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Blockchain-based bilateral bidding market mechanism with carbon allocation on both supply and demand sides

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Recently, the interaction between power generation and load has been continuously strengthened, providing a new method for realizing the decarbonization of the power system. An auction method that enables a low-carbon economic dispatch is designed in this study, where blockchain technology is employed to record the whole process data credibly. According to the regional division of market entities, carbon emissions are allocated to regions and entities by using the Owen value method. In order to reduce the expected carbon emission of entities, an additional carbon price is added to the original quotation through the allocation results. Case studies are conducted based on the typical output characteristics of wind and photovoltaic power generation in the bilateral bidding market hourly. The results have validated that the proposed method can reduce expected carbon emissions in the power system by prioritizing zero-carbon and low-carbon generation units.

KEYWORDS

carbon allocation, cooperative game, low carbon electricity, Owen value, blockchain

1 Introduction

Global warming is becoming more obvious with the intensification of greenhouse gas emissions (Zhongming et al., 2021). Hence, how to reduce carbon is an important topic. Among the carbon emissions of energy, the power system accounts for about 40% so that electricity will play an increasingly important role in the future. In building a power system with a higher proportion of renewable energy, how to realize renewable energy consumption is also crucial. In the first 5 months of 2022, China's entire society consumed 3,352.6 billion kWh of electricity. As of the end of May 2022, coal-fired thermal power accounted for 49.3% of the installed capacity of power plants. At the same time, gas-fired power generation accounted for 4.8%, and grid-connected wind power and photovoltaic power generation accounted for only 15.1% and 9.3%, respectively. Currently, China has established a carbon market to meet the checks and balances on the proportion of renewable energy and thermal power units through carbon trading(Yue et al., 2021). The

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emission reduction of the power industry is a vital and challenging task, and people are constantly exploring solutions (Wang and Feng, 2017; Zhang D et al., 2017; Zhang X et al., 2017; Zhang Y.-J et al., 2017; Cheng Z et al., 2018). Therefore, in this study, it is significant to allocate the carbon cost to each entity and guide each entity to reduce emissions and actively participate in renewable energy consumption by means of a market mechanism.

Since the power industry and carbon emissions coexist, figuring out how to achieve decoupling between power and carbon emissions is a critical step in evaluating carbon properties. Some of the methods currently start from the input–output model to track the emission flow through the power grid and quantify the pollution, which is reflected in the power production, exchange, and final consumption of the corresponding institutions (De Chalendar et al., 2019). Some studies use input-output analysis (EIOA) to calculate the implied emission intensity or use a top-down approach to assess the environmental footprint (Gao et al., 2022), while other studies convert electricity into emissions by using the EIA factor to convert fuel into emissions where energy carbon emissions are imputed (Goldstein et al., 2020). Furthermore, some studies decouple the two models by introducing the concept of carbon emission flow attached to the power flow (Kang et al., 2015; Tian et al., 2015). In addition, relevant operating scenarios also include quantifying carbon emissions related to energy transmission (Cheng Y et al., 2018) and conversion processes in multi-energy systems, coordinating energy generation, transmission, conversion, and utilization across different energy sectors, paths, and time scales, and implementing a hierarchical process (Huang et al., 2020). Some studies start from the perspective of cooperative games. At first, the



Shapley value method can be used for the benefit distribution of internal members in the electricity market (Lei et al., 2020). Later studies regard the whole process of carbon allocation as a cost allocation process and use the Shapley value to share carbon emissions on the load and power generation sides (Chen et al., 2017). However, since the entire system is time-varying and carbon emissions are spatially and temporally different, a reasonable distribution cannot be achieved by simply describing them with fixed parameters.

At present, the existing research hardly allocates the carbon emissions to the power generations and loads. It changes the current trading cycle of quotations by considering the allocation to promote the realization of carbon reduction on both sides of power generation and loads. At the same time, in the current ideas of cooperative games, all market entities are regarded as the same class for cost allocation. However, due to the different geographical distribution of each entity as well as issues such as administrative divisions and policies, it does not take into account the impact of the division of the major leagues. Thus, they belong to cooperative games with coalition structure constraints, which is consistent with the characteristics of Owen's value method. The research (Fiestras-Janeiro et al., 2015) proves the hypothesis of the Owen value. Moreover, the design of a bilateral bidding mechanism considering carbon allocations on both sides starts from the problems mentioned previously by adopting the two-level structure of region-entity. According to the Owen value method and the rules of bilateral bidding, the carbon allocation is set for the entity in each game.

The carbon allocation result is transmitted as a price signal to original quotations of entities, and the final transaction is cleared according to the quotation after considering the change of carbon allocations.

Due to the process of quotation changing in the middle, in order to guarantee the traceability of the data, blockchain technology is used to realize the whole process recording on the chain. At present, some research studies have designed the transaction and settlement framework in the electricity market by blockchain (Xu et al., 2021), as well as simplified the transaction process and realized mutual trust between users (Zhou et al., 2021). At the same time, a decentralized trading market is designed in the distribution market for bilateral energy trading (Morstyn et al., 2018). Due to the blockchain consensus mechanism, various information silos are connected (Hamouda et al., 2020), and the problem of information tampering in distributed energy transactions is prevented (Tonev and Nikolaev, 2020). Some studies use private key encryption to protect the privacy of the auctioneer in the bidding process (Zhang et al., 2019). At the same time, there is also a two-level structure of the agent and transaction settlement in the transaction (Luo et al., 2018). It also establishes a regional structure for many types of entities and realizes inter-regional and cross-regional transactions (Li et al., 2021). In order to fit with the aforementioned region-entity twolevel structure, a two-level blockchain structure with the main chain of the whole system and sub-chains of each region is designed. Blockchain technology is used in the overall process,



and cross-chain between multiple blockchains is used. A blockchain is set up in the overall system to realize the traceability record of regional data and critical data in the transaction, and the aforementioned method is written into the smart contract to transmit the optimal unit output across the chain to each regional blockchain for scheduling. The innovations are as follows:

- The Owen value method is used to allocate carbon emissions in two levels. The Shapley method is used between regions, and the improved Shapley method is used for each entity in the region to achieve a reasonable allocation.
- 2) The process of carbon allocation is integrated into the quotations of the current trading bilateral market. In that case, the factor of carbon allocation can be fully considered while satisfying social welfare in the final transaction.
- 3) Blockchain technology is used to record the corresponding data of the whole process to the chain and realizes the operability and achievability of the process by adopting a two-level blockchain structure.

The remaining contents of this article are as follows: Section 2 introduces the operation process of the whole market

mechanism. Section 3 introduces the main models and methods. Section 4 analyzes the results through simulation. Section 5 gives the conclusion and summarizes the whole article.

2 Bilateral electricity market framework with carbon emission allocation

The article follows the general idea of using carbon emissions as a price signal to change the auction before the completion of the current transaction. A cooperative game method is used to allocate carbon emissions to the power generation and load entities participating in the market. A new quotation is formed, taking into account the allocation results and the original quotation.

As shown in Figure 1, in the traditional bilateral bidding model in the electricity market, each entity quotes according to its economic cost while ignoring the environmental value, which is the impact of greenhouse gas emissions. In bilateral electricity trading considering carbon attributes, reasonable carbon allocation should be carried out according to the marginal effect of carbon emissions caused by the participation of each



entity in the market. Market organizers make revisions based on the original quotation and the allocation results before the transaction. It considers not only the physical value of electricity but also its environmental value, which can promote the development of low-carbon generation and make the entire system develop toward the goal of a low-carbon and economical operation.

The whole framework is divided into four parts, namely, transaction preparation, modification of quotations, final market transaction, and transaction data recording. The whole framework is shown in Figure 2.

In the transaction preparation process, each power generation entity reports data, including the upper and lower limits of output $\hat{P}_{Gi}P_{Gi}$, corresponding quotation λ_{Gi} , the carbon emission intensity of generation *its_i*, and regional locations *local_i*. Each load entity reports data simultaneously, including the upper and lower limits of the load demand $\hat{P}_{Li}P_{Li}$, the corresponding quotation λ_{Li} , and the regional location *local_i*.

In the modification of the quotation process, the Owen value method is used to allocate carbon emissions to each market entity by integrating the previously reported data. The allocation is to divide regional carbon emissions on the basis that each market entity is divided into regions. Then, the entities are divided according to the regional division results, and a ladder carbon price is set according to the allocation results, so as to make a new quotation. At the same time, according to the characteristics of the bilateral bidding market mechanism, in order to promote the low-carbon economic dispatch of the whole system, the additional quotation of the load entities should be negatively correlated with their carbon allocation. The additional quotation of the power generation entities should have a positive correlation with their carbon allocation.

In the final market transaction process, the latest quotation can be obtained based on the quotation of the market entity in the first step and the revised quotation in the second step. Based on the latest quotation, clearing is carried out according to the bilateral bidding model, and the output of each generation and the consumption of each load are distributed.

In the transaction data recording process, its aim is to record the data of the previous three steps, respectively, and realize the



automatic implementation of the process through smart contracts. This process establishes a two-level blockchain system. First, the sub-chains in each region are used for the record of report data of market entities and the settlement after the final transaction. Then, the main chain for all regions is used to integrate the report data of each sub-chain to form and record a new quotation.

$\max_{P_G P_D} \left(\sum_{i=1}^{D} P_{Di} \bullet \lambda_{Di} - \sum_{j=1}^{G} P_{Gj} \bullet \lambda_{Gj} \right),$ $\sum_{i=1}^{D} P_{Di} = \sum_{j=1}^{G} P_{Gj},$ s.t. $\breve{P}_{Gj} \leq P_{Gj} \leq \widetilde{P}_{Gj},$ $\breve{P}_{Di} \leq P_{Di} \leq \widetilde{P}_{Di},$ $f_1 \leq \widetilde{f}_1,$ (1)

3 Model and methodology

In this section, detailed modeling regarding the whole mechanism is introduced.

3.1 Bilateral bidding market model

The electricity market clearing model based on bilateral bidding is as follows:

where
$$P_{Di}$$
 represents the actual consumption of load *i*, the actual output of generation *j*, and, respectively, represent the quotations for load *i* and generation *j*, respectively, the minimum and maximum output of generation *j* and minimum and maximum consumption of load *i*.

Since in the bilateral bidding model, the buyer's quotations are sorted from high to low, the buyer can get more advantage in the auction if his bid is higher. The seller's quotation is sorted from low to high, so the lower the seller's quotation, the more the transaction will be prioritized, as shown in Figure 3.



3.2 Carbon allocation model based on the Owen value method

At present, the carbon emission allocation method is adopted from the perspective of a cooperative game. Under the premise of considering axiomatic standards, the Shapley method has good fairness and consistency. At the same time, considering that each entity is divided according to administrative regions, it follows the priority alliance's structure. Therefore, when allocating carbon emissions, it is necessary to separate them from the regional level first and then carry out the allocation of market entities according to the regional allocation results. In summary, this section adopts a two-level structure of the Owen value, and each level is allocated by the Shapley method. Each region can be treated as a priority alliance, and each market entity can be treated as a player in the game.

The carbon allocation of the entity is the additionally expected carbon emissions after the entity joins the bilateral market trading. However, the transaction is based on the modified quotations, which consider every allocation of the entity. The result of carbon allocation can be seen as a factor in the final transaction. Assume that all market players form a set $N = \{L_1, L_2, ..., L_{nl}, G_1, G_2, ..., G_{ng}\}$. Different market players are located in different positions, thus forming different alliances. Therefore, the set of all market players can be divided into *m* parts according to the alliance structure S,

$$S = \{\zeta_1, \zeta_2, ..., \zeta_m\}, \zeta_k = \{L_{k1}, L_{k2}, ..., L_{kl}, G_{k1}, G_{k2}, ..., G_{kg}\},$$
(2)

where ζ_k represents the area k, which consists of loads and generations in the area k, which is the group $\{L_{k1}, L_{k2}, ..., L_{kl}, G_{k1}, G_{k2}, ..., G_{kg}\}$ Therefore, the calculation of the Owen value method is divided into two steps as follows:

1) First, use the Shapley value method to allocate carbon among regions, which is a priority alliance. The method is as follows:

$$x_{\zeta_k} = \sum_{L \subset \left(S \setminus \{\zeta_k\}\right)} P(L) \left(c \left(L \cup \{\zeta_k\}\right) - c(L)\right), \tag{3}$$

where x_{ζ_k} represents the carbon allocation in the local k, L represents a subset of the set S with ζ_k removed, c(A) represents a function which calculates the carbon emissions when the market consists of entities in the alliance structure A, and $c(L \cup \{\zeta_k\}) - c(L)$ represents the



additional carbon emissions caused when joining in the market.

When considering that all regions are randomly arranged, the probability of the region L is as follows:

$$P(L) = \frac{n_L!(m - n_L - 1)!}{m!},$$
(4)

where *m* represents the number of alliance structures and n_L represents the number of the probability of forming subset *L*.

2) According to the idea of the first step, the market players in a certain regional structure are first considered. The Shapley value method is used to calculate the marginal effect of additional allocation to the area when an entity joins the area. The method is as follows:

$$x_{ci} = \sum_{R \subset \left(\zeta_k / \{ci\}\right)} P(R) \cdot (s(R \cup \{ci\}) - s(R)),$$
(5)

where x_{ci} represents the entities *ci* in the area *k* alliance structure, which is the load entity or generation entity; *R* represents a subset of the set ζ_k with *ci* removed; and s(R) represents carbon allocation in the area *k* when the market entities in area *k* only have all the elements in *R*. When considering that all entities are randomly arranged, the probability of the subset *R* is as follows:

$$P(R) = \frac{n_R!(n_k - n_R - 1)!}{n_k!},$$
(6)

where n_k represents the number of entities in the area k and n_R represents the number of the entities in the subset *R*.

Specifically, the combinations of entities need to be considered in order to calculate all the marginal effects. When





Output characteristic curve of the renewable energy unit.

calculating the carbon allocation of the entity C_i , it also needs to be carried out in two steps by the improved Shapley method as follows:

- 1) Since the carbon allocation of the target entity is calculated by the marginal effect of the region when the entity participates in each combination, we need to calculate the carbon allocation of the region ζ_k in each combination of the entities. When some combination of set *L* after removing ζ_k is chosen, we can get the total carbon emissions of the system if a bilateral transaction is carried out in this combination. When we add the region ζ_k to the set *L*, we can get the total carbon emissions of the system again. In this way, the marginal effect of one combination case is obtained. Finally, we can get all the case results and add them to get the region of the target entity allocation, shown in Figure 4A.
- 2) When some combination R is chosen, the distribution of the entities in the region will change. Thus, Eq. 3 is used to calculate the allocation of the region, which is where the



FIGURE 10

Consumption or generation of entity and carbon allocation result comparison. (A) Space-time distribution of power consumption. (B) Space-time distribution of load carbon allocation. (C) Space-time distribution of power generation. (D) Space-time distribution of generation carbon allocation.

target entity is located. Then, since the previous combinations did not contain the target entity, the combination R is mixed with the target entity C_i and the new combination is obtained. Finally, Eq. 3 is used to calculate the allocation of the region with further distribution of the combination. In this way, the marginal effect of a combined case can be obtained. The final carbon allocation is obtained when we add all the cases together by Eq. 5. The whole process of this step is shown in Figure 4B.

In calculating the marginal effect, this method needs to consider all the combinations of the entities and the marginal effect of the target entity in each combination. However, in the calculation steps of the Owen method, it needs to consider combinations of regions and entities in the region. It means that minimum supply exceeds maximum demand or minimum demand exceeds maximum supply. Hence, the improved Shapley method will exclude the inappropriate combination to reduce computational complexity. The improved method adds a filter function of combinations compared with the original method.



3.3 Design of auction rules

This subsection introduces a method to add the carbon cost based on the allocation of the Owen value method on the quotation in the bilateral market. This mechanism changes the clearing sequence to make low-carbon generations or loads preferentially traded. Considering the characteristics of bilateral auction market clearing, it is necessary to design the quotations of load and power generation entities accordingly.

Then, considering the carbon emissions of load allocation and the additional carbon price, there is a negative correlation, so that the load with more carbon emissions will be ranked lower in the auction, which is consistent with the original goal. The expected additional carbon price for the carbon emissions shared by power generation is positively correlated, so that the units with fewer carbon emissions will accumulate less additional carbon price, and the rank will be relatively high.

The total amount is first calculated through the stepped carbon price in the extra carbon price setting. The carbon price is set to transmit the carbon price of the entity's allocation to its quotation. In this way, the quotation in the final transaction includes economic and environmental factors. If the carbon price is too low, it is impossible to achieve low carbon by changing the quotation order. Therefore, the setting of carbon price should satisfy such that the low-allocation entity can gain an advantage in the quotation. In Section 3.3, the carbon price(40\$/t) is a relatively acceptable empirical value obtained



by many experiments, which is close to the international carbon price. The detailed method is as follows:

$$\begin{cases} c = 5 \times (x_{ci} - x_{\min}), & x_{ci} \le x_{avg}, \\ c = 5 \times (x_{avg} - x_{\min}) + 10 \times (x_{ci} - x_{avg}), & x_{\min} \le x_{ci} \le x_{avg}, \end{cases}$$
(7)

where *c* represents the whole cost for the carbon allocation, x_{ci} represents the carbon allocation of generation or load entity, x_{\min} represents the minimum allocation in generation or load entities, and x_{avg} represents the average allocation of generation or load entities. Then, the additional carbon price is set to

TABLE 1 Configuration of each node port.

| Entity_name | Container_name | Port | |
|-------------|------------------------|------|--|
| Organizer | peer0.org1.example.com | 7051 | |
| G1 | peer1.org1.example.com | 7151 | |
| G2 | peer2.org1.example.com | 7251 | |
| G3 | peer3.org1.example.com | 7351 | |
| G4 | peer4.org1.example.com | 7451 | |
| G5 | peer5.org1.example.com | 7551 | |
| L1 | peer6.org1.example.com | 7651 | |
| L2 | peer7.org1.example.com | 7751 | |
| L3 | peer8.org1.example.com | 7851 | |
| L4 | peer0.org1.example.com | 7951 | |

$$\lambda_{ci} = \frac{c}{p_{ci}},\tag{8}$$

where p_{ci} represents the output or load allocated in the original bilateral bidding scenario. Then, the latest quotation of the load entity is $\lambda_{Di} = \lambda_{Di0} - \lambda_{Li}$, and the latest quotation of the power generation entity is $\lambda_{Gi} = \lambda_{Gi0} + \lambda_{Gi}$.

3.4 Application of blockchain technology

In the market transaction of the proposed model, blockchain technology is used in the whole transaction process. Since the transaction process involves multiple regions, many transaction entities, steps, and blockchain technology makes the whole process traceable and simple while preventing tampering, which is suitable for the intermediate quotation change step.

Due to the previous calculation of the Owen value method, the blockchain application is divided into a twolevel structure. The target of the sub-chain is to collect the quotation information and deliver the transaction results. Thus, it is aimed at the primary users, such as generation or load entities. The target of the main chain is to evaluate the allocation and supervise the performance of the users. Thus, it is aimed at the market organizers. Meanwhile, it can

| Header | "number" 5 | "previous_hash" :'sydv453a0737bdi 963jnv11k521yu53 29iuy3361f, | "data_hash" :' sydv453a0737b di963jnv11k521 yu5329iuy3361 f' | Header | "number" :12 | "previous_hash" :' gR91q43wk4u7 z16ypv93m0qm78 dyo50sg61i53s', | "data_hash" b91mb82b02v 02nb71b71n80 bp5qb8wucm2 bh ' |
|--------|--|---|--|--------|---|---|---|
| - | Signature <buffer 05="" 16="" 34="" 4f="" 79=""></buffer> | | > | Data | Signature <buffer 13="" 26="" 2d="" 3f="" 6e=""></buffer> | | |
| | Payload channel_id:" region_a" | | | | Payload channel_id:" region_a" | | |
| Data | Value: { "G1_NAME" = "G1", "G1_PRICE" = "16", "G1_UPPER" = "500", "G1_LOWER" = "100" "G2_NAME" = "G2", "G2_PRICE" = "18", "G2_UPPER" = "300", "G2_LOWER" = "60" "L1_NAME" = "L1", "L1_PRICE" = "25", "L1_UPPER" = "50", "L1_LOWER" = "20" | | , , , , | | Value: { "G1_NAME" = "G1", "G1_PRICE" = "16", "G1_GENERATION" = "100 "G2_NAME" = "G2", "G2_PRICE" = "18", "G2_GENERATION" = "60", "L1_NAME" = "L1", "L1_PRICE" = "25", "L1_CONSUMPTION" = "50", | | |
| | | • | } | | | • | } |

Schematic diagram of block information. (A) Transaction preparation step. (B) Settlement step.

reduce the risk by setting up a separate sub-chain in each region when some sub-chain encounters a mistake.

Therefore, when building the blockchain and establishing a two-level structure, the main chain of the total area and the subchain of a single region are set, respectively, as shown in Figure 5. The members of the sub-chain consist of load entities, generation entities, and market organizers in the region, while the members of the main chain consist of market organizers of all regions. They all have a set of the ledger separately and, by using a consensus mechanism, all the members keep the ledgers identical when the transaction is finished. The Kafka algorithm used in the Fabric 1.4 version of the consensus mechanism is adopted. After the peer node requests endorsement, all transaction information is sorted through the Kafka cluster, and then the block is packaged through the order node. Finally, each peer node verifies the correctness of the block (Androulaki et al., 2018).

In the sub-chain of each region, each market entity represents a node in the blockchain system. At the same time, a super node is saved, which is the market organizer of every region for the information interaction with the main chain and the supervision and management of the sub-chain. The functions in the smart contract of the sub-chain are as follows:

- 1) Record the public key of the market organizer on the subchain.
- 2) Realize the quotations and relevant data of users reporting and record the information after encrypting the aforementioned data with the public key.
- Record the modification quotation and the transaction results on the chain.
- Realize query of final results for all users and decoding of quotations for market organizers through their private key.

The main chain comprises market organizers, mainly used to record the data interacted through the cross-chain and change the quotation through the smart contract in the main chain. Since market organizers join both blockchains, they can be seen as the bridge between the main chain and sub-chain. Relevant operations are implemented through smart contracts. The functions in the smart contract of the main chain are as follows:

- 1) Record the quotations of all regions on the main chain through market organizers.
- 2) Calculate the carbon allocation of all entities according to the reporting data through the Owen method and modify the quotation based on allocation.
- 3) Fulfill the final transaction by the modification quotations.
- 4) Realize the query of final results for market organizers.

The process is divided into three steps: the quotation report, the quotation modification, and the transaction settlement.

In the first step as shown in Figure 6, by querying transaction notifications in the main chain, the information is transmitted to each market entity in the sub-chain through cross-chain operations of market organizers. The organizers in each region use the RSA algorithm to generate a new public and private key and broadcast the public key to the region by sub-chain. Each market entity reports the quotation and the maximum and minimum values of its demand or output according to its situation and encrypts it with the public key of the super node.

In the second step as shown in Figure 7, the super node of each regional sub-chain decrypts the reporting data by the market entity through its private key and uploads it to the main chain. The smart contract in the main chain uses the data to modify the quotation and record those in the main chain. Then, the transaction is carried out according to the new quotation. Finally, super nodes will transmit the clearing result and the changed quotation to the sub-chain.

In the last step, the clearing results are transferred from the main chain to each sub-chain. Each market entity on the sub-chain finishes the settlement of electricity or funds according to the final result. Relevant power generation entities are dispatched according to requirements. The settlement of the final electricity and the funds' performance is written into each region's blockchain.

4 Case study

The example analysis in this section is divided into two parts. The first part is in the PJM-5 system simulation, using five nodes to simulate five areas. Then, each power generation entity and load entity are set up in different groups. Carbon allocation is carried out through the Owen value method and transmitted to the quotations of each entity as a price signal. Considering the typical daily output characteristics of renewable energy units, the transactions of 24 times a day are simulated. After adopting this method, the impact on the overall carbon emissions and the changes in the dispatch of low-carbon or zero-carbon units are analyzed. The second part implements the aforementioned process in the blockchain by using the Hyperledger Fabric and verifies the practicability of blockchain technology by setting up a two-level structure of the main chain and sub-chain.

4.1 PJM-5 node example

The network structure of the PJM-5 system is shown in Figure 8, where each node is regarded as a region. G3 and G14 are photovoltaic power generation; G4, G9, and G15 are wind power generation; and G11 and G12 are natural gas generators. The remaining units are conventional thermal power units. The AB and ED lines' capacity is 400 and 240 MW, respectively. There is no capacity limit on other lines.

The capacity of two photovoltaic power generations is 200 MW; and the capacity of three wind power generations is 200 MW. Their typical day output curves are shown in Figure 9.

Within 24 h, the load and power generations declare the quotations separately. If the transaction is formed according to the bilateral bidding mode, the consumption of each load and the output of each unit are shown in Figures 10A, C and through the Owen value method, carbon allocation is carried out on 24 trading points. The results are shown in Figures 10B, D. It can be seen that the carbon allocation result is positively correlated with the assumed original bilateral market clearing result. Loads with higher power consumption have higher carbon allocation results, and thermal power units with higher power generation also have higher carbon allocations. The results show that the hydropower units have a negative allocation lower than the photovoltaic or wind power. The reason is that the calculation of carbon allocation is based on the marginal carbon effect in the hypothetical bilateral bidding scenario. Due to the lower quotation of hydropower units, hydropower units are easier to be traded in the bilateral market. The participation of the hydropower unit will reduce the carbon emissions of the entire system, so its allocation is lower. However, due to their high quotations of renewable units, they are not easily traded in the bilateral market. Its participation will not change the output distribution of units, so its allocation is higher than that of hydropower units.

Next, each entity will tier carbon prices based on the Owen value allocation in each trading period. A ladder carbon price of 40\$/t, 20\$/t, and 0\$/t is set according to the maximum allocation, average allocation, and minimum allocation among entities of the same type, respectively. The subsequent clearing is carried out by adding the ladder carbon price to the original quotation. The change in carbon emissions of the entire system is shown in Figure 11. It can be seen that this method can promote the utilization rate of the low-carbon and zero-carbon units and reduce the carbon emissions of the system effectively. The results show that calculating the carbon allocation and considering it as a cost to loads and power generations. On the one hand, an additional carbon price is added to the quotation of power generations so that the environmental value of zero-carbon or low-carbon generators can be exerted and their priority dispatch can be realized. On the other hand, after allocating carbon to loads and attaching a carbon price,

according to the results of their quotations, the expected usage of loads with excessively allocated carbon emissions will be reduced in the bilateral market, so as to guide load entities to energy conservation and emission reduction.

4.2 Implementation based on blockchain technology

Hyperledger Fabric 1.4 version is used in the process of building a blockchain. The operating environment is Linux CentOS 7.5. According to the operation characteristics of the fabric, the smart contract and ledger in each channel will be shared. Therefore, different nodes will be grouped into other channels to realize the construction of different blockchains. As shown in Figure 12, P1 represents the nodes of the power generation entity, P2 represents the nodes of the load entity, and the number of entities determines the number of nodes. P0 represents the organizational nodes in each region. Channels 1-5 represent the sub-chains of five regions; Channel 0 represents the total chain for the entire region. At the same time, because the market organizer in each region is located in the sub-chain of its region and the general chain of the whole region, it has not only the account book of its region but also the account book of the general chain. So it acts as a bridge between the total chain and the sub-chain.

In the application of the sub-chain, it mainly involves the transaction information reporting step and the last transaction step. Therefore, region A was taken as an example to display the relevant results. In region A, there are five power generation nodes, four load nodes, and an organizer's super node. The simulation method is a single machine with a multi-node network. Each node port is set as follows in Table 1.

Two blocks representing two steps in the sub-chain are selected, as shown in Figure 13. A block consists of a header, data, and metadata in blockchain technology. The relevant hash value and block number are displayed in the header. The data in the middle show the signature and value, etc., where the value is parsed by converting the relevant data into the JSON format. Metadata have little effect on the application, which is not shown in the figure.

According to the analysis of the results, it can be seen that the adoption of blockchain technology can automate the entire process. The process involves multiple steps, but it can be finished quickly. In addition, the data of the whole process are recorded on the blockchain. This makes the entire quote change process traceable and credible. Finally, due to the consensus mechanism, all data are tamper-proof, thus ensuring the secure and smooth operation of the market.

5 Conclusion

This article proposes a method to allocate the carbon emissions of the power system to the power generation

entities and load entities according to the Owen value method region-entity principle and establish a model that can realize the transaction of the allocation result as a price signal in the bilateral trading market in this period. Based on the example analysis, the following conclusions are drawn:

- The allocation results of the Owen value method show that its allocation to the power generation and load sides is more reasonable. For different generation units The allocation results correspond to the different generation units' carbon emissions from electricity generation. For other loads, the allocations are consistent with their consumption behavior. At the same time, the difference in carbon emissions caused by geographical factors caused by line capacity is also considered.
- 2) In the bilateral bidding scene, the allocation calculated by the Owen value method is used as a price signal to update entity quotations, which can realize the combination of policy and market mechanisms to realize the dispatch of low-carbon and zero-carbon generators so as to realize the low-carbon economic operation of the whole system.
- 3) The two-level blockchain structure can fit with the entire process. On the one hand, it realizes the contract by a code during the entire process, thereby simplifying the process. On the other hand, during changes in the quotation, it realizes the record of the overall data on the chain to ensure the credibility and traceability of the entire process.

Data availability statement

The raw data supporting the conclusion of this article will be made available by the authors, without undue reservation.

Author contributions

RL and ZL contributed to the conception and design of the study. RL organized the simulation, analyzed the results, and wrote parts of the manuscript. HW gave suggestions to technical support. HD and HJ contributed to writing parts of the manuscript. CX contributed to the improvement of the study design.

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Conflict of interest

Authors HD and CX was employed by the company Electric Power Research Institute of State Grid Zhejiang

Electric Power Co., Ltd., Hangzhou, China. HJ was employed by the company Power Dispatching and Control Center, State Grid Heilongjiang Co., Ltd., Harbin, China.

The remaining 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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