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A neural network-based adaptive power-sharing strategy for hybrid frame inverters in a microgrid

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The capacitive-coupling inverter (CCI) is more cost-effective in reactive power conditioning and enhanced reactive power regulation ability when compared with the inductive-coupling inverter (ICI). As power conditioning capability is vital for a microgrid (MG) system, a new MG frame with hybrid parallel-connected ICIs and CCIs was proposed in this paper. With lower DC-link voltage for the CCI, an adaptive power sharing method was proposed for reducing total rated power and losses. A power-sharing control layer based on a back-propagation neural network that guarantees rapid and accurate sharing ratio computation was investigated as well. The results of simulations and experiments were used to verify the effectiveness of the proposed method.

KEYWORDS

microgrid, capacitive-coupling inverter, unequal power sharing, power capacity, BPNN

1 Introduction

Distributed generation (DG), which offers distinct advantages, including lower transmission losses, greater power control flexibility, high energy efficiency and lower pollution dissipation, is playing an important role in the development of next-generation power grids (Deng et al., 2021; Liu et al., 2019a). A microgrid (MG) is a self-sufficient energy system based on DG that serves a discrete geographic footprint, such as a college campus, business center or neighborhood. Ensuring a stable point-of-common-coupling (PCC) voltage in a MG is vital, and requires that the DG unit have a wide reactive power control capability to cope with load fluctuations through its inverter interface (Ahmed et al., 2021; Murty and Kumar, 2022).

With large active power transfer capacity, the inductive-coupling inverter (ICI) is used in most DG applications (Zhang et al., 2019). However, the reactive power compensation capability of the ICI relies on a DC-link voltage that is higher than the PCC voltage, resulting in high conversion losses and increased operational breakdown risks (Deng et al., 2020). The reactive power compensation capacity of ICI is also limited by the output capacity of the DG system (Ziyi et al., 2023); therefore, additional reactive compensation

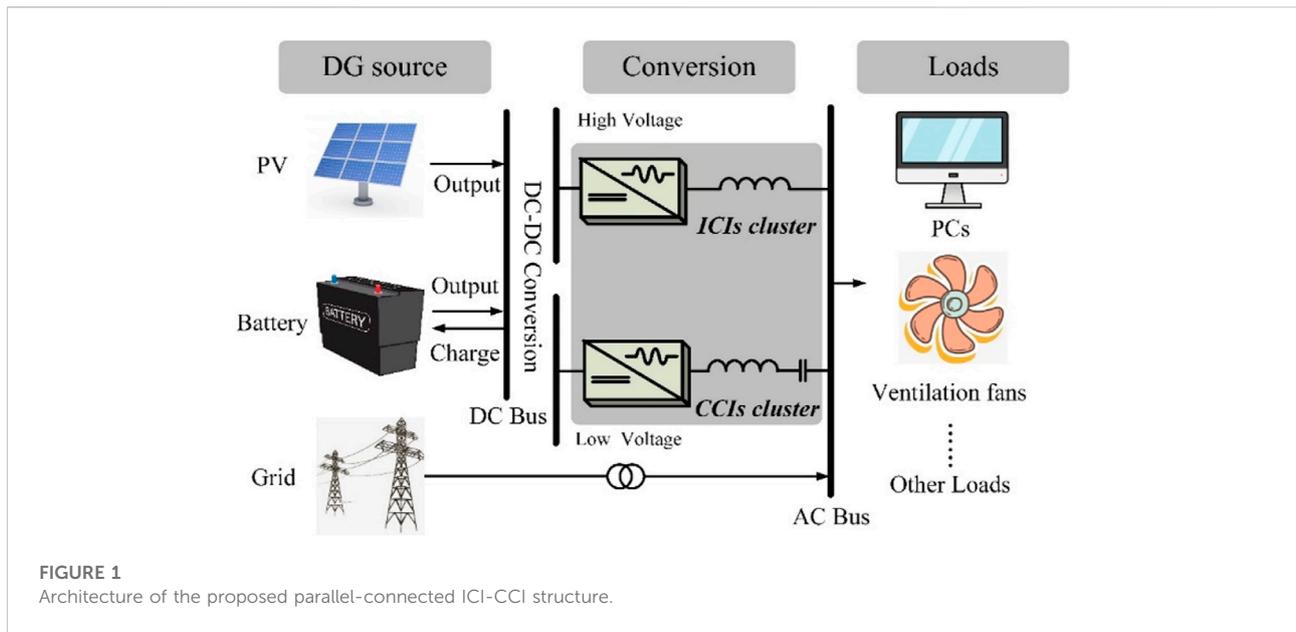


TABLE 1 Comparison of different inverters.

	P output	Q compensation	DC requirement
CCI	The coupling impedance is large, and the active output range is narrow at the same DC voltage.	At the same DC voltage, the reactive power compensation range for inductive loads is wide.	When the DC voltage is lower than the double bus voltage, the active output and reactive power compensation for inductive loads can be realized.
ICI	The coupling impedance is small, and the active output range is wider at the same DC voltage.	At the same DC voltage, the reactive power compensation range for inductive loads is narrow.	When the DC voltage is lower than the double bus voltage, reactive power compensation for inductive loads cannot be realized.

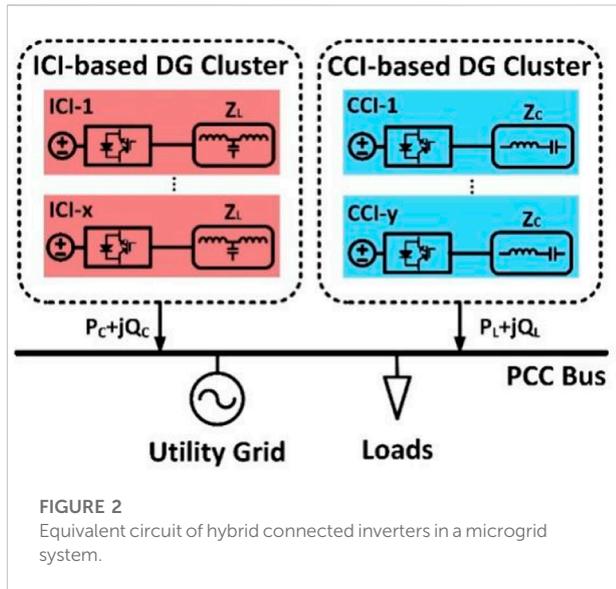
devices, such as the static VAR compensator (SVC), are necessary for an ICI-based MG system, leading to increased investment cost.

In order to have enhanced reactive power compensation capability with reduced conversion loss and cost, as well as increased control flexibility and operational stability, a capacitive-coupling inverter (CCI) has been proposed by researchers (Pang et al., 2022). It was first presented in the form of a reactive power compensator and then, was studied as a standalone generation unit (Fujita and Akagi, 1991; Deng et al., 2019; Gong et al., 2021). Because the capacitor was series-connected, the CCI had a better reactive power control capability with a lower DC-link voltage than the ICI (Sou et al., 2022). To fully utilize the power regulation capabilities of both ICI and CCI, and keep the conversion loss and equipment cost low, a new hybrid consisting of parallel-connected ICIs and CCIs was proposed in this paper (Figure 1). A comparison of CCI and ICI is presented in Table 1.

For the proposed MG system, coordinative operation of the inverters is required to regulate the DG output according to changes in load demand (Yazdani et al., 2020). Power-sharing

control plays a pivotal role in the inverter’s coordinative operation, and this was investigated by (Mousazadeh Mousavi et al., 2018; Qi et al., 2020). Droop control is simple and widely used since it can regulate inverters locally and achieve power sharing automatically. However, the controller characteristics were easily affected by mismatched feeder impedance, unbalanced loads, and other factors (Mousazadeh Mousavi et al., 2018; Liu et al., 2019b; Razi et al., 2020). For this reason, modified droop control strategies were proposed. To compensate for differences in feeder impedance, a virtual-impedance was devised (Mousazadeh Mousavi et al., 2018)–(Razi et al., 2020). Using specifically designed parameters, the virtual-impedance could also tune the resistive and capacitive feeder impedances and maintain satisfactory power-sharing (Liu et al., 2019b)–(He et al., 2019). Furthermore, by revising the control algorithm, droop control could be enhanced to compensate for inaccurate power-sharing (Lao et al., 2019).

The modified controllers are only for the ICI, and are inappropriate for the proposed MG system. This is first because of the differences in DC-link voltage, which means that the power transfer range was not the same for the CCI and ICI, and the



targeted power may surpass the control range of the CCI or the ICI with conventional equal power-sharing control. Secondly, the equal power-sharing ratio was not optimized for inverter total power reduction. For these reasons, a neural network (NN)-based adaptive power sharing scheme was proposed in order to calculate the optimal power-sharing ratio and maintain minimum power capacity. The main contributions of this work include:

- Proposal of a new hybrid system of parallel-connected ICIs and CCIs for better power control flexibility of a microgrid.
- Investigation of the effect of power-sharing ratio on the inverters' power capacity reduction and its arithmetic computation methods.
- Proposal of a power-sharing controller based on an artificial neural network (NN), to achieve fast, accurate, adaptive power-sharing.
- The principles underlying the NN model applied to the proposed method were studied and evaluated.

This paper is organized as follows. Section 2 introduces the power delivery characteristics of ICIs and CCIs. In Section 3, power flow equations are provided and we investigate how different power sharing ratios affect the proposed MG system's performance. Section 4 presents the power-ratio restraints and computations based on minimum system cost. A power sharing controller based on an artificial neural network model is also presented in this section. The results of simulations and experiments are given in Section 5 to verify the effectiveness of the proposed control method. The conclusions and implications for adoption are presented in Section 6.

2 Hybrid connected inverters in the MG

2.1 Modelling of the hybrid connected inverters

The equivalent circuit of the hybrid connected inverters is shown in Figure 2. ICI-based and CCI-based DG units are coupled to the power grid with different DC sources. The general power flow from the DG unit to the power grid can be expressed as:

$$\begin{cases} P = \frac{VV_{pcc}}{Z} \cos(\theta - \delta) - \frac{V_{pcc}^2}{Z} \cos \theta \\ Q = \frac{VV_{pcc}}{Z} \sin(\theta - \delta) - \frac{V_{pcc}^2}{Z} \sin \theta \end{cases} \quad (1)$$

where V_{pcc} and V are the magnitudes of the voltage of the power grid and inverter, and δ is the phase difference between the voltage of the inverter and the power grid. Z and θ are the magnitude and phase angle of the coupled impedance. With different coupling structures for ICI and CCI, Z and θ are also different as shown in Eqs 2, 3:

$$\begin{cases} Z_L = \omega L \\ \theta_L = 90^\circ \end{cases} \quad (2)$$

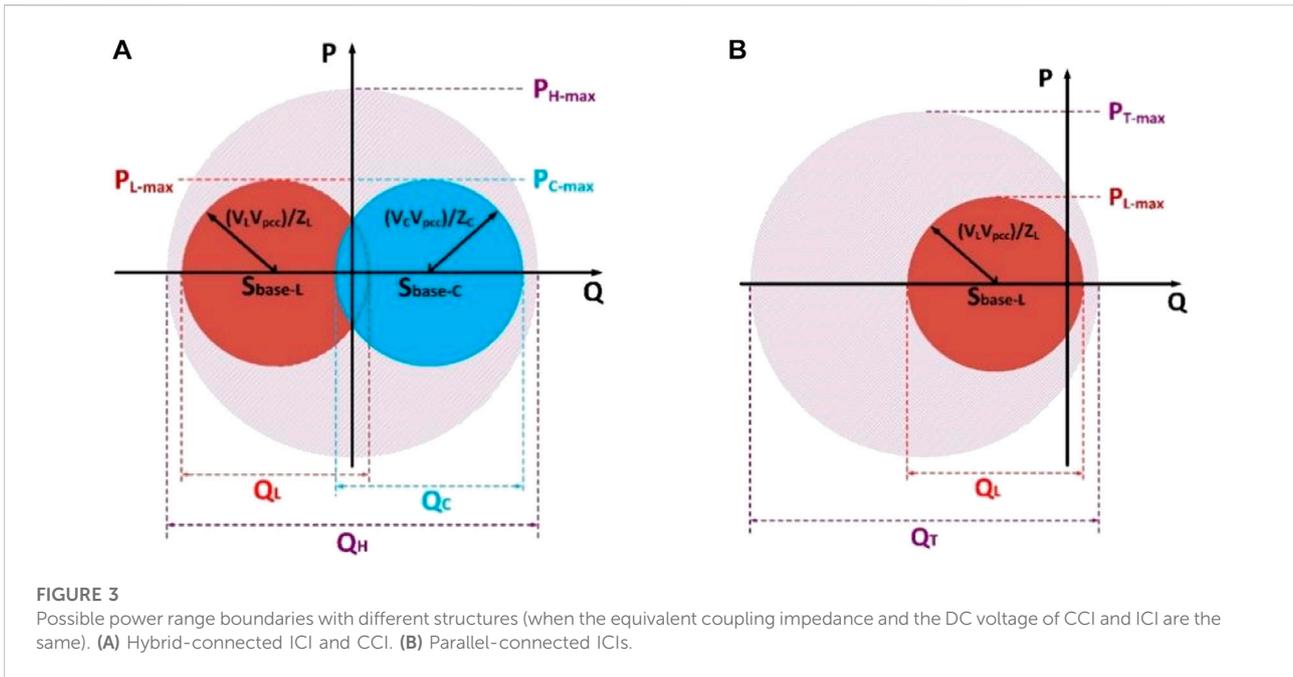
$$\begin{cases} |Z_C| = \frac{1}{\omega C_C} = \frac{1}{\omega C} - \omega L_C \\ \theta_C = -90^\circ \end{cases} \quad (3)$$

To generalize the power flow, a power base is defined as:

$$\begin{cases} S_{base-L} = \frac{V_{pcc}^2}{Z_L} \\ S_{base-C} = \frac{V_{pcc}^2}{|Z_C|} \end{cases} \quad (4)$$

By standardizing all ICIs as one ICI unit and all CCIs as one CCI unit, the power output of the inverters in Figure 2 can be expressed as:

$$\begin{bmatrix} P_L \\ Q_L \\ P_C \\ Q_C \end{bmatrix} = \begin{bmatrix} \frac{\sin \delta}{V_{pcc}} S_{base-L} & 0 \\ \frac{\cos \delta}{V_{pcc}} S_{base-L} & 0 \\ 0 & \frac{\sin \delta}{V_{pcc}} S_{base-C} \\ 0 & -\frac{\cos \delta}{V_{pcc}} S_{base-C} \end{bmatrix} \begin{bmatrix} V_L \\ V_C \end{bmatrix} + \begin{bmatrix} 0 \\ -S_{base-L} \\ 0 \\ S_{base-C} \end{bmatrix} \quad (5)$$



2.2 Power control range

Usually, there are three factors that affect the power control range of the inverter: the property of the coupling impedance (inductive or capacitive), the output voltage, and the equivalent value of the coupling impedance. The following discussion focuses on these three aspects in describing the advantages of the proposed MG system. Since the output voltage is proportional to the DC-link voltage, these two voltages are not distinguished in the following discussion.

The main difference between ICI and CCI is the property of the coupling impedance, which leads to different power delivery characteristics. Assuming the direction of active power injection and reactive power compensation is positive, the power base and the DC-link voltage of the ICI and CCI are the same, and the last term is $> \sqrt{2}V_{pcc}$. Thus, the power range boundary of the ICI and CCI can be illustrated as a circle, in which the points are at the power base and the radii of ICI and CCI are $V_L V_{pcc}/Z_L$ and $V_C V_{pcc}/Z_C$, where V_L and V_C are the output voltage of the ICI and CCI, respectively. Figure 3A depicts the positions of the power range circles of the ICI and CCI, and the total power range of the hybrid connected CCI and ICI. The total power ranges of the traditional parallel-connected ICIs are shown in Figure 3B.

In Figure 3, P_{L-max} and P_{C-max} are the maximum active power transfer boundaries of ICI and CCI, respectively; Q_L and

Q_C indicate the reactive power compensation range of these two inverters, respectively; P_{H-max} and P_{H-max} are the maximum active power transfer boundaries of the proposed hybrid connected system, respectively; and Q_H is the power compensation range of a parallel-connected ICI system.

According to Figure 3, it can be found, that with the proposed system, the active power transfer range of the two MG systems is the same, and the reactive power compensation range of the proposed system is much larger than the traditional one.

The output voltage of the inverter will also affect the power transfer. Figure 4 shows the power range with the variation of the output voltage of ICI and CCI.

The CCI achieved a wide range of reactive power compensation with lower output voltage, and the DC-link voltage requirement of CCI could be effectively reduced, which increased the adaptiveness and flexibility of the MG system in reactive power compensation (Figure 4).

The equivalent value concept of the coupling impedance can be replaced by the power base. Practically, the power base of the CCI is usually set equal to the nominal load demand of the MG. In this paper, the CCI's power base is set as a reference like the following:

$$S_{base} = S_{base-C}$$

The ratio of the ICI power base to the reference power base was set as:

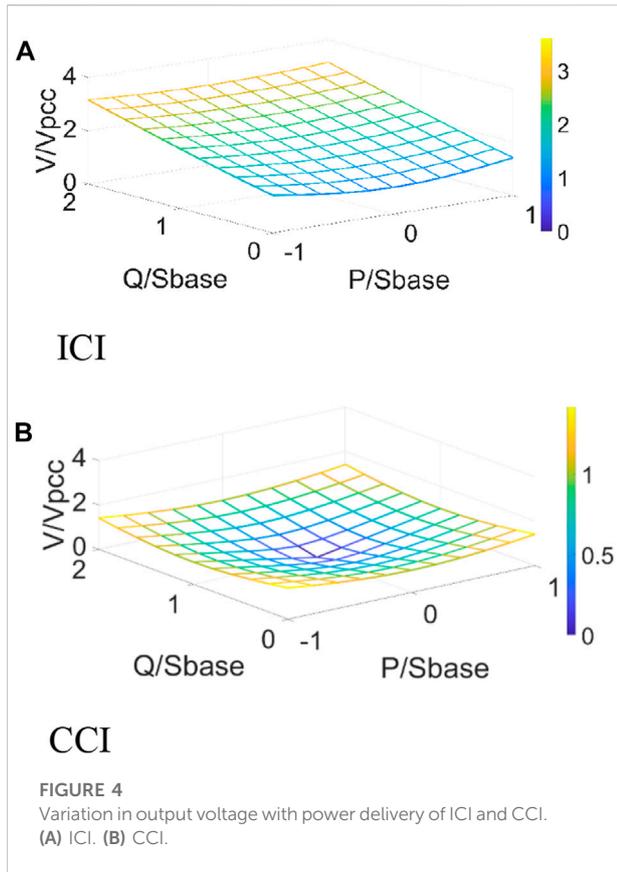


TABLE 2 Power boundary of the hybrid MG system.

Inverter	Boundary
ICI	$0 \leq P_L \leq \lambda V_L / V_{pcc} S_{base}$
	$\lambda (V_L / V_{pcc} - 1) S_{base} \leq Q_L \leq \lambda (V_L / V_{pcc} + 1) S_{base}$
CCI	$0 \leq P_C \leq V_C / V_{pcc} S_{base}$
	$(1 - V_C / V_{pcc}) S_{base} \leq Q_C \leq (V_C / V_{pcc} + 1) S_{base}$

$$\lambda = \frac{S_{base-L}}{S_{base}} \tag{6}$$

If the battery charging in a MG is not considered, the power range boundaries of the proposed MG system can be deduced (Table 2).

To sum up, we believe that the proposed hybrid of ICIs and CCIs connected to a MG system is able to achieve a wider reactive power compensation range. Since the CCI is able to work under a lower DC-link voltage, an optimal power sharing method could reduce the total power capacity compared with a traditional microgrid using ICIs only.

TABLE 3 Power reference coefficients in the case studies.

Case	Power reference coefficients
1	$\begin{cases} r_p = 1.8 \\ r_q = 0.35 \end{cases}$
2	$\begin{cases} r_p = 1.4 \\ r_q = 1 \end{cases}$
3	$\begin{cases} r_p = 0.7 \\ r_q = 0.35 \end{cases}$
4	$\begin{cases} r_p = 0.7 \\ r_q = 1.6 \end{cases}$
5	$\begin{cases} r_p = 0.3 \\ r_q = 1 \end{cases}$

3 Power-sharing for the proposed MG system

3.1 Power sharing ratio

The power-sharing ratio was defined for the proposed MG system. The power reference of the hybrid-connected ICI and CCI can be described as:

$$\begin{aligned} P_L &= m * P_{out} \\ Q_L &= n * Q_{out} \\ P_C &= (1 - m) * P_{out} \\ Q_C &= (1 - n) * Q_{out} \end{aligned} \tag{7}$$

where P_{out} and Q_{out} represent the total power injected into the PCC, P_L and Q_L are the power references of the ICI, and P_C and Q_C are the power references of the CCI.

To normalize the power demands, power reference coefficients are defined as:

$$\begin{bmatrix} P_{out} \\ Q_{out} \end{bmatrix} = \lambda S_{base} \begin{bmatrix} r_p \\ r_q \end{bmatrix} \tag{8}$$

The power capacity of ICI and CCI can be deduced as:

$$\begin{cases} S_L = V_L i_L \\ S_C = V_C i_C \end{cases} \tag{9}$$

where i_L and i_C are the output currents from ICI and CCI, respectively. They are expressed as:

$$\begin{cases} i_L = \frac{\sqrt{P_L^2 + Q_L^2}}{V_{pcc}} = \frac{\sqrt{(mP_{out})^2 + (nQ_{out})^2}}{V_{pcc}} \\ i_C = \frac{\sqrt{P_C^2 + Q_C^2}}{V_{pcc}} = \frac{\sqrt{[(1-m)P_{out}]^2 + [(1-n)Q_{out}]^2}}{V_{pcc}} \end{cases} \tag{10}$$

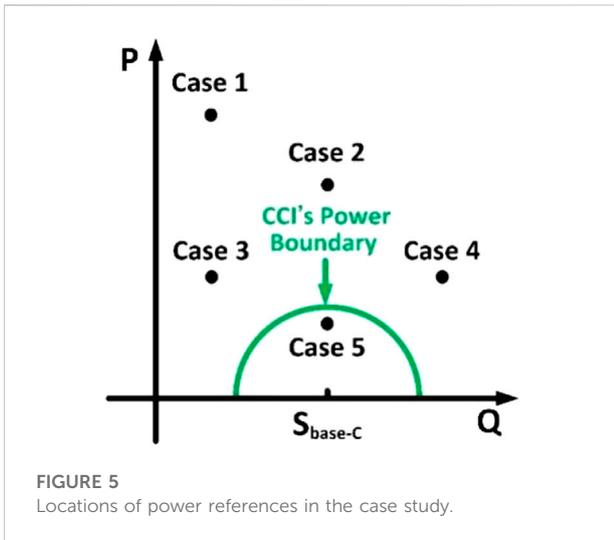


FIGURE 5 Locations of power references in the case study.

According to Eq. 9 and Eq. 10, the power capacity of an inverter changes with the power ratio for a given set of active and reactive power references. A detailed discussion will be provided hereinafter.

3.2 Effect of power sharing ratio on power capacity

To better evaluate the effect of power sharing ratio on power capacity, a case study is essential.

Assuming that the power base of the ICI was 1/8 times that of the CCI and the ratio of DC-link voltage to PCC voltage for the ICI was 1.5 and for the CCI, 0.7, six cases with random power reference coefficients are listed in Table 3.

The positions of the case studies in terms of the power delivery range of the CCI are shown in Figure 5. As shown in the figure, the active power reference in case 1 was much larger than the reactive power reference. In cases 2, 3, and 4, there was little difference between the active and reactive power references, which were uniformly distributed outside the CCI's power output range. The power references in case 5 both fell within CCI's power range. To keep the proposed system always working within the minimum power capacity, after receiving the power demand signals, the current from the CCI should be as large as possible because of the lower DC-link voltage, and to keep minimum power capacity the current from the CCI must obey the relationship:

$$\frac{i_c}{i_l} \leq \frac{V_l}{V_c}$$

By substituting the power reference coefficients into (Eq. 11), we noted that the current from the ICI varied according to the different power-sharing ratios (Figure 6). Thus, the power sharing ratios significantly affected the output current.

According to Figure 6, the lowest point is always located at (0, 0) in the x-y plane, which means the ICI should never output any current on mathematical calculation. However, due to the limitation of voltage and the power base, a single CCI may not always satisfy the power demands. Therefore, to find the

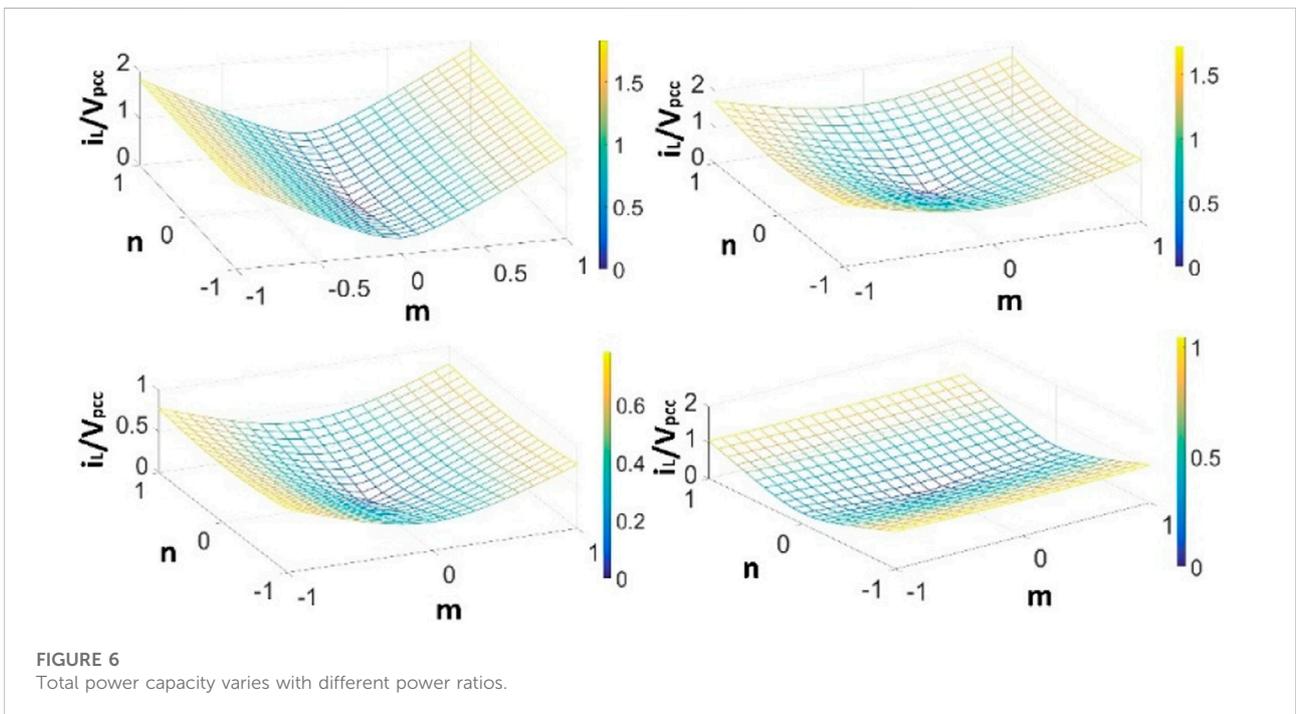


FIGURE 6 Total power capacity varies with different power ratios.

Since only the CCI is active in case 5, power sharing is not available. The current vector graphs for power-sharing are illustrated in Figure 7 for cases 1–4.

In the figure, i_{L-e} is the output current from the ICI under equal power sharing, and it can be obtained as:

$$i_{L-e} = \frac{\sqrt{r_p^2 + r_q^2} S_{base}}{2V_L} \quad (14)$$

For this case study, the number of ICIs under equal power sharing was assumed to be 2.

In Figure 7, i_{p-ref} , i_{q-ref} are the active and reactive power vectors in the reference current, respectively. They can be obtained as:

$$\begin{cases} i_{p-ref} = \frac{r_p}{V_{pcc}} S_{base} \\ i_{q-ref} = \frac{r_q}{V_{pcc}} S_{base} \end{cases} \quad (15)$$

where i_{qC} is the reactive power vector current when the CCI compensates for the reactive power equal with the power base as:

$$i_{qC} = \frac{S_{base}}{V_{pcc}} \quad (16)$$

The radius of the circle of the CCI's power range can be described as:

$$r_c = \left| \frac{V_C}{V_{pcc}^2} * S_{base} \right| \quad (17)$$

The initial point and the location of the power reference are marked as α_1 and α_3 , respectively. Their positions on the coordinate are:

$$\begin{cases} \vec{\alpha}_1 = 0 + j0 \\ \vec{\alpha}_3 = i_{q-ref} + j i_{p-ref} \end{cases} \quad (18)$$

and i_C and i_L can be calculated as:

$$\begin{cases} i_L = \vec{\alpha}_3 - \vec{\alpha}_2 \\ i_C = \vec{\alpha}_2 - \vec{\alpha}_1 \end{cases} \quad (19)$$

Assuming the acute angle between r_c and the horizon axis is β , the point α_2 gives the maximum current from the CCI with a given power reference. It has three positions:

1. When $r_q > 1$, the current is:

$$\vec{\alpha}_2 = \left(r_c \sqrt{\frac{1}{1 + \tan^2 \beta}} + i_{qC} \right) + j \left(r_c \sqrt{1 - \frac{1}{1 + \tan^2 \beta}} \right) \quad (20)$$

The power-sharing ratio is:

$$\begin{cases} m = \frac{\left(i_{p-ref} - r_c \sqrt{1 - \frac{1}{1 + \tan^2 \beta}} \right) V_L}{r_p S_{base}} \\ n = \frac{\left[i_{q-ref} - \left(r_c \sqrt{\frac{1}{1 + \tan^2 \beta}} + i_{qC} \right) \right] V_L}{r_q S_{base}} \end{cases} \quad (21)$$

and:

$$\tan \beta = \frac{i_{p-ref}}{i_{q-ref} - i_{qC}} \quad (22)$$

2. When $r_q < 1$, the current is:

$$\vec{\alpha}_2 = \left(i_{qC} - r_c \sqrt{\frac{1}{1 + \tan^2 \beta}} \right) + j \left(r_c \sqrt{1 - \frac{1}{1 + \tan^2 \beta}} \right) \quad (23)$$

The power-sharing ratio is:

$$\begin{cases} m = \frac{\left(i_{p-ref} - r_c \sqrt{1 - \frac{1}{1 + \tan^2 \beta}} \right) V_L}{r_p S_{base}} \\ n = \frac{\left[i_{q-ref} - \left(i_{qC} - r_c \sqrt{\frac{1}{1 + \tan^2 \beta}} \right) \right] V_L}{r_q S_{base}} \end{cases} \quad (24)$$

and:

$$\tan \beta = \frac{i_{qC} - i_{q-ref}}{i_{p-ref}} \quad (25)$$

3, When $r_q = 1$, $\beta = 90^\circ$, and r_p is larger than the maximum active power transfer of CCI, the current is:

$$i_{qC} = i_{q-ref} \quad (26)$$

the expression for α_2 is:

$$\vec{\alpha}_2 = i_{qC} + j i_{p-ref} \quad (27)$$

and the power-sharing ratio is:

$$\begin{cases} m = \frac{\left(i_{p-ref} - 8 \frac{V_L}{V_{pcc}^2} S_{base} \right) V_L}{r_p S_{base}} \\ n = 0 \end{cases} \quad (28)$$

• If the power references satisfy:

$$\sqrt{(1 - r_q)^2 + r_p^2} \leq \frac{V_C}{V_{pcc}} \quad (29)$$

the power-sharing ratio is:

$$\begin{cases} m = 0 \\ n = 0 \end{cases} \quad (30)$$

TABLE 4 Summary and comparison of power capacities of different systems.

Case	Total power capacity (S/S_{base-c})				
	ICIs equal sharing	ICI and CCI			
		Adaptive sharing	m	n	
1	2.7546	1.9812	1	-0.16	
2	3.4181	1.5463	0.69	0	
3	2.9984	0.7913	0.87	-0.18	
4	3.1026	1.0675	0.51	-0.94	
5	1.3254	1.0189	0	0	

4.3 Effectiveness of the adaptive power sharing method

According to Eqs 14–30, the power sharing ratios can be obtained for cases 1–4. The power sharing ratio in case 5 is determined as (0, 0). Thus, the corresponding power ratios under the proposed method and the conventional method were calculated and are summarized in Table 4. As the comparison shows, the proposed power sharing method for the hybrid MG system can effectively reduce the power capacity.

5 Realization of the NN-based power sharing controller

In the last section, a complete power sharing calculation process is provided. It can be seen that the calculation process is complicated and includes some complex forms of computation, such as trigonometric functions and square roots. It could be difficult for the controller chip to compute when the load demands vary rapidly. To solve this problem, an NN-based power sharing model is proposed in this work.

5.1 Training and evaluation of the NN model

The architecture of an NN model is based on neurons and can be divided into three main parts: the input layer, the hidden layer, and the output layer (Wang et al., 2020). The input layer receives signals from the outside world. The hidden layer is the intermediate layer connecting the input layer and the output layer, and constitutes the core of the NN as being composed of many neurons. The output layer returns the results of the mathematical operations on the input (Liu et al., 2021).

As a representative of a NN the application areas of a back-propagation neural network (BPNN) include function

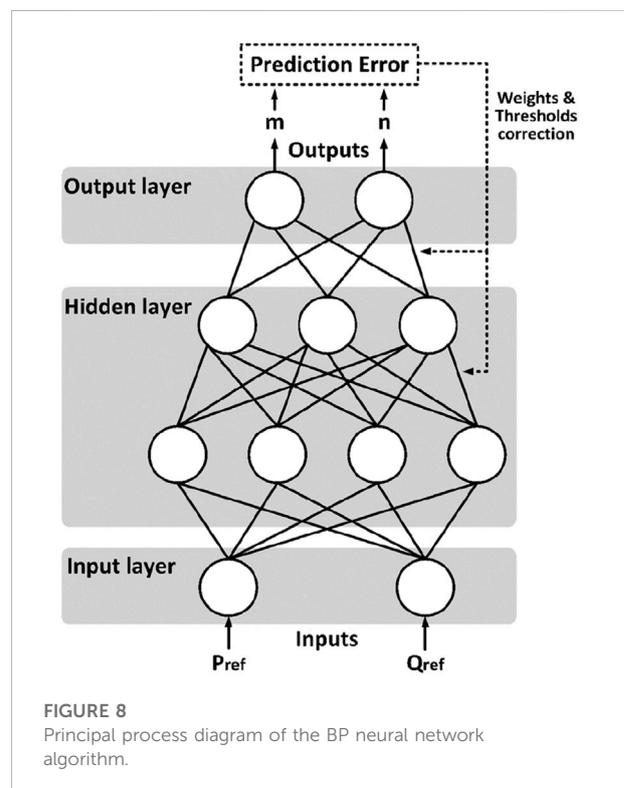


FIGURE 8 Principal process diagram of the BP neural network algorithm.

approximation, regression analysis, numerical prediction, and classification and data processing (Chen et al., 2020). The BPNN calculates the final network error based on the forward operation of the input data, transmits the error in the opposite direction, and contains a mechanism for adjusting the weights and thresholds of the corresponding layers according to certain rules when crossing different layers. After a large number of data samples are trained, an algorithmic model that can accomplish complex nonlinear mapping is finally constructed. For such networks, the input layer is responsible for feeding data into the neural network. The hidden layer fits and optimizes the parameters to the

TABLE 5 Partial equation calculation results.

Case	MSE
1	
2	
3	
4	
5	

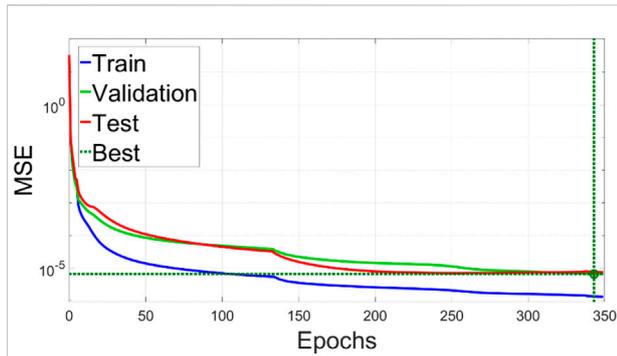


FIGURE 9 Training results of the proposed BPNN.

neural network by model training. These include weights and biases as well as the selection of various parameters and activation functions. The output layer is responsible for outputting the computational results from the neural network (Wang et al., 2022). The principal algorithm process of the BPNN model in this paper is illustrated in Figure 8.

The BPNN model can be expressed mathematically as $f(\sum_i X_i W_i + b_i)$, where f is the activation function, which introduces non-linear factors to a linear model, thus solving problems that are difficult to solve with a linear model. X_i represents the output data, W_i is the weight between the connected input data and the hidden layer, while b_i is the bias vector. During the training process, the accuracy of the NN

model is improved by continuously changing the values of the parameters W_i and b_i .

A loss function was set to examine the difference between the predicted value and the known answer, and also to evaluate the accuracy of the trained NN model. With enough training sessions, the loss function will be stable in the vicinity of a very small value, which can be regarded as the prediction error of the NN model (Huang et al., 2021). In this paper, the sigmoid function was chosen as the activation function and the mean square error (MSE) as the loss function to describe the prediction error.

5.2 Effect of the prediction error

For a BPNN, the number of neurons, hidden layers, as well as training sessions, have a high positive correlation with the final MSE. However, an excessive number of neurons and hidden layers will increase the computational power required of the chip in the power sharing controller. In addition, an increase in the amount of training data will also increase the difficulty of BPNN training (Huang et al., 2021). Therefore, it is essential to choose the proper numbers of neurons and hidden layers, and also the amount of training data. The selection principle is based on experience, as well as the required prediction error of the BPNN.

To evaluate the impact on the total power capacity due to the prediction error, a partial derivative equation is presented. Assuming that the prediction error of the total power capacity is defined as:

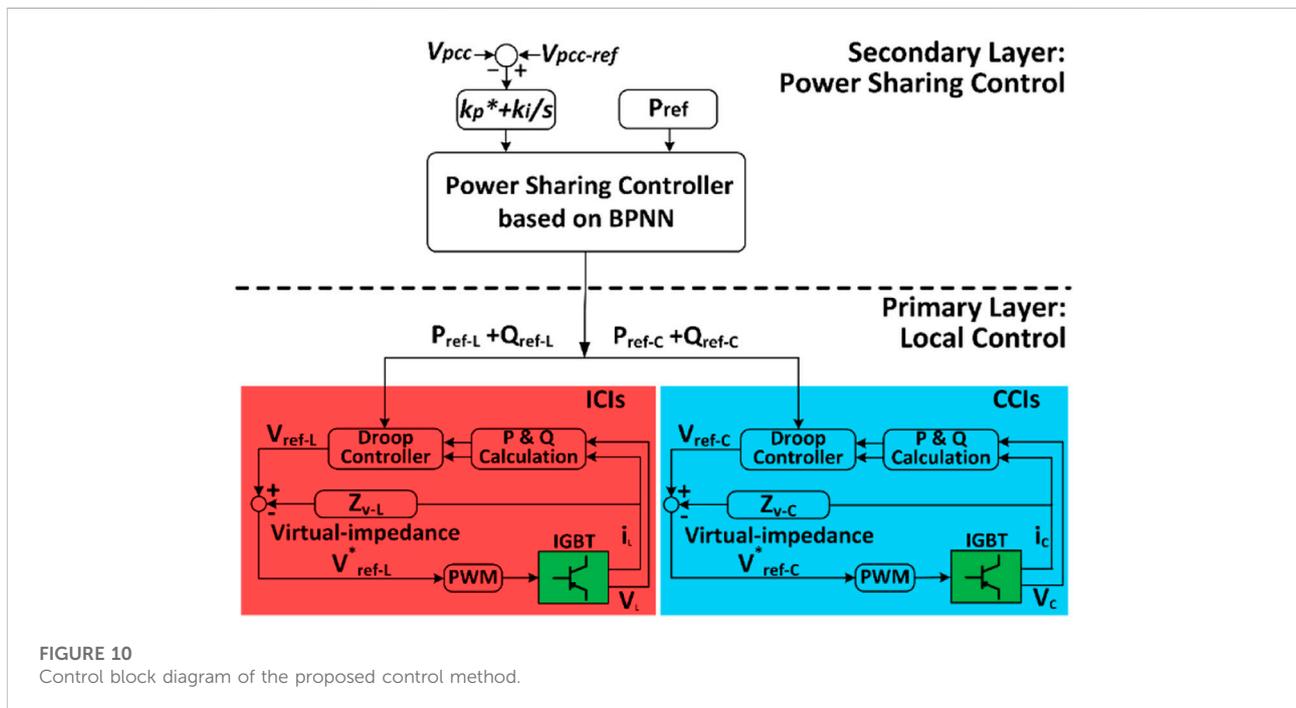


FIGURE 10 Control block diagram of the proposed control method.

TABLE 6 Summary and comparison of power capacities of different systems.

System	
V_{PCC}	110 V/50 Hz
ICI	
V_{DC-L}	170 V
L_i	4 mH
S_{base-L}	9680 Var
Active power transfer	$1.1S_{base-L}$
Reactive power compensation	$0.1S_{base-L}$
CCI	
V_{DC-C}	80 V
L_c	2 mH
C_c	340 μF
S_{base-C}	1210 Var
Active power transfer	$0.5S_{base-C}$
Reactive power compensation range	$(0.5 \sim 1.5)S_{base-C}$

$$\Delta s = \frac{\partial S}{\partial m} \Delta m + \frac{\partial S}{\partial n} \Delta n \quad (31)$$

there are:

$$\frac{\partial S}{\partial m} = \frac{\partial S_L}{\partial m} + \frac{\partial S_C}{\partial m} \quad (32)$$

$$\begin{aligned} \frac{\partial S_L}{\partial m} = & \frac{1}{2} \frac{2mr_p^2}{\sqrt{(1+nr_q)^2 + (mr_p)^2}} \sqrt{(nr_q)^2 + (mr_p)^2} S_{base-L} \\ & + \sqrt{(1+nr_q)^2 + (mr_p)^2} \frac{2mr_p}{\sqrt{(nr_q)^2 + (mr_p)^2}} \frac{1}{2} S_{base-L} \end{aligned} \quad (33)$$

$$\begin{aligned} \frac{\partial S_C}{\partial m} = & (m-1)r_p^2 S_{base-C} \\ & \times \left[\frac{\sqrt{[(1-m)r_p]^2 + [(1-n)r_q]^2}}{\sqrt{[1-(1-n)r_q]^2 + [(1-m)r_p]^2}} + \frac{\sqrt{[1-(1-n)r_q]^2 + [(1-m)r_p]^2}}{\sqrt{[(1-m)r_p]^2 + [(1-n)r_q]^2}} \right] \end{aligned} \quad (34)$$

and:

$$\frac{\partial S}{\partial n} = \frac{\partial S_L}{\partial n} + \frac{\partial S_C}{\partial n} \quad (35)$$

$$\begin{aligned} \frac{\partial S_L}{\partial n} = & (1+nr_q)r_q S_{base-C} \\ & \times \left[\frac{\sqrt{(mr_p)^2 + (nr_q)^2}}{\sqrt{(1+nr_q)^2 + (mr_p)^2}} + \frac{\sqrt{(1+nr_q)^2 + (mr_p)^2}}{\sqrt{(mr_p)^2 + (nr_q)^2}} \right] \end{aligned} \quad (36)$$

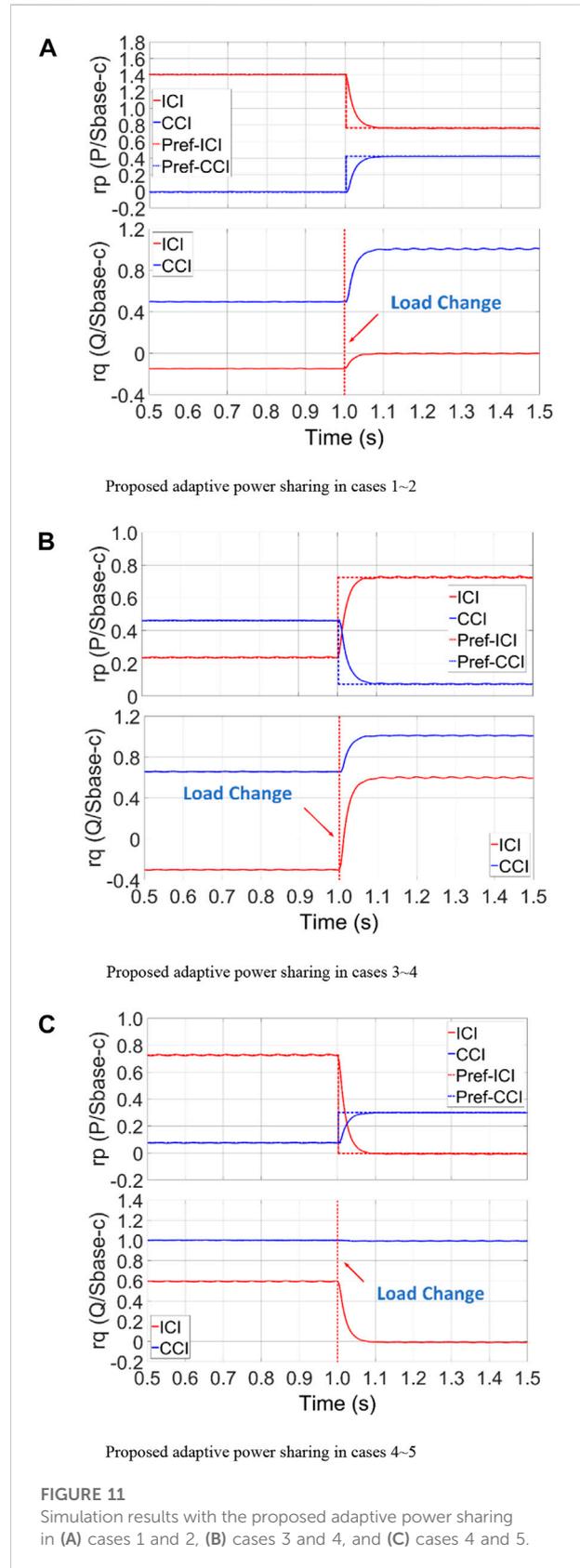
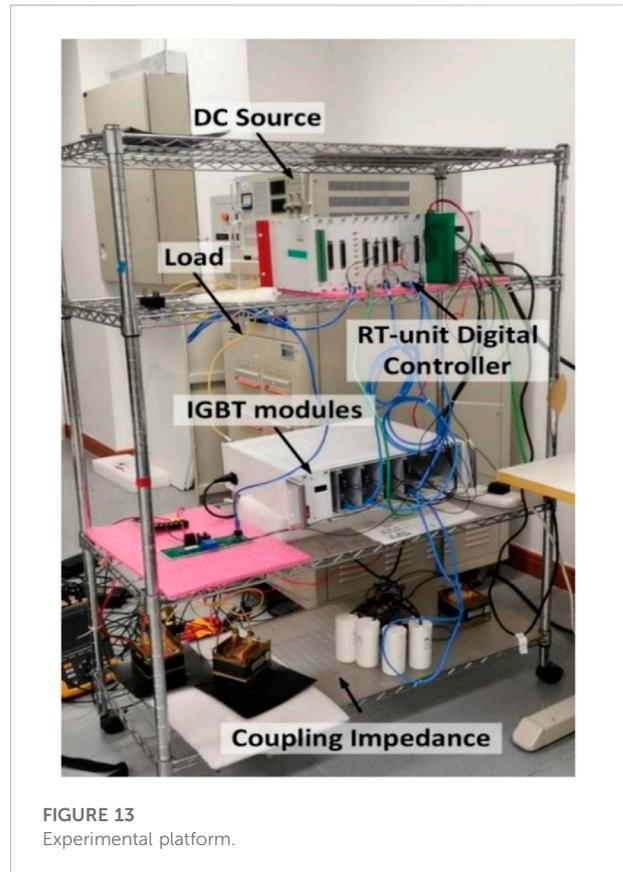
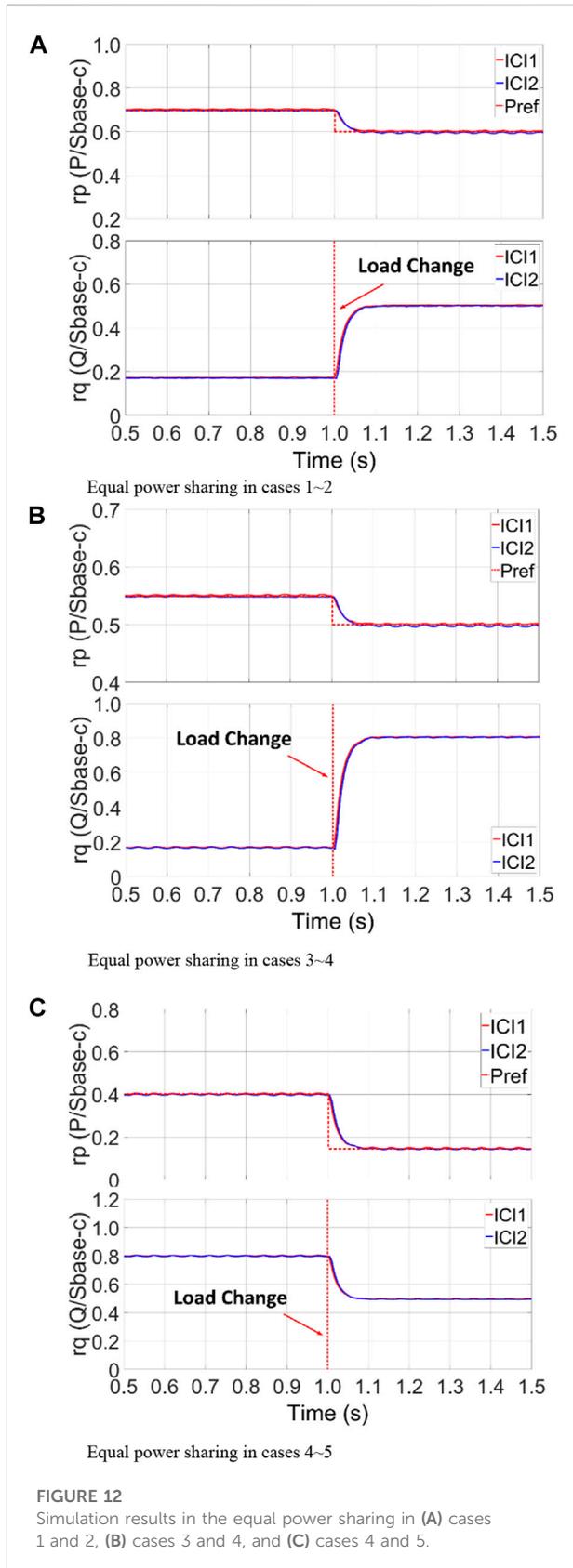


FIGURE 11 Simulation results with the proposed adaptive power sharing in (A) cases 1 and 2, (B) cases 3 and 4, and (C) cases 4 and 5.



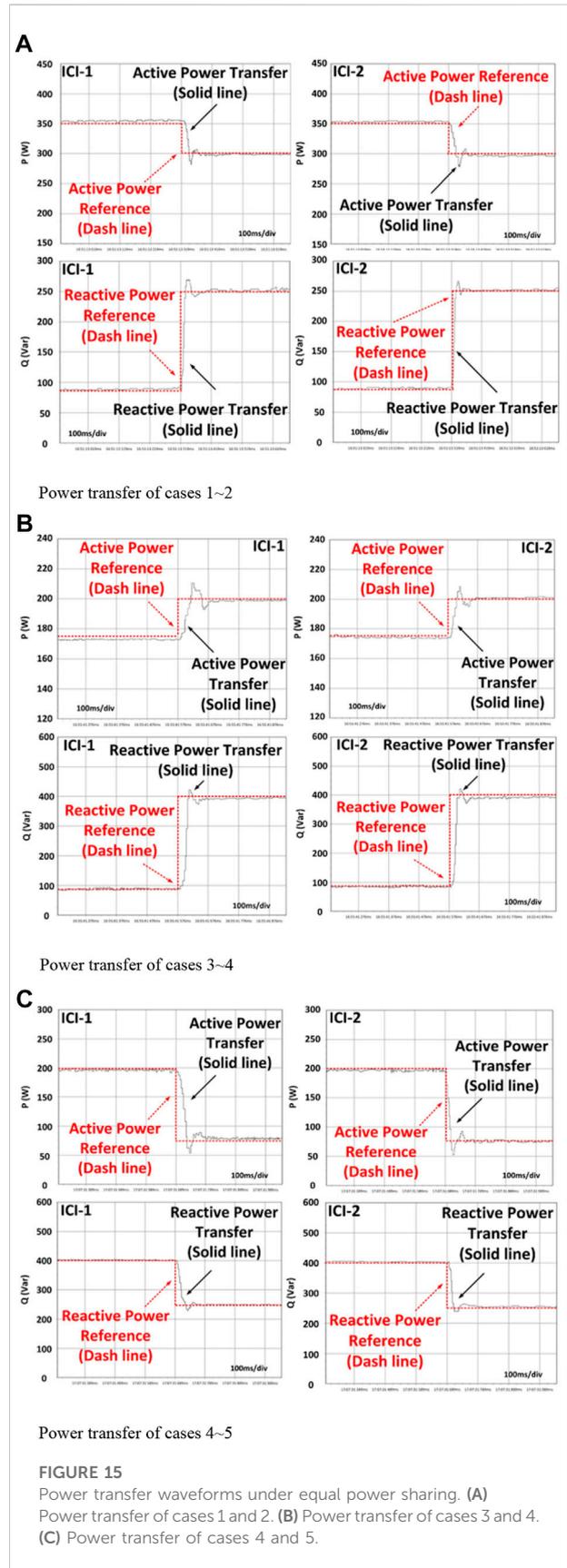
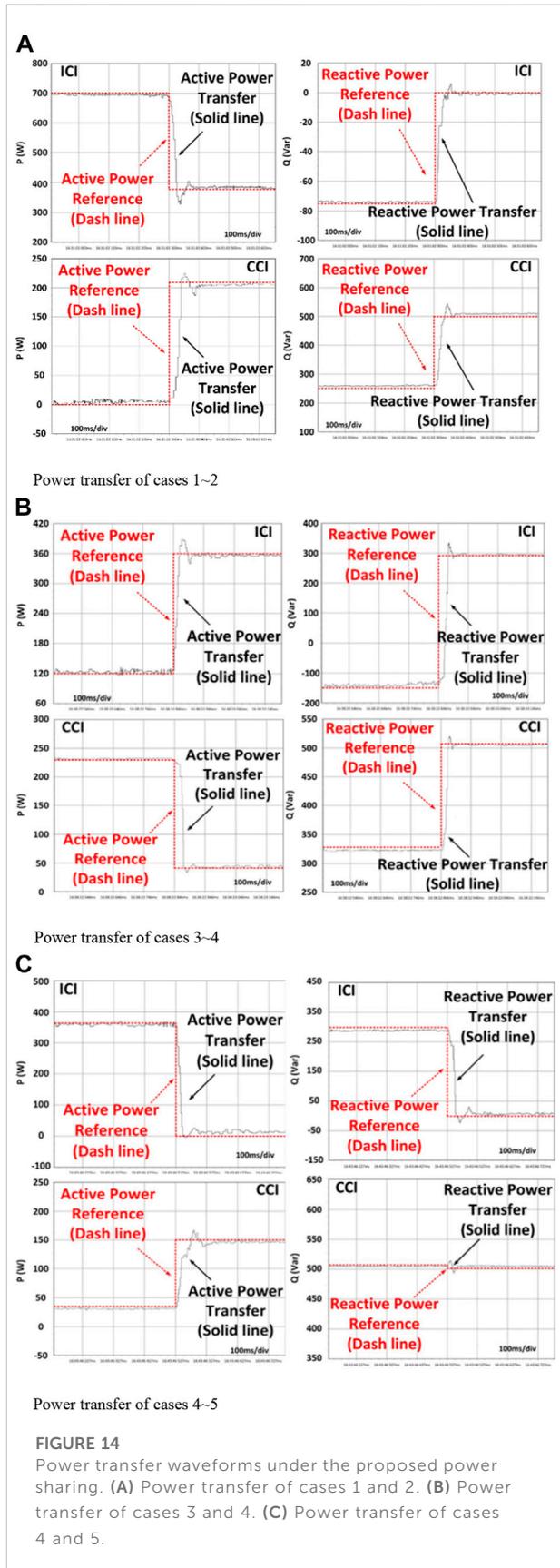
$$\frac{\partial S_C}{\partial m} = [1 - (1-n)r_q]r_q S_{base-c} \times \left[\frac{\sqrt{[(1-m)r_p]^2 + [(1-n)r_q]^2}}{\sqrt{[1 - (1-n)r_q]^2 + [(1-m)r_p]^2}} + \frac{\sqrt{[1 - (1-n)r_q]^2 + [(1-m)r_p]^2}}{\sqrt{[(1-m)r_p]^2 + [(1-n)r_q]^2}} \right] \quad (37)$$

The results from inserting the parameters from cases 1 to 5 into (Eq. 26) are summarized in Table 5.

As in Table 4, the degrees of impact of m and n on the total power capacity are not the same. As the reactive power sharing ratio will more critically affect the prediction error, a larger active power demand will increase the effect of the reactive power sharing ratio error on the power capacity. A smaller reactive power output will reduce the influence of the active power sharing ratio on the power capacity.

Assuming that the total power capacity bias should not exceed 5%, the prediction errors of the active and reactive power sharing ratio must be smaller than 0.5%, which means that the MSE of the BPNN cannot be larger than 0.005 to keep the total power bias within a satisfactory range.

The BPNN model includes two hidden layers, and a total of 15 neurons were proposed. The size of the training set was 380 groups of power references and sharing ratios. Like the



training results shown in Figure 9, after training, the MSE of the BPNN model reached a value on the order of 10^{-5} , which is satisfactory for generating power sharing ratios.

The control block diagram of the proposed power sharing method for the hybrid ICI and CCI system microgrid is presented in Figure 10. As the figure shows, the power demands will be sent to the secondary power sharing layer with a well-trained BPNN model and generate an optimal power sharing ratio, which can reduce the total power capacity requirement of the hybrid MG system to the minimum value. The power sharing ratio then will be used to calculate the actual power references for the local inverters. Virtual-impedance loops are employed in the local control layers to eliminate the tracking errors by droop control.

6 Simulations and experimental verification

The proposed power sharing method aims to keep the hybrid MG system working with the lowest power capacity and to ensure accurate power transfer. Since the power capacity cannot be obtained by the results of the test, therefore, in this section, all tests will follow the case study strictly, and the results will verify the accuracy of power transfer under the proposed power sharing method.

6.1 Simulation results

Simulation tests were conducted in Matlab/Simulink, to verify the effectiveness of the proposed power sharing strategy. Test cases are listed in Table 3, and the control block diagram is given in Figure 10. Simulation parameters are listed in Table 6. All tests were performed with ICIs under equal power sharing and using the proposed adaptive power sharing model. The simulation results with the proposed MG system and adaptive power sharing method are illustrated in Figure 11 and the results of equal power sharing with the ICIs in the MG system are shown in Figure 12. Power references for each inverter were calculated based on the total power references and power sharing ratio; the references were varied by 1.5 s in each case. Based on the results, the power sharing algorithm proved effective and the output power tracked the reference smoothly and accurately.

6.2 Experiment verification

To verify the power control effectiveness of the proposed method on the hardware level, a prototype was implemented

in the lab (Figure 13). The configuration of the testing system was the same as the simulation. The parallel-connected inverters were controlled by a DSP-based controller. All tests followed the simulation settings. The power transfer waveforms under the different cases are shown in Figure 14.

Similarly, the conventional equal power sharing method was also tested on parallel-connected ICIs. The power transfer waveforms are provided in Figure 15.

7 Conclusion

The CCIs and ICIs have advantages in active and reactive power transfer, respectively. To enhance the reactive power compensation capacity, lower the total power capacity, and reduce transmission losses in a MG system, a hybrid of ICIs and CCIs grid-tied to the MG system was proposed. With its different power characteristics, the conventional equal-sharing method was not suitable for the proposed system. In order to keep the MG system operating continually at minimum power capacity, a power-sharing ratio calculation method for adaptive power sharing was proposed. Simulations and experimental results proved that the hybrid MG system can work well with the proposed control strategy.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary Material; further inquiries can be directed to the corresponding author.

Author contributions

WD and YT provided the ideas and theory of the paper, QL and YY helped to implement the experimental platform, and YZ supervised the research work.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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