Hybrid Energy System for EV Charging Station Using PV Fed Z-Source Modified Luo Converter and Fuel Cell Integration

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Abstract:

Petroleum-based vehicles are rapidly being phased out of transportation sector as number of Electric Vehicles (EVs) rises. Nonetheless, timely and well-coordinated growth of EV Charging Stations (CSs) are crucial for quick implementation of EVs. Since randomly distributed Photovoltaic (PV) system causes high power losses and voltage variances that exceed permitted limitations, integrating EVCSs into the current distribution network is difficult. This research proposes an EV charging solution supported by PV systems and fuel cells (FC) to sustain a variable load under an optimized energy management strategy. To overcome the intermittent nature of PV system, a Z-source modified Luo converter is employed, which provides higher efficiency and voltage gain. A Walrus optimized Proportional Integral (PI) controller is utilized to improve performance of developed converter with stabilized voltage. The performance of developed FC system is improved by using Boost converter with PI controller. In order to meet load requirement of the grid-connected charging station during unavailability of PV or FC shutdown condition, the grid source power the EV, thus ensures effective EV charging even under uncertain environmental conditions. The developed work is verified through MATLAB/Simulink and outcomes reveal higher converter efficieny of 97.14% with improved voltage gain, offering uninterrupted power supply to EVs.

Keywords:

1. Introduction

The rise in the world's need for power has prompted the utilization of fossil fuel reserves and negatively impacted environmental situations, ultimately leading to global warming. With zero tailpipe emissions, EVs are now regarded as one of the most efficient forms of transportation [1-4]. However, EVs can't directly utilize fossil fuel resources; rather, their indirect fossil fuel consumption is increased by the energy derived from the system that distributes fossil fuels. Furthermore, EVs only be maintainable while the electrical energy necessary for charging is developed from RES [5-8]. Microgrids are a means of achieving consistent and dependable power generation through the use of renewable energy [9, 10]. Among all the RES that is incorporated into CS system, the PV source is acknowledged as one of the most practical options [11-13]. Regarding the increased need for fast charging during the day, the quick growth of PV generation maximizes power usage during peak hours by providing sufficient daytime generation. One practical way to support more clean energy, lower carbon emissions, and lessen the demand on the local grid for electricity is to directly integrate PV system with EV charging stations [14,15]. Because Renewable Energy Sources (RESs), especially PV is intermittent, DC-DC converters need to be employed to stabilize the power output fluctuations [16]. In [17] Boost converter is developed that increases the input voltage to a higher output voltage. However, it produces significant output voltage ripple, requiring additional filter to guarantee smooth output. According to its design, the SEPIC converter [18, 19] operate in boost or buck mode by utilizing the PV system's maximum power. However, at greater duty ratios, the SEPIC converter is no longer able to generate a substantial voltage gain. In [20], the Luo converter is demonstrated offering impressive features including increased voltage gain, less component count, minimal capacitor stress and low voltage stress on power switches However, the voltage stress on the switches of converter is not equal to the output voltage. In [21] Z-source converter is developed that step up and down the voltage, providing greater flexibility. Nevertheless, extra components are required for Z-network, which improve the

complexity and cost of the overall system. Therefore, Z-source modified Luo converter is developed for improving voltage in this paper.

PI controllers are employed for enhancing grid-connected PV system's transient stability and several optimization approaches are used for best possible tuning of PI control parameters. The literature [22] presents Particle Swarm Optimization (PSO) is comparatively simpler than other optimization methods. PSO is good at avoiding local optima and investigating the global search space. The initial placements of the particles have a significant impact on PSO performance; incorrect initial positions might result in more iterations and longer convergence periods. The harmony search method, which is developed in [23], has a more robust convergence when it comes to the accuracy and efficiency of the solution while optimizing objective functions. Nevertheless, if the optimization algorithm concentrates on exploration, it will converge very slowly. Better performance is assured with optimal PI controller tuning, however these optimization procedures frequently have lengthy execution times and are complex. An Artificial Bee Colony optimized PI Controller is developed in [24], which has less settling time. However, the computational complexity and the requirement for quick convergence make the implementation of ABC-optimized PI controllers difficult. In [25] Cuckoo search optimization algorithm is presented, which has strong global search capabilities, which helps in finding the optimal PI controller parameters. However, there is a risk of premature convergence, where the algorithm get trapped in a local optimum rather than predicting global best solution. To overcome the above mentioned problems, this work develops Walrus optimization algorithm for refining PI controller efficiency.

Energy Management System (EMS):

Seamless power is supplied to the EV system by the integration of dynamic EMS that regulates

the co-ordinated operation of PV, FC, battery and grid source. The EMS maximizes the use of RES by extracting power from PV system if solar irradiance is adequate. In the absence of PV system, FC is turned ON to provide supplemental power through PI-controlled Boost converter. If both PV and FC are inadequate, the grid system charges the EV battery effectively.

The objectives of this paper are,

- Utilization of Z-source modified Luo converter for boosting PV system voltage for EV charging stations.

- For controlling the output of developed converter Walrus optimized PI controller is exploited.

- For supplying extra power supply to EV battery the fuel cell is employed with Boost converter.

- The battery with bidirectional converter is exploited to provide continuous supply to Evs.

- For control of bidirectional converter the ANN controller is employed on load side.

2. Proposed Methodology

The developed PV-Fuel cell based EV charging stations is illustrated in Fig. 1. Normally, the output voltage of PV system is low owing to variation in sunlight conditions. To maximize the PV system's low voltage, a Z-source modified Luo converter is employed that improves the efficiency of overall system. Meanwhile the converter's output is unstable, which is controlled through the implementation of Walrus tuned PI control method that provides the stabilized power. Also, the Fuel cell acts as another resource and its low voltage is enhanced by boost converter in which PI controller is used to control it. Then, the 3ϕVSI is used to convert the DC supply into AC, which is also regulated by PI controller and PWM generator is exploited to improve the inverter's functionality, guaranteeing low-distortion and high-quality AC power The power supply is passed to 3ϕ grid through LC filter that removes unwanted noise and harmonics presented in the supply. The inverter enables grid-connected operation, allowing bidirectional power exchange and

synchronization with the utility grid. This setup supports microgrid functionality, also as a stabilizing and

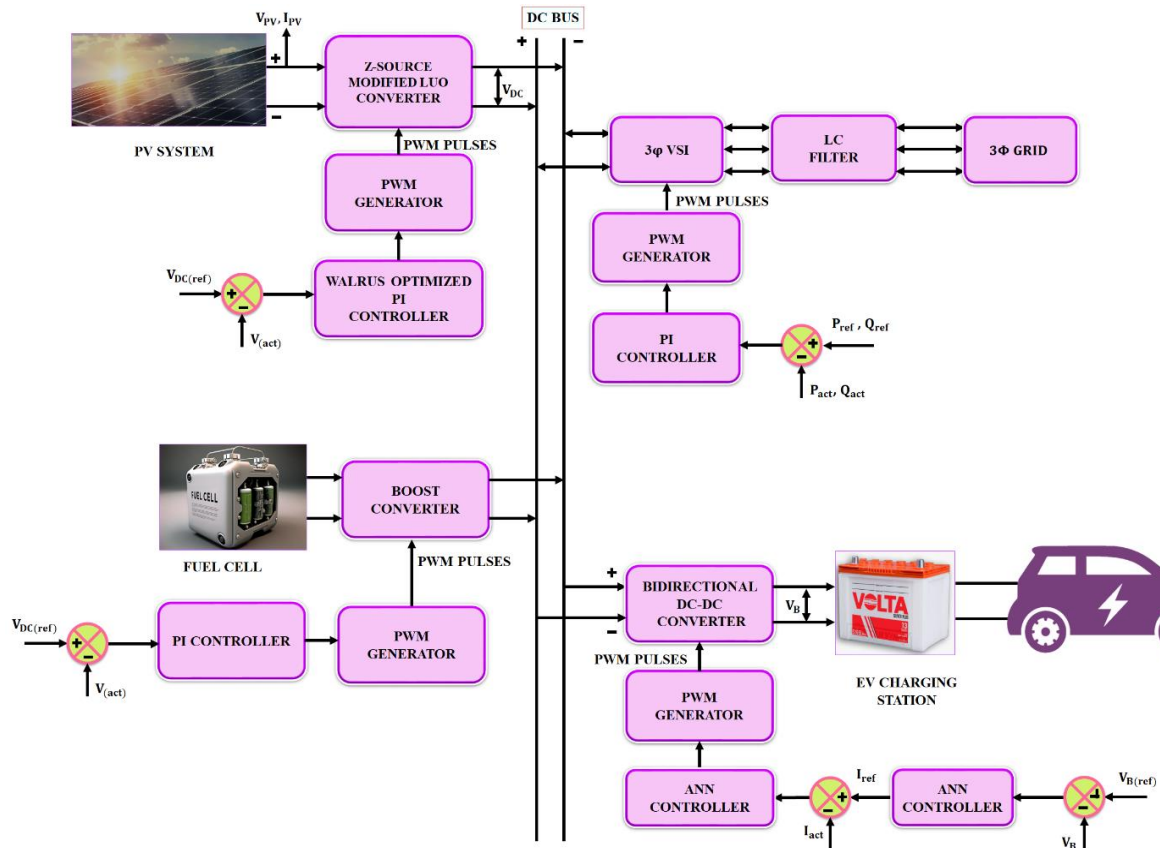supplementary source during dynamic load conditions.

Fig. 1. PV-FC based EVCS

Subsequently, additional energy from both PV and FC system is given to the grid system while bidirectional

converter performs charging and discharging operation based on the need of converter in which its function

is regulated by ANN controller, thereby delivers consistent power supply to EVs.

2-1- PV system

To create solar panels, many PV cells are linked in parallel or series. In a parallel design, the current rises

while voltage remains stable, but in a series configuration, the voltage increases while the current remains

constant. N_p and N_S represent the total number of solar cells attached in shunt and series. The total

photocurrent, is the result of solar irradiation and cell temperature. Fig. 2 displays the single diode circuit
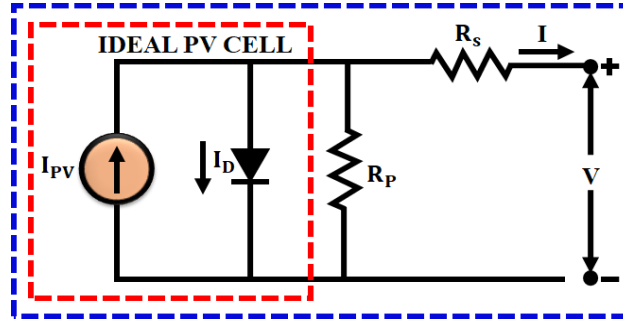
diagram for PV system.

Fig. 2. Circuit of PV model

The output voltage and current is related as,

$$I_{PV} = I_{ph} - I_O \left[e^{\frac{q(V_{PV}+I_{PV}R_S)}{N_S knT}-1} \right] - \frac{V_{PV}+I_{PV}R_S}{R_{sh}} \tag{1}$$

Parameter n denotes ideality factor of the PV cell, while k represents Boltzmann's constant ($k = 1.3806 \times 10-23 J/K$), and T is operating temperature of PV cell. Resistances R_S and R_{Sh} correspond to series and shunt resistances, respectively. V_{PV} and I_{PV} indicate output voltage and current of PV module. Symbol q refers to elementary charge of an electron, and I_O is reverse saturation current. To overcome inherently low voltage output of PV system, a z-source modified luo converter is utilized to effectively boost the voltage level.

2-2- Z-Source Modified Luo Converter

A Z-source modified Luo converter is like to that of a switching mode boost converter behavior. The output DC voltage of this step-up power converter is higher than the input. With a relatively smaller number of components, it has a high gain. When attached in a cascade, this gain rises arithmetically step by step. The positive output Z-source modified Luo converter is shown in Fig. 3. This converter includes of input voltage (V_{PV}), input inductor L_1, capacitors(C_1, C_2 and C_3), switch S, diodes(D_1 and D_2), inductors (L_2 and L_3) output capacitor(C_O), output resistor(R_O) and output voltage(V_O).
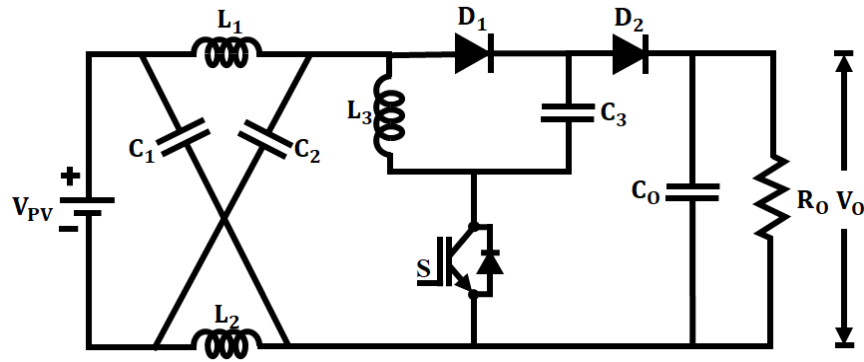
Fig. 3. Proposed converter

2-2-1- Mode 1

In this operating mode, switch S is turned ON, resulting in charging of inductor L_3 and simultaneous

discharge of capacitor C_3, as showed in Fig. 4. Diode D_1 conducts forward, though diode D_2 remains reverse

biased, leading to discharge of output capacitor C_O. During this phase, input section is electrically decoupled

from load. Inductors L_1 and L_2 accumulate energy, while capacitors C_1 and C_2 release their stored charge.

Applying Kirchhoff's Voltage Law (KVL), following relationships are established:

$$V_{PV} - V_{C1} - V_{L2} = 0 \tag{2}$$

$$V_{PV} - V_{L1} - V_{C2} = 0 \tag{3}$$

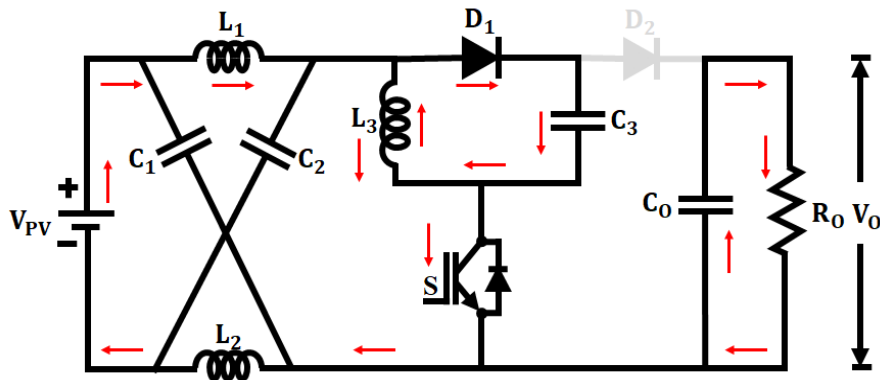$$V_{L3} - V_{C3} = 0 \tag{4}$$

$$V_{CO} = V_O \tag{5}$$

Fig. 4. Mode 1 of developed converter

2-2-2- Mode 2

In this mode, switch S is OFF thereby inductor L_3 is discharged making the capacitor C_3 charge. Here, diode D_1 is reverse biased, D_2 is forward biased and capacitor C_O is charged. Inductors L_1 and L_2 are discharged and capacitors C_1 and C_2 are charged, as depicted in Fig. 5.

$$V_{L2} - V_{C2} - V_{L3} - V_{C3} - V_{CO} = 0 \tag{6}$$

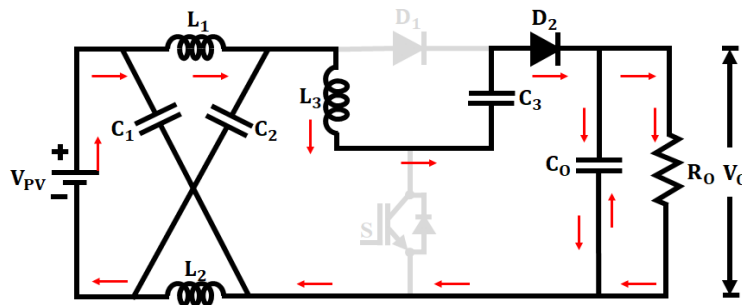$$V_{C1} - V_{L1} - V_{L3} - V_{C3} - V_{CO} = 0 \tag{7}$$

Fig. 5. Mode 2 of developed converter

Subtracting equation (6) from (5)
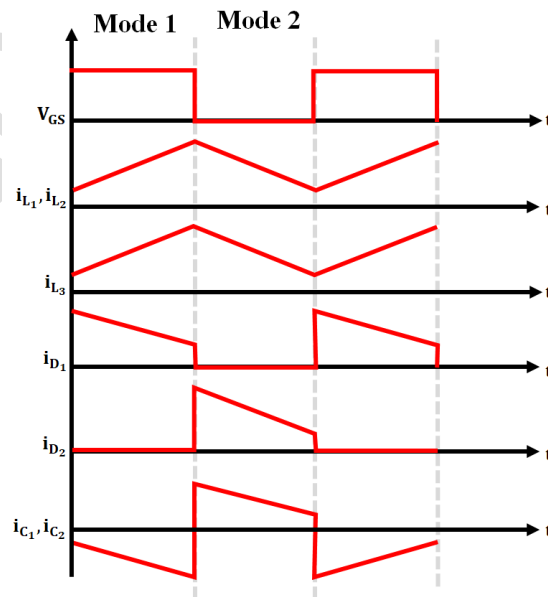
$$V_{L2} - V_{C2} - V_{C1} - V_{L1} = 0 \tag{8}$$

Fig. 6. Waveform of developed converter

Fig. 6 represents the represents timing diagram of proposed converter and its voltage gain is computed as,

$$\frac{V_o}{V_{in}} = \frac{2-k}{1-k} \tag{9}$$

Here, k represents duty ratio of the switching signal applied to the main switch S, defined as:

$$k = \frac{t_{ON}}{T} \tag{10}$$

Where, t_{ON} is time duration in which switch S is in the ON state, T is the total switching period. Then, the function of converter is managed by walrus optimized PI controller, which continued voltage of PV system.

2-3- Walrus Optimized PI Controller

Walrus optimized PI controller incorporates the Walrus optimization algorithm into the PI controller design. For optimal control performance this method leverages the unique behaviours of walrus to improve the process of tuning. This algorithm is inspired by behavioural and social pattern of walruses.

2-3-1- Feeding strategy

The food is searched by the strongest walrus, similar to detecting optimal solution in search space. Walruses migrate to rocky beaches, indicating the exploration of distinct search areas for better solutions. To refine candidate solutions these behaviors simulate local exploitation. Arbitrarily initialize a walrus's population, each indicating a candidate solution for PI controller parameters such as proportional gain $(K_{p,i})$ and integral gain $(K_{i,i})$.

$$X = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_N \end{bmatrix} \tag{11}$$

Where $X_i = [K_{p,i}, K_{i,i}]$ for the ith walrus. For each walrus, the objective function is related to performance metrics like Integral of Squared Error (ISE).

$$F_i = \int_0^T e^2(t)dt \qquad (12)$$

Where the error signal among the actual and desired output is denoted as $e(t)$. The optimization aim for tweaking PI controller parameters is specified as minimizing the Integral of Squared Error (ISE) stated in Eq. (12). To produce balanced transient and steady-state responses, further performance weighting terms are incorporated.

$$J = \int_0^T \left(w_1 e^2(t) + w_2 e^2(t)\right)dt \qquad (13)$$

Here w_1 and w_2 denote the relative weights of error magnitude and rate of change, respectively. The tuning ranges of K_p and integral gain K_i are limited to [0,5] and [0,0.05] respectively, maintaining system stability and limiting excessive overshoot. The WaOA algorithm iteratively adjusts these gains to reduce J, resulting in ideal control settings with the lowest ISE and smoothest voltage regulation.

Phase 1: Feeding strategy (Exploration)

Walruses graze by grazing on sea floor and using their sensitive vibrissae and forceful flipper motions to identify food. Benthic bivalve mollusks, particularly clams, are preferred by walruses. As tgroup searches for food, walrus with tallest tusks and strongest physique takes lead. Proposed solution with highest objective function value is the group's smartest walrus. WaOA's ability to explore in global search is improved by Walruse's search behavior, which produces results in many scanning regions of search space. A new position is created as

$$X_i^{Pl} = X_i + rand_{i,j} \cdot (X_{SW} - X_i) \cdot I_{i,j} \qquad (14)$$

Where random integer (1 or 2) is denoted by $I_{i,j}$ and strongest walrus's position is indicated by X_{SW}. The position is updated as,

$$\text{If } F_i^{P1} < F_i \text{ then } X_i = X_i^{P1} \qquad (15)$$

Phase 2: Relocating

One of walruses' typical behaviours is to migrate to rocky beaches due to late-summer air warming. This migratory strategy is used by WaOA to assist walruses in locating suitable spots inside search space. A new position is created as

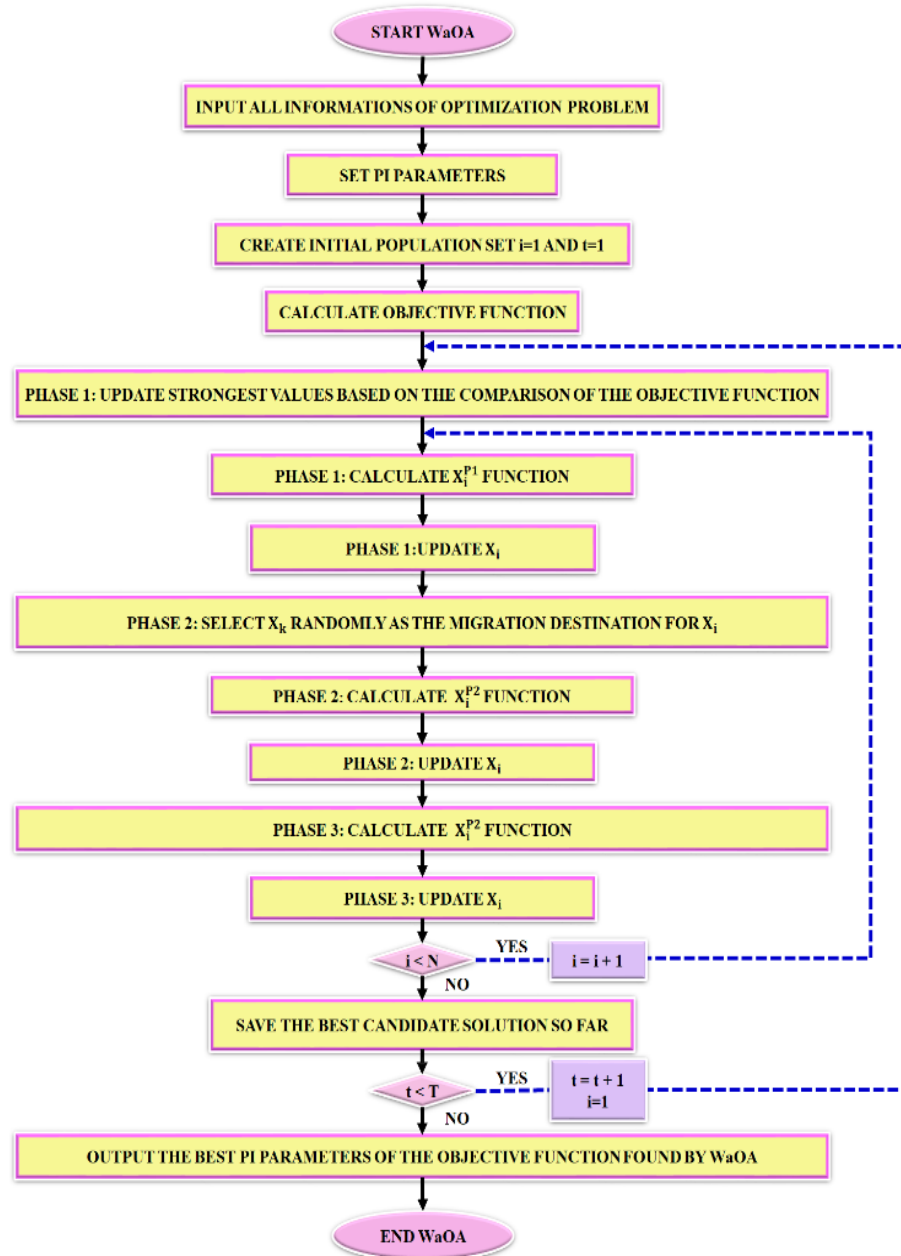$$X_i^{P2} = \frac{1}{N-1} \sum_{k=1, k \neq i}^{N} X_K \tag{16}$$

Fig. 7. Flowchart of Walrus optimized PI controller

Fig.7. shows the flowchart of proposed Walrus optimized PI controller. The position is updated as,

$$\text{If } F_i^{P2} < F_i \text{ then } X_i = X_i^{P2} \tag{17}$$

Phase 3: Exploitation

The killer whale and polar bear are constant threats to walruses. By mimicking this walrus's natural behavior, WaOA's exploitation capability in local search in problem-solving space around possible solutions is increased. New position in local neighborhood is generated as,

$$X_i^{P3} = X_i + rand_{i,j} \cdot \left(ub_j^{local} - lb_j^{local}\right) \tag{18}$$

Where jth variable's local bounds are indicated by ub_j^{local} and lb_j^{local}.

The position is updated as,

$$\text{If } F_i^{P3} < F_i \text{ then } X_i = X_i^{P3} \tag{19}$$

2-3-2- Iteration process

The first WaOA iteration is finished when the walruses' position has been updated based on the phases' implementation. New values are then computed for walruses' position and the objective functions. Using the WaOA procedures as a guide, candidate solutions are updated and improved. Until a stopping condition is satisfied, repeat steps two and three.

2-3-3- Final solution

WaOA presents the optimal candidate solution discovered during algorithm execution as solution to specified problem when algorithm has finished running. Return the best candidate solution as the PI controller's optimal parameters

$$X_{best} = \arg\min F_i \tag{20}$$

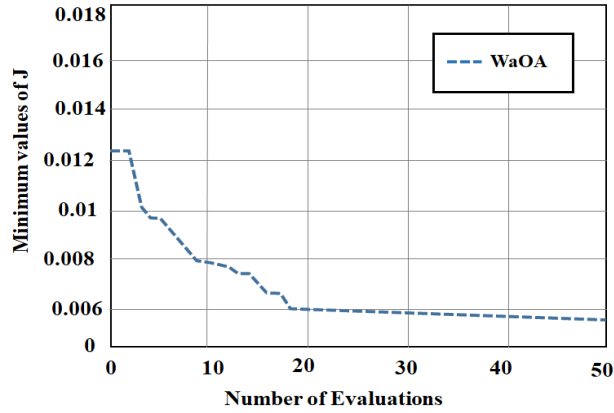The objective function value amid all walruses is denoted by F_i.

Fig. 8. Convergence curve of WaOA.

Fig. 8. Displays convergence curve of proposed optimized controller. Based on the execution of the WaOA, optimal proportional gain (K_p) and integral gain (K_i) for the developed converter were found to be K_p = 2.35 and K_i = 0.015. These values yielded the lowest Integral of Squared Error (ISE) and ensured rapid transient response with minimal overshoot and steady-state error.

By simulating walrus behaviours, walrus optimized PI controller attains a balance among exploration and exploitation, leading to optimal solutions. Another RES is Fuel cell and its voltage is improved by Boost converter.

2-4- Fuel Cell-Fed Boost Converter

A fuel cell is an energy transformation device, which directly transforms fuel chemical reactions into electrical energy with some heat, while producing water as a by-product. where $VFC \ and \ IFC$ are voltage and current of fuel cell, R_W is resistance coefficient, R_{act} stands actual value of the resistance, E_{Nernst} is the output voltage of each component, and R_{con} is the resistance concentration loss. An important component of an FC regulator is the excess ratio of oxygen, or λ_o. Equations (21) and (22) each provide an independent description of the necessary stack current I_st.

$$W_{H2,reac} = 2M_{H2} \left(nI_{st}/4F \right) \tag{21}$$

$$W_{O2,reac} = 2M_{O2}(nI_{st}/4F) \tag{22}$$

In this case, M_H stands for mass of hydrogen, M_o for mass of oxygen, F for Faraday constant, and n for number of cells. The movement of hydrogen is represented by $W_{H2,reac}$ and the stack is denoted by I_{st}. Equation (23) states that in order to meet the current demand, a connection between flow of oxygen and stack is required.

$$\lambda_{o2} = {W_{O2,ca,in}} \Big/ {W_{O2,reac}} \tag{23}$$

Equation (20) is the sole basis for oxygen excess ratio, which is expressed as λ_{02}. Flow of oxygen disbursed is characterized by $W_{O2,reac}$, and movement of cathode oxygen is shown by $W_{O2,ca,in}$. Equation (24) represents the steady-state equation:

$$V_{FC} = A - B.\ln\left(1 + \frac{i_{FC}}{C}\right) - D.e^{\frac{i_{FC}}{E}} \tag{24}$$

Here $A, B, C, and\ D$ tuned parameters and stack current of FC is expressed as I_{FC}. Fig. 9 depicts the traditional boost converter topology. Fig. 10 depict the boost converter's two modes of operation. The boost converter, which is needed for low-to-medium power levels, is an easy and sufficient solution for applications where extremely high voltage step-up ratios are not required. The direct current input of converter is powered by DC sources like fuel cells. In this system, the input voltage is raised by means of three primary components: an inductor, a diode, and a power switch. Boost converter's output is coupled to a capacitor that reduces and stabilizes output voltage fluctuations. Lastly, an output capacitor is connected to a load in form of adjustable resistors.
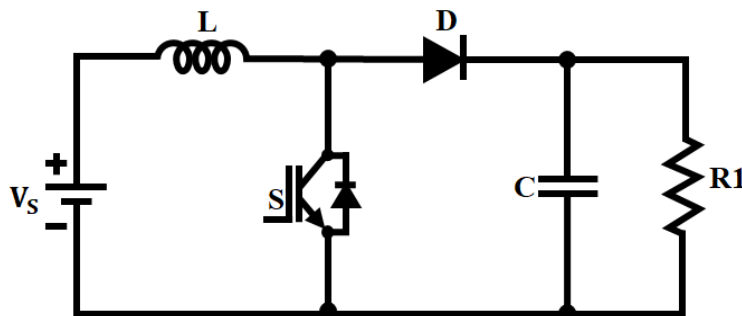
Fig. 9. Boost converter

In stage1, the switch S is active, causing in improving inductor current. In this stage the expression becomes,

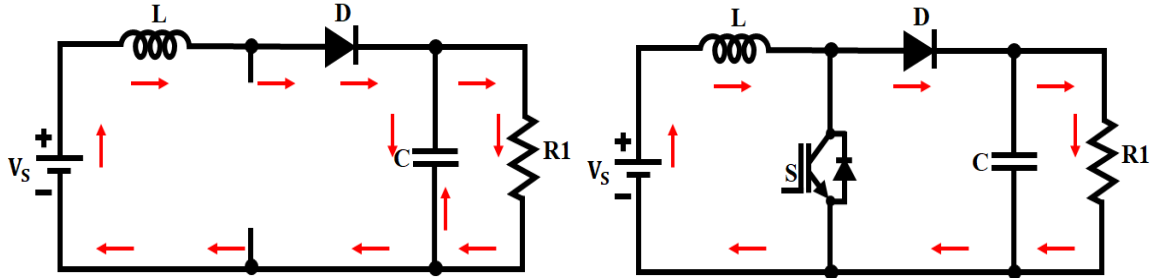$$\Delta_{i\ lON} = \frac{DT}{L} V_S \tag{25}$$

Fig. 10. Operating Modes of Boost converter

In stage 2, switch S is inactive, the inductor current passes via load and the expression becomes,

$$\Delta_{i\ lOFF} = \frac{(V_S - V_O)(1-D)T}{L} \tag{26}$$

By utilizing the CCM operation and volt-second balance principle, the static gain is expressed as,

$$G_v = \frac{V_o}{V_i} = \frac{1}{1-D} \cdot \frac{1}{\left(1 + (R_L/((1-D)^2 \cdot R_o))\right)} \tag{27}$$

Here, the fuel cell based Boost converter provides additional power to charge EVs. Also, the battery with bidirectional converter is exploited for EV charging when there is no power available from HRES.

2-5- Battery System with Bidirectional Converter

A bidirectional converter in the battery system depicted in Fig. 11 enables both power mode operation while the battery is being charged and regeneration mode operation when the battery is being discharged. Power transfers from DC link to battery during battery bank charging and back to DC link during battery bank discharge. When the inductor is charged, diode D_2 reverse bias causes S_1 is in the OFF state and S_2 is in the ON state. When both switches are in ON state, the diode D_2 is forward biased.
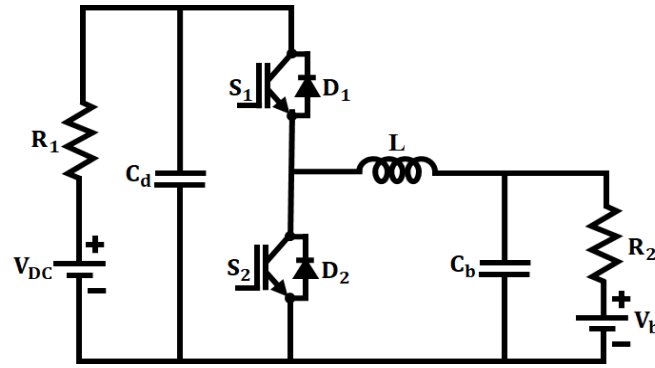
Fig. 11. Bidirectional converter

Similarly, when S_1 is in ON state and S_2 is in the OFF state during discharging, the inductor's current drops to zero and polarity charges. Diode D_1 turns forward bias when S_1 and S_2 are in the ON state. However, achieving a stable voltage requires an efficient controller, in which an ANN controller is exploited in this work to monitor the battery's state of charge.

2-6- ANN Controller

Several factors that are pertinent to the control procedure are usually included in the input signals sent to an ANN controller for managing switching power signals from a battery bank. The proposed ANN controller employs a feed forward neural network architecture with two hidden layers, each comprises of 10 neurons. The sigmoid activation function is utilized in both hidden layers for ensuring non-linear mapping, while the output layer integrates a linear activation function for generating precise switching control signals. The network is trained using the backpropagation algorithm with gradient descent optimization, ensuring convergence and adaptability to dynamic operating condition The Levenberg-Marquardt back propagation algorithm is applied, with a learning rate of 0.01 and a momentum coefficient of 0.9.s. The input signals are battery voltage, current, SOC, temperature, load demand, and perhaps other pertinent factors. These inputs are normalized and fed into the ANN to evaluate the system's operational state. Switching control signals for power switches connected to the battery bank, determining ON/OFF states to optimize energy flow and maintain grid stability. The fundamental neuron model used in ANN controller is,

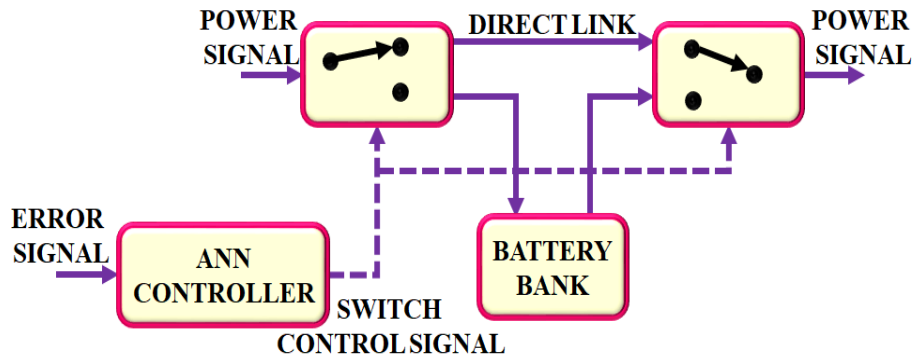$$\alpha = \left(\sum_{j=1}^{n} w_j u_j + \Phi \right) \tag{28}$$

Fig. 12. Energy management using ANN controller

Fig. 12 depicts the basic design of ANN controller. This ANN-assisted control approach enhances battery reliability, reduces switching losses, and ensures continuous grid support under varying load and environmental conditions.

3. Result and Discussions

Waveforms of proposed hybrid energy system for EV charging stations is examined in this section. Proposed work is executed in MATLAB/Simulink, and its proficiency is demonstrated by comparison with traditional methods. Parameters specifications for proposed research are specified in Table 1.

Table 1. Specification of parameters

Parameter	Specification
PV System	
Rated Power	10kW
Open Circuit Voltage	37.25V
Short Circuit Current	8.95A
No. of Panels in Parallel	2
No. of Panels in series	21
Temperature	$35\,°C$
Irradiance	$1000\ W/m^2$
Z-source modified Luo Converter	
Switching frequency	10KHz
Input Voltage	230V
Output Voltage	600V
L_1, L_2, L_3	4.7mH
C_1, C_2, C_3	22µF
C_O	2200µF
Fuel cell	

| Maximum Operating Point [$I_{end}(A)$]	33.2 A
Maximum Operating Point [$V_{end}(V)$]	104.86 V
No. of cells	119
Nominal supply pressure (Air[bar])	1 bar
Nominal composition (%) [$H_2, O_2 H_2 O$]	[50, 21, 3] %
Nominal Operating Point [$I_{nom}(A)$]	30 A
Nominal Operating Point [$V_{nom}(V)$]	106.5 V
Operating temperature (Celsius)	600 °C
Nominal Air flow (Ipm)	635 L min⁻¹
Nominal supply pressure (fuel[bar])	1.35 bar
Battery (Li-ion)	
SOC	90%
Nominal Voltage	60V
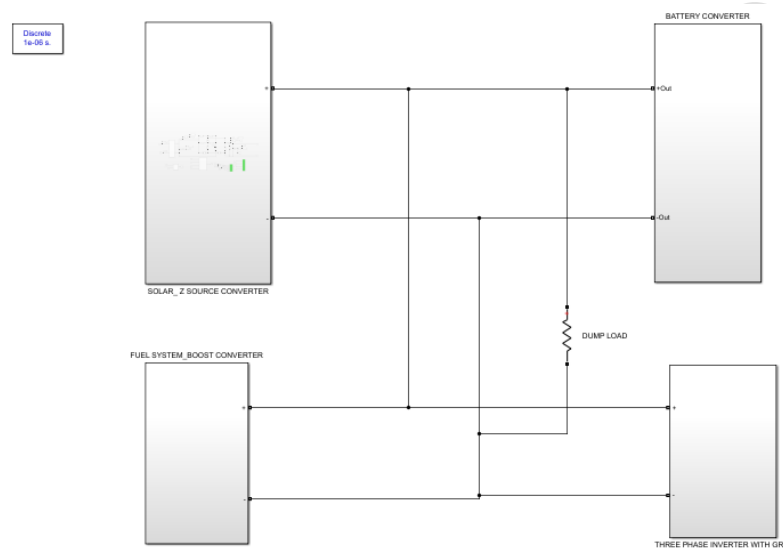
Fig. 13. Simulink model of proposed hybrid PV-Fuel Cell EV charging system.

Fig. 13 system includes a solar Z-source modified Luo converter and a fuel cell boost converter, whose outputs are combined to power the DC link. A bidirectional battery converter controls charging and discharging operations, and surplus energy is released via a dump load. The three-phase inverter connects the DC link to the utility grid, allowing bidirectional power exchange while ensuring grid stability.
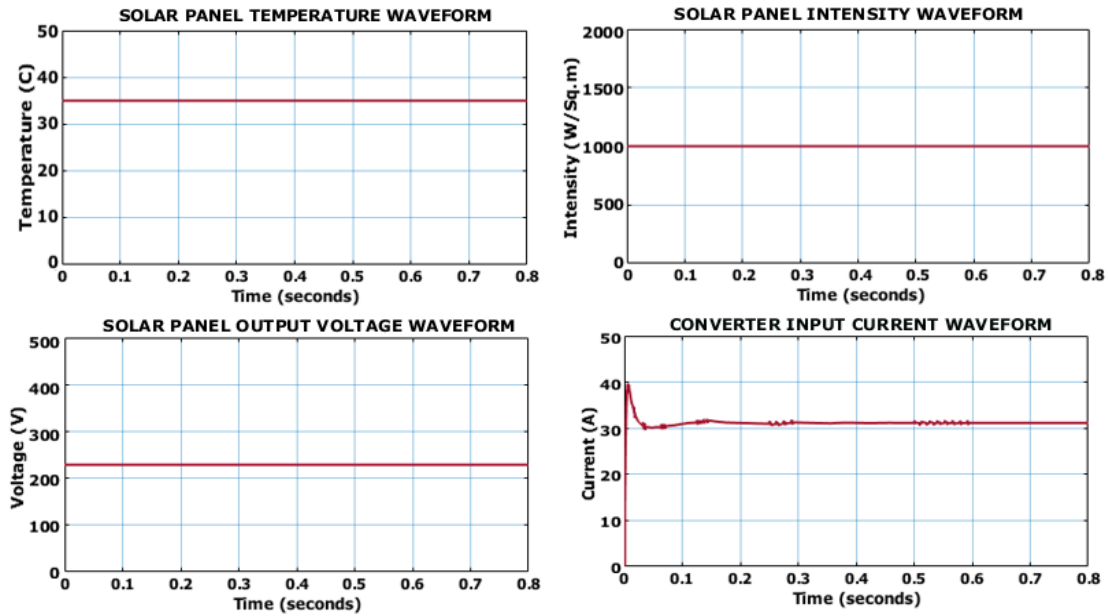
CASE 1: Constant temperature and intensity

Fig. 14. PV panel waveforms.

Waveform for solar panel's properties under conditions of constant temperature and intensity is presented in Fig. 14. System as a whole maintains solar panel's temperature at a constant 35°C. Additionally, solar panel maintains a consistent intensity value of $1000 \, W/m^2$ throughout the system. After that, solar panel's voltage is maintained throughout system at 230V. Also, output current of solar panel is initially varied and sustained at a value of $32A$ after an initial transient fluctuation of ±0.8 A. Based on this values, converter's output are varied, which is discussed below.
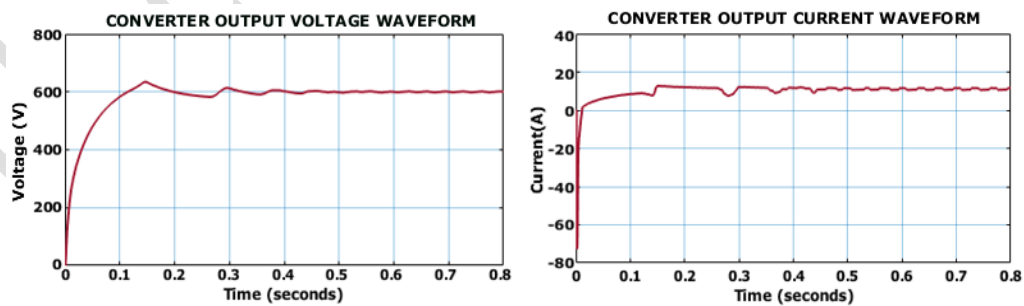

Fig. 15. Converter output waveform under case 1

Waveform for developed converter is represented in Fig. 15. The converter's output voltage continuous at 600 V, with a maximum transient fluctuation of ±2 V (0.33%). At first, the converter's output current is varied and maintained to stable value of 10A. Under constant environmental circumstances, indicate strong dynamic stability and quick voltage regulation.
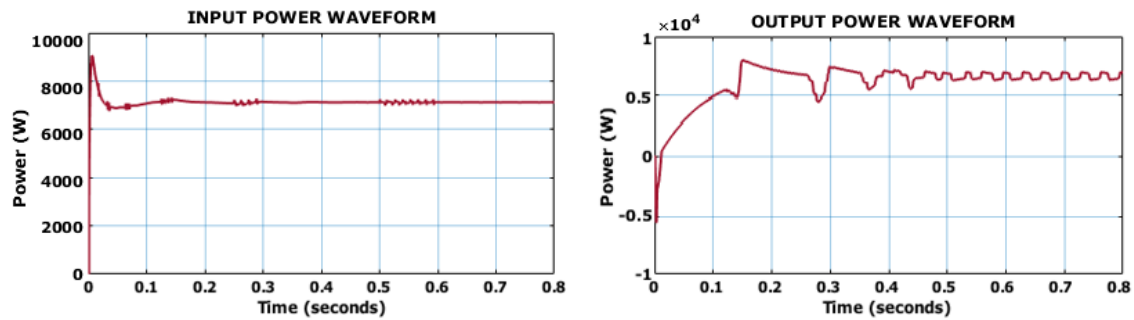

Fig. 16. Waveform of power at case 1

The waveform for input and output power is indicated in Fig. 16. Input power fluctuates during the transient period before stabilizing at an average of 7000 W with a variance of ±90 W. The output power is stable at $6.8 \times 10^3\ W$, with slight variations within ±120 W.
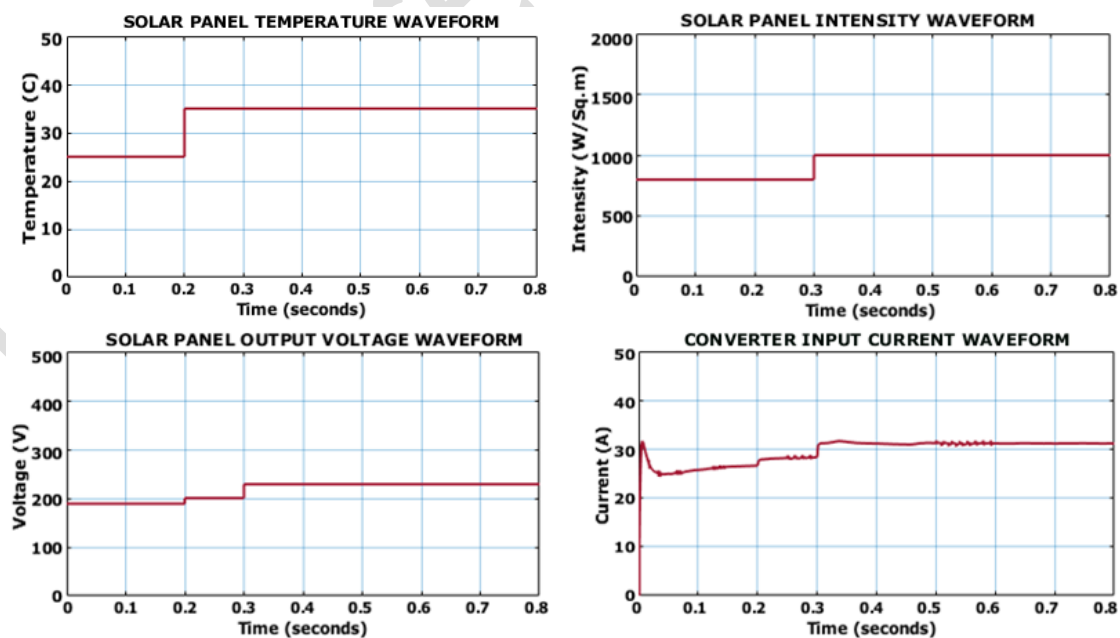
Varying temperature and intensity

Fig. 17. PV system waveforms.

The waveform of solar panel in changing temperature and intensity is indicated in Fig. 17. When irradiance ranges from $800 - 1000 \ W/m^2$ and temperature ranges from $25 - 40 \ °C$, the PV output voltage averages 230 V with ±3 V (1.3%) variations, and the current ranges from 8.5 A ± 0.25 A. These low-magnitude oscillations demonstrate that the control strategy efficiently reduces transient impacts induced by environmental changes.
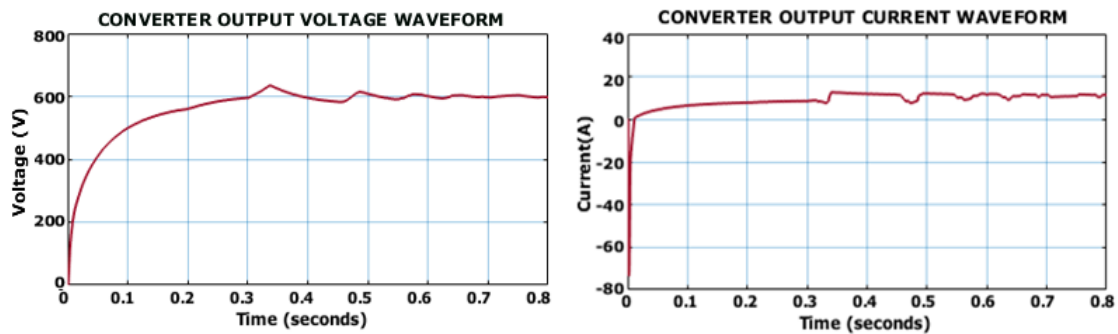

Fig. 18. Output waveform of proposed converter.

Fig. 18 signifies proposed converter output voltage and output current waveform. Throughout the transient interval, the converter's output voltage rises and stabilizes at 600 V with a maximum variance of ±2.5 V (0.4%). The output current stabilizes at 10 A with oscillations of ±0.3 A during steady state. These findings show the converter's fast settling reaction and effective voltage regulation in dynamic operation situations.
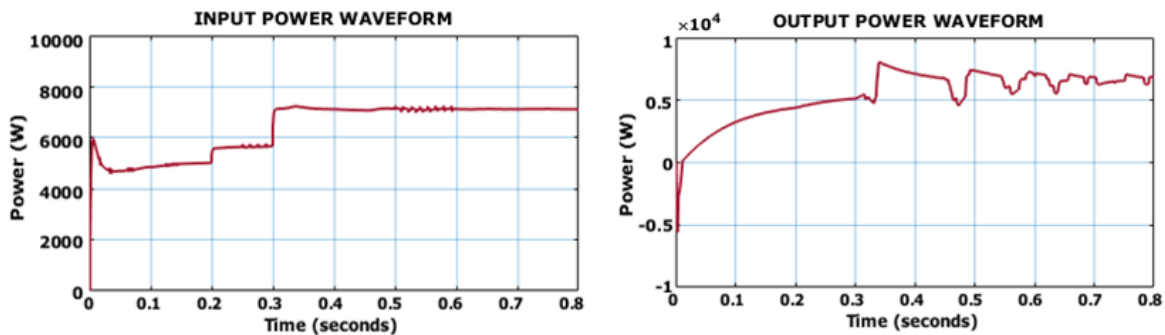

Fig. 19. Power Waveform at case 2

Under variable temperature and intensity, the power waveforms shown in Fig. 19. During the initial transients, the input power swings before settling at an average of 7000 W, with a variance of ±100 W

(1.4%). The output power averages 6.8×10^4 W, with minor variations of ±150 W (2.2%). These findings demonstrate that the suggested converter maintains high power stability and little fluctuation despite dynamic environmental disturbances.
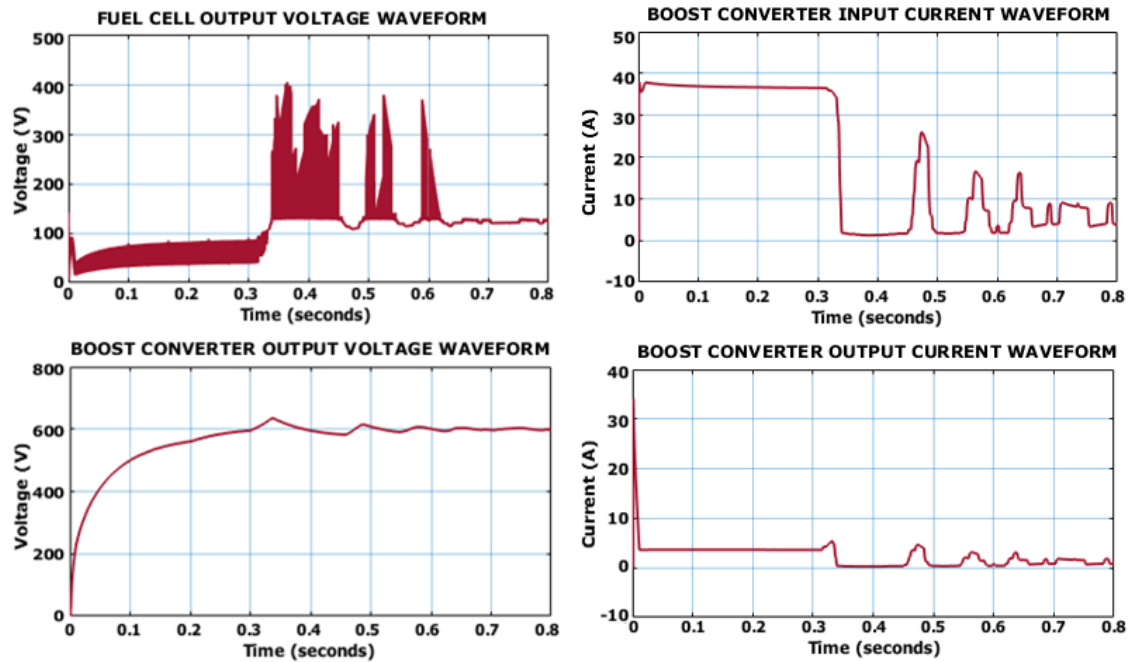

Fig. 20. Waveforms of Fuel Cell with Boost converter output

The outcomes of fuel cell with converter output is shown in Fig. 20. Initially, the fuel cell voltage fluctuates somewhat before settling at 110 V, with a variance of ±2 V. The boost converter's input current peaks during starting and stabilizes at 5 A, with fluctuations within ±0.15. Based on the fuel cell's input voltage, the output voltage is enhanced with Boost converter with the value of $600V$. Accordingly, the output current is stabilized in the beginning and finally has some fluctuations with a value of 2A.
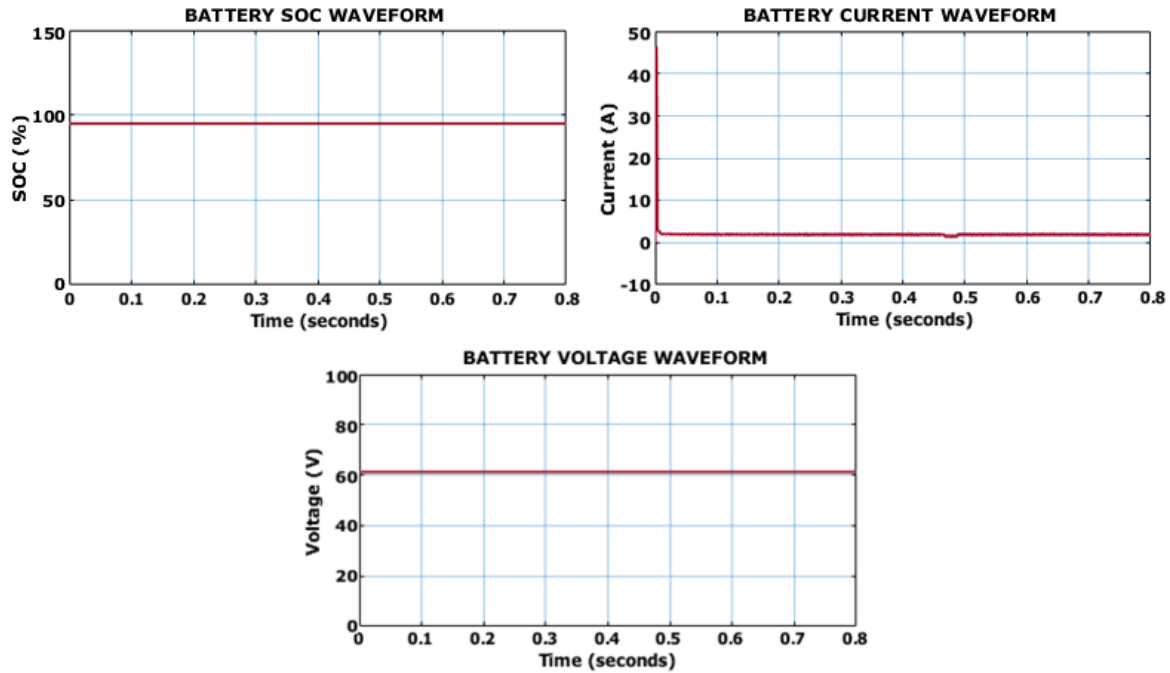
Fig. 21. Waveform of battery

The battery characteristics waveforms in Fig. 21 illustrate the effectiveness of control strategy in regulation. The SOC waveform remains constant at 90% during simulation period, ensures representing that battery is in a high state of charge and held in reserve for potential energy exchange. The current waveform shows an initial transient spike that peaks close to 40A, which is likely due to load fluctuation. After this transient period, the current stabilizes near 3A. Similarly, the voltage waveform remains steady at approximately 60 V over the entire duration, demonstrating effective voltage regulation by the ANN controller.
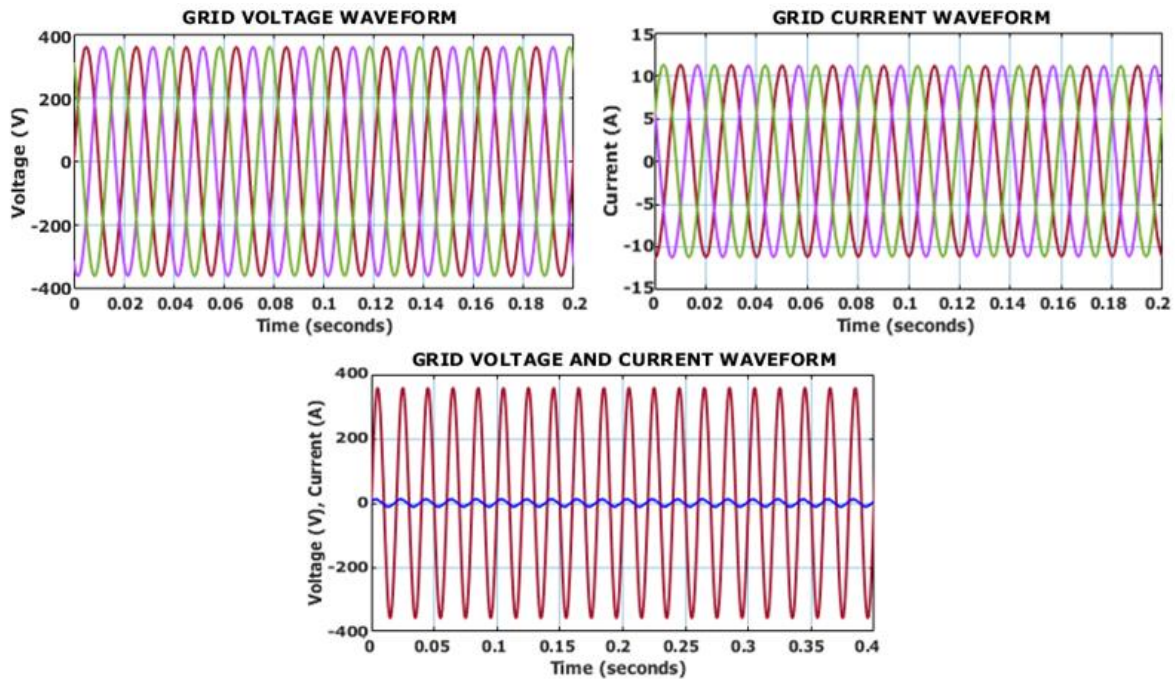
Fig. 22. Waveform of grid

Fig. 22 indicates waveform of grid. The voltage of grid is conserved a stable value of $350\,V$ during entire period. Consequently, grid current is continued to stable value of $12\,A$. Also, in-phase grid voltage and current is continued to value of $350\,V$. It denotes that current and voltage in the grid is controlled by PI controller to continue a stable and constant level. This confirms the grid-tied nature of the system, with stable voltage and current profiles indicating continuous synchronization and support from the grid, characteristic of a grid-connected microgrid.
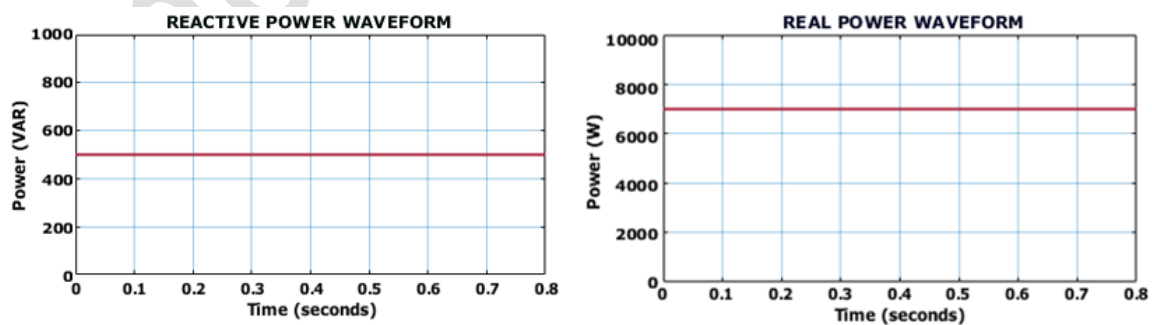

Fig. 23. Waveform of real and reactive power

Real and reactive power waveform are illustrated in Fig. 23. Reactive power consumption is decreased, which makes it possible to use active power in the system more effectively, which raises total power efficiency and improves real power performance.
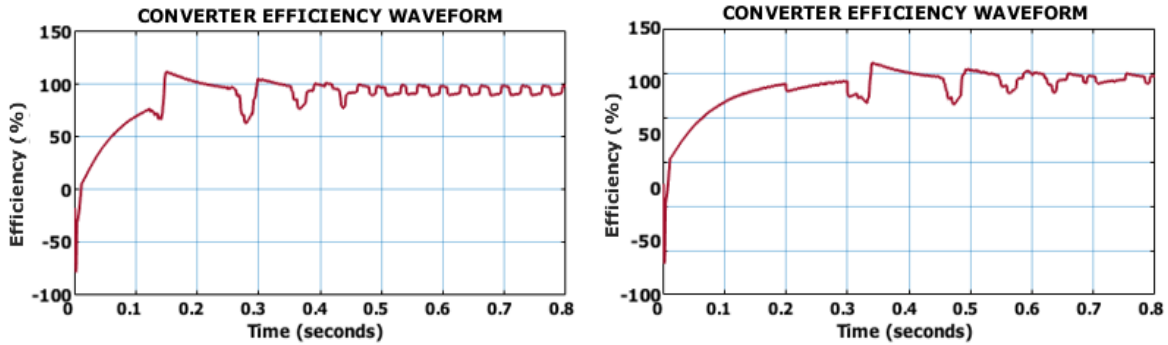

Fig. 24. Waveform of converter efficiency under constant and varying condition

Figure 24 shows converter maintains a consistent efficiency of 97.14%, under constant and varying conditions. The overall converter efficiency is calculated using the MATLAB/Simulink model's total DC output power to total DC input power ratio. It is described as,

$$\eta = \frac{P_{out}}{P_{in}} \times 100$$

The average input voltage and current are represented by V_{in} and I_{in}, correspondingly. The Z-source modified Luo converter's output quantities are denoted by V_{out} and I_{out}. The average results were obtained once the system had reached steady-state functioning. The computed efficiency of 97.14% indicates the steady-state power conversion efficiency, which includes conduction and switching losses, as assessed using MATLAB/Simulink simulation data.
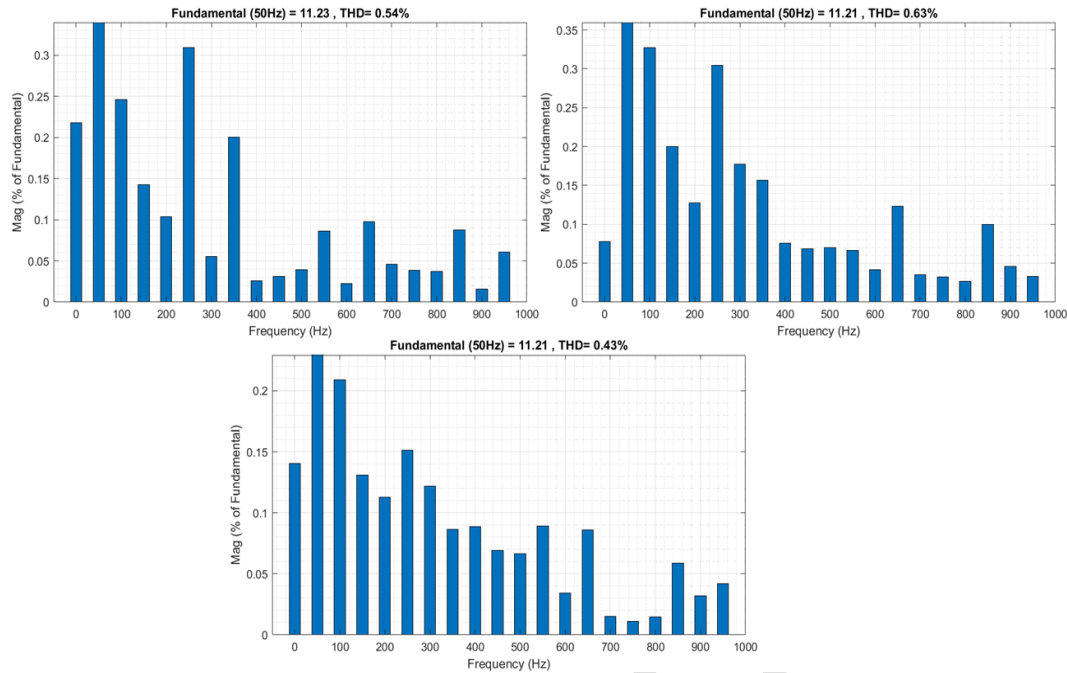
Fig. 25. Waveform of THD

The waveform of Total Harmonic Distortion (THD) spectra for R, Y and B phases are illustrated in Fig. 25, showing respective THD values of 0.54%, 0.63%, and 0.43%. Among them, the B phase exhibits the lower distortion level, signifying better power quality compared to the R and Y phases.

Fig. 26. Analysis of efficiency

Fig. 26. illustrates the comparative efficiency analysis of various converter topologies, such as High Step-Up Switched Z-Source Converter (HS-SZSC) [26], Trans-quasi-Z-source (TQZS) Boost [27], KY [28],

Super-lift Luo and Boost Converter (SLBC) [29], SEPIC-Luo [30], and the proposed converter. The respective efficiency values recorded are 83.60%, 92%, 93%, 94.80%, 96%, and 97.14%. Among these, proposed converter validates enhanced performance in effi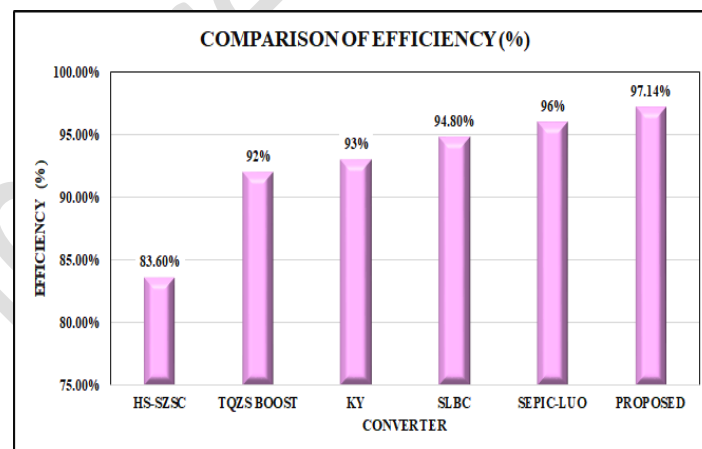ciency, overtaking all existing designs. It clearly validates its superior energy conversion capability and optimized performance under the tested conditions.

Table.2 Comparative analysis of converter with control topologies.

References	[26]	[30]	Proposed
System Components	PV, Grid.	PV, Wind, Grid, Battery, EV Load	PV, FC, Grid, Battery.
Converter Topology	HS-SZSC	SEPIC-Luo	Z-Source Modified Luo Converter
Control Algorithm	PI Controller	PI Controller	Walrus Optimized PI controller
Peak Efficiency (%)	83.60%,	96%	97.14%
Voltage Gain (G)	Moderate	High	Very High
Input Voltage (V)	35V	120V	230V
Output Power (W)	1200W	4500W	6800W
Switching Frequency (KHz)	30KHz	100KHz	10KHz

Table.2 presents the comparative study of different converter with control approaches highlighting their system configurations, peak efficiencies, voltage gain capabilities, input voltage, output power and switching frequency. It demonstrates that the proposed converter with walrus optimized PI controller outperforms existing designs by achieving the highest efficiency (97.14%) and voltage gain, while maintaining a compact architecture with minimal storage dependency.
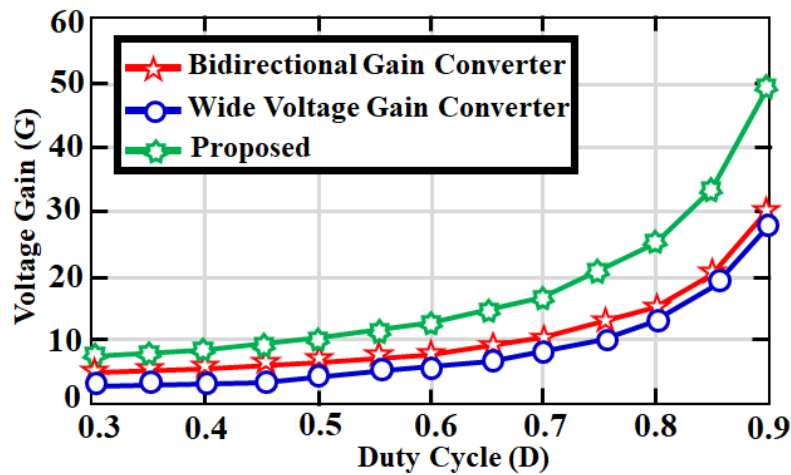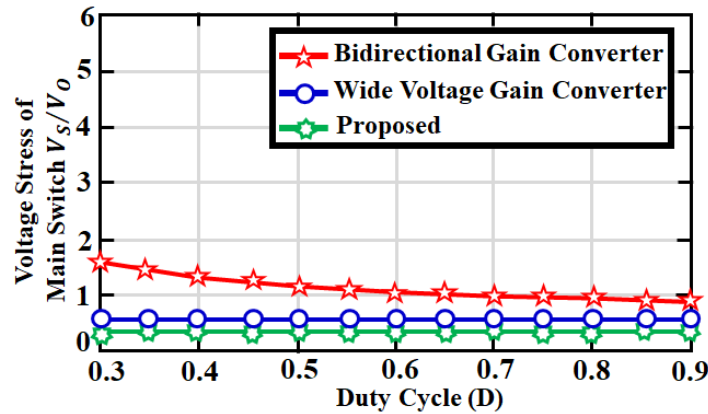

Fig. 27. Analysis of voltage gain

Fig. 28. Analysis of voltage stress on switch

Fig. 27 illustrates the analysis of voltage gain for bidirectional DC-DC [31], wide voltage gain [32] and developed converter. Proposed converter has highest voltage gain than conventional converters, indicating its maximum efficiency. The voltage stress between bidirectional DC-DC [31], wide voltage gain [32] and proposed converter is shown in Fig. 28, it has lesser voltage stress than existing converter topologies, which enhances converter efficiency.

To ensure an impartial assessment of performance, the PSO-PI and GWO-PI controllers are developed in MATLAB/Simulink under similar converter and load conditions. The controller parameters and tuning approaches are taken from [33] and [34], respectively, and used the standard gain selection and convergence settings. These reference models offer valid baseline performance indicators for testing the proposed Walrus-optimized PI controller's settling and rising times.

Table 3. Analysis of PI controllers

Approaches	Settling Time (s)	Rising Time (s)
PSO-PI [33]	0.0075 s	0.00815 s
GWO-PI [34]	0.00406 s	0.000228 s
PROPOSED	0.00089 s	0.00013 s

Table 3 represents the analysis of optimized-PI controllers for PSO [33], GWO [34] and developed algorithm. The settling time for developed Walrus optimized PI controller is 0.00089 s, which is better than settling time of PSO-PI (0.0075s) and GWO-PI (0.00406s). Also, the rising time for PSO-PI, GWO-PI and

developed method is 0.00815 s, 0.000228 s and 0.00013 s. The developed method attains the better performance interms of rising time than listed approaches.

4. Conclusion

This research develops a novel hybrid energy system for EV charging station using PV fed Z-source modified Luo converter and fuel cell based Boost converter. The PV system's low voltage is enhanced by Z-source modified Luo converter with higher efficiency. Furthermore, the Walrus optimized PI controller manages the performance of converter with stabilized voltage. Also, the FC based Boost converter with PI controller is employed to charge the electric vehicles. The battery with bidirectional converter is exploited to provide continuous power supply to Evs in the necessary period. The overall research is implemented in MATLAB/Simulink and outcomes demonstrate that developed method provides higher efficiency of 97.14%, ensures the constant power supply to EVs.

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