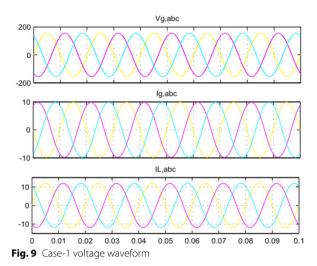
Fig. 8 Fuel cell V-I characteristics and output power curve

Table 2 Rule based representation

Controllers	Gains
PI controller	$K_p = 10, K_i = 250$
PI-RC controller	$K_r = 1, K_p = 5, K_i = 0.5$
Fuzzy controller	Adaptive self-tuned by algorithm
ANFIS PI optimized controller	Adaptive self-tuned by algorithm

power quality issues in various case studies, and the results are listed in Table 3. The case studies considered are:

- Case 1: Grid supplying sinusoidal voltage to linear load
- Case 2: Grid supplying sinusoidal voltage to nonlinear load.
- Case 3: Grid supplying distorted voltage to nonlinear load.
- Case 4: Grid supplying distorted voltage with frequency variation of 50 Hz to 49 Hz to nonlinear load.



Case 1: As shown in Fig. 9, where the grid voltage V_g is sinusoidal, DG supplies i_l to linear local load and also transfers surplus power to grid. Under such operating conditions, the THD of the grid current i_g at PCC is 1.57% which is well within the IEEE519 limits.

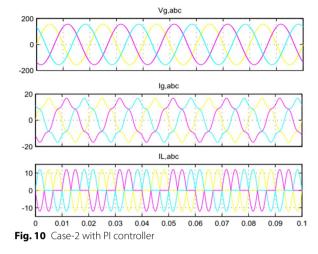
Case 2: As shown in Figs. 10, 11, 12 and 13, where grid voltage V_g is sinusoidal, DG supplies nonlinear local load current i_l and also transfers surplus power to the grid. Under such operating conditions, 54.74% of THD is introduced in the grid current because of the nonlinear load. DG with the PI current controller compensates harmonics by reducing THD of the grid current i_g at PCC to 11.9% but it fails to meet the limits of IEEE519 as shown in Fig. 10 and Table 3.

However, DG current controllers of PI-RC, Fuzzy and ANFIS-PI optimized controller as seen in Figs. 11, 12 and 13, are able to compensate harmonics by reducing the THD of the grid current $i_{\rm g}$ at PCC to 1.76%, 1.35% and 1.35%, respectively, which are all well within the IEEE519 limits.

Case 3: As shown in Figs. 14, 15, 16 and 17, where the grid voltage V_g is non-sinusoidal, i.e., distorted grid voltage with a THD of 4.82%, DG supplies i_l to nonlinear local load and also transfers surplus power to grid. Under

Table 3 Comparative study of DG current controllers for harmonic mitigation

Harmonic mitigation	Grid voltage	Local nonlinear load i _i	Compensated grid current i_g			
	distortion V_g		PI	PI-RC	Fuzzy	ANFIS-PI optimized
Case-1 (THD-%)	_	_	1.57	1.57	1.57	1.57
Case-2 (THD-%)		54.74	11.9	1.76	1.35	1.35
Case-3 (THD-%)	4.82	54.74	18.6	1.85	1.35	1.35
Case-4 (THD-%)	24.91	61.35	22.6	25.2	12.4	4.5



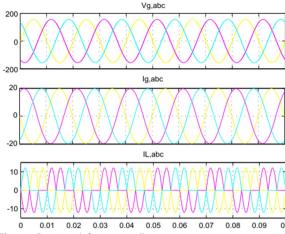
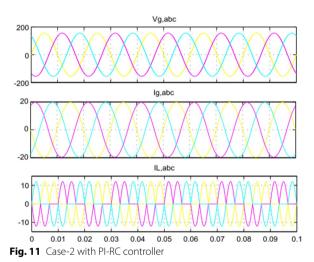


Fig. 12 Case-2 with fuzzy controller



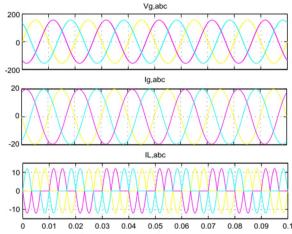


Fig. 13 Case-2 with ANFIS-PI controller

such operating conditions, 54.74% of THD is introduced by the nonlinear load with an additional 4.82% of THD from the distorted grid voltage. DG with the PI current controller compensates harmonics by reducing the THD of the grid current i_g at PCC to 18.6% but it fails to meet the IEEE519 limits, as shown in Fig. 14 and Table 3.

However, PI-RC, Fuzzy and ANFIS-PI optimized controller as seen in Figs. 15, 16 and 17 are able to compensate harmonics by reducing the THD of the grid current $i_{\rm g}$ at PCC to 1.85%, 1.35% and 1.35%, respectively, which are well within the IEEE519 limits.

Case 4: As shown in Figs. 18, 19, 20 and 21, the effect of grid frequency variation from 50 to 49 Hz is introduced along with the non-sinusoidal grid voltage V_g with a THD of 24.91%. DG supplies nonlinear local load current i_L and also transfers surplus power to the grid. Under such operating conditions, 61.35% of the THD is introduced by the nonlinear load with 24.91% introduced by

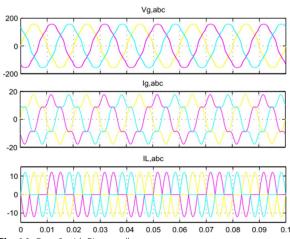
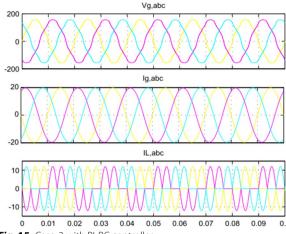
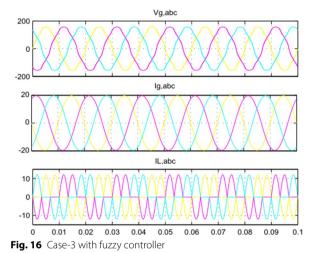


Fig. 14 Case-3 with PI controller







Vg,abc

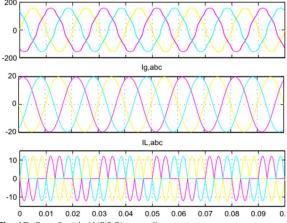


Fig. 17 Case-3 with ANFIS-PI controller

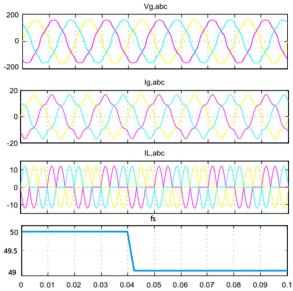


Fig. 18 Case-4 with PI controller

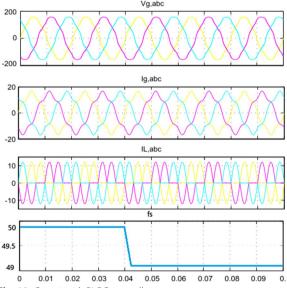
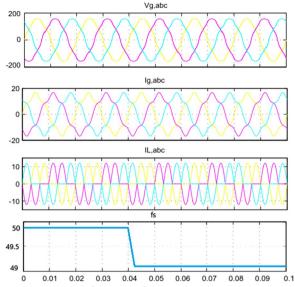


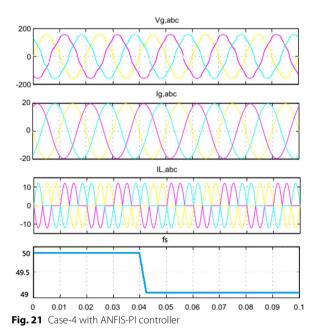
Fig. 19 Case-4 with PI-RC controller

the distorted grid voltage. DG with PI, PI-RC and Fuzzy current controller compensates harmonics by reducing THD of the grid current i_g at PCC to 22.6%, 25.2% and 12.4%, respectively. However, all the 3 controllers fail to compensate harmonics sufficiently to meet the IEEE519 limits as shown in Figs. 18, 19 and 20.

However, the proposed ANFIS-PI optimized DG controller shown in Fig. 21 is able to compensate harmonics by reducing the THD of the grid current i_g at PCC to 4.5% which is now within the IEEE519 limits. For all four







case studies, the results obtained from each type of controller tabulated in Table 3 are also represented using a bar graph in Fig. 22.

6 Conclusion

An advanced DG current control strategy is developed to supply power to local load and transfer surplus power to the grid along with the additional benefit of improving the grid power quality. This paper demonstrates the importance of considering the effect of grid frequency

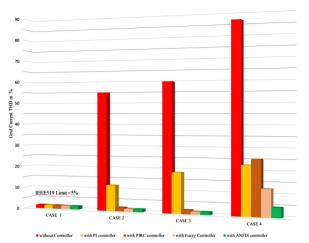


Fig. 22 Bar graph comparing THD% for all 4 case studies

variation while designing and deploying the controller for power quality restoration, and the need for adaptive intelligent controllers to compensate power quality issue. The proposed intelligent current controller eliminates the effect of grid frequency variation, voltage distortion and nonlinear loads simultaneously. The results show that the ANFIS-PI optimized DG current controller has better harmonic compensating capability by limiting harmonics well within the limits defined in the IEEE519 standards, whereas other controllers fail to compensate sufficiently the THD to meet the limits. Thus, this work conveys the effectiveness of adapting intelligent techniques for tuning PI (conventional) controllers to tackle multiple power quality problems simultaneously. The proposed method is economical as it does not require any additional hardware installation for measuring local load current for feedback control. Thus, it can be easily coupled with existing control schemes.

Authors' contributions

All authors read and approved the final manuscript.

Author's information

Srishail K. Bilgundi born in Karnataka, India, received Bachelor of Engineering degree from PDA college of Engineering in 2015 and Master of Technology degree from The National Institute of Engineering, Mysuru in 2017. Currently pursuing Ph.D. as full time research scholar in the Dept. of Electrical and Electronics Department at The National Institute of Engineering, Mysuru, Karnataka, India under Visvesvaraya Technological University Belgaum. Area of research is Power Systems which includes distributed energy resources, Renewable Energy, Power Quality, FACTS, Energy Management and Smart Grid.

R. Sachin born in Karnataka, India, received Bachelor of Engineering degree from Bangalore Institute of Technology in 2015 and Master of Technology degree from The National Institute of Engineering, Mysuru in 2017. Currently pursuing Ph.D. as full time research scholar in the Dept. of Electrical and Electronics Department at Bangalore Institute of Technology, Karnataka, India under Visvesvaraya Technological University Belgaum. Area of research is Power Systems which includes distributed energy resources, Renewable

Energy, Energy Management, Power Quality, FACTS, Power systems dynamic and controls and Smart Grid.

- **H. Pradeepa** Received Bachelor of Engineering degree from Banglore University, M. Tech. from National Institute of Technology Karnataka, Surathkal, and Ph.D. from Visvesvaraya Technological university, Belagavi. Currently working as an Associate Professor in the Dept. of Electrical and Electronics Engineering at The National Institute of Engineering, Mysore, Karnataka, India. Research interest is in power system which includes distribution system, renewable energy, and smart grid. Currently working on a project titled "Digital Protection Algorithms for Renewable Integrated Grid", Sponsored by VGST, and Govt. of Karnataka".
- **H. B. Nagesh** Received Bachelor of Engineering degree and M. Tech. from Mysore University, and Ph.D. from Visvesvaraya Technological University, Belagavi. Currently working as an Associate Professor in the Dept. of Electrical and Electronics Engineering at Bangalore Institute of Technology, Karnataka, India. Research interest is in Power Systems which includes Renewable Energy, Energy Management, Power Quality, Power system voltage stability, Power systems dynamic and controls and Smart Grid.
- M. V. Likith Kumar received BE degree in EEE from VTU, Belagavi, Karnataka in 2010. M. Tech. in Power Systems and PhD in faculty of Electrical Sciences from The National Institute of Engineering, Mysuru affiliated to VTU, Belagavi, Karnataka in 2016. He is currently working as Associate Professor in the Department of Electrical and Electronics Engineering, The National Institute of Engineering, Mysuru, India. He has published 20 papers in indexed journals and reputed conferences. His research interests include soft computing application in power system Engineering, Power Quality, Distributed Generation, Smart and Microgrid.

Availability of data and materials

The Authors confirm that the data supporting the findings of this study are available within the article and any further data required data will be provided by corresponding author on request.

Declarations

Competing interests

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Author details

¹Department of Electrical and Electronics Engineering, The National Institute of Engineering, Mysuru, Karnataka, India. ²Department of Electrical and Electronics Engineering, Bangalore Institute of Technology, Bangalore, Karnataka, India.

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