A Transition Mechanism for the Participation of Renewable Energy Generation Companies in Competitive Electricity Spot Markets

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Under the China’s ‘dual carbon’ national goal—reaching peak carbon emissions by 2030 and achieving carbon neutrality by 2060, one of the key issues in China is how to smoothly transit from a fixed-price mode to a competitive market pricing mode for renewable energy generation companies. Aiming at minimizing governmental subsidies and maximizing the fairness among renewable energy generation companies, a multi-agent three-layer transition mechanism with the transactions of green certificates considered is proposed in this paper. Through adjusting subsidy policies, the developed transition mechanism can stimulate the renewable energy generation companies to gradually participate in the competitive electricity spot market. Specifically, a multi-market multi-agent transaction framework in the transition mechanism is first established. Then, in order to derive the important parameters of the transition mechanism, a method that decouples the electricity market and the green certificate market is designed. Finally, the feasibility and efficiency of the proposed transition mechanism are demonstrated through numerical examples.

Keywords: carbon neutral, renewable energy generation, electricity spot market, tradable green certificate, contract coverage

1 INTRODUCTION

The Renewable Energy Sources (RESs) in China have been developing rapidly with the goal of reaching peak carbon emissions by 2030 and achieving carbon neutrality by 2060 (The State Council Information Office of the People’s Republic of China, 2020). By the end of 2020, the total installed capacity of RES-based generation in China reached 530 million kW, accounting for 25.5% of the total generation installed capacity. The rapid development of RESs also results in the large amount of subsidies to RES investors. In order to mitigate financial burdens on the government due to RES subsidies, the energy administrative authorities have been continuously lowering the benchmark electricity prices for photovoltaic and wind power generation. In June 2021, the National Development and Reform Commission issued a relevant document (National Development and Reform Commission of China, 2020), stipulating that the central government would no longer subsidize the newly registered centralized photovoltaic power plants, industrial and commercial distributed photovoltaic projects, and newly approved onshore wind power projects in 2021. This will help to alleviate the financial burden, but it may lead to a sharp decline in the growth rate of RES installations.
In order to alleviate the financial burden of subsidy while promoting the development of renewable energy generation, mature electricity markets have taken a number of measures. The German Ministry of Energy adopted the policy of feed-in premium in 2014, which guide RESs to participate in the competition in the electricity spot market (Schallenberg-Rodriguez and Haas, 2012). It is a transitional way for RESs to gradually shift from full acquisition to full competition (Gawel and Purkus, 2013). In contrast, the United States uses the renewable energy quota system to promote the development of RESs (Barbose et al., 2016; Luo, 2016). At present, more than 30 states in the United States have established a renewable energy quota system, which is conducive to encouraging RESs to adjust its own output according to market supply and demand, and reduce the pressure on system operation. The United Kingdom has implemented a policy of renewable obligation (RO) since 2002, which is similar to a quota system (García-Alvarez et al., 2017). Since the implementation of the RO policy, the RESs has developed rapidly in United Kingdom, and the average annual growth rate of power generation has reached 12%. In addition, both the United States and the United Kingdom allow RES units to participate in the electricity spot market (Kilinc-Ata, 2016). Overall, the combination of “financial contract + spot market” and quota system is often adopted in practical electricity markets around the world (Ren et al., 2022), in which RES units participate in the electricity spot market equally with other types of generation units, and the risks of spot market prices are hedged by signing Contracts For Difference (CFD) between power supply companies and RES generation companies. Meanwhile, the establishment of a marketplace for Tradable Green Certificates (TGCs) is also deployed as an important measure to encourage the development of RES generation. The fixed-price procurement of RES generation as a subsidy for RES generation can be gradually replaced by governmental compensations to TGC transactions. This will not only help mitigate the governmental financial burdens but also promote the participation of RES generation units in competitive electricity spot markets, which is align with the requirements of sustainable development.

Currently, there is still no effective market mechanism to support the participation of RES generation units in electricity spot markets in China. On the one hand, the income of RES-based generation companies cannot be guaranteed after the participation, which will damage the confidence of RES-based generation investors, and is not conducive to the long-term RES generation development. On the other hand, the participation of RES generation units in the electricity spot market may increase the market volatility and lead to severe fluctuations in spot market prices (Xu et al., 2020; Yang et al., 2020). Therefore, the participation of RES generation units in electricity spot markets requires an effective transition mechanism.

Some publications are available on the participation mechanism/ mode of RES generation units in electricity spot markets. Specifically, the impacts of RESs on electricity spot market due to the uncertainty are examined from both theoretical and empirical perspectives in (Li and Xu, 2021; Zhao et al., 2021). An electricity market transition model considering is proposed in (Shinde et al., 2021) to represent a possible requirement to undertake system balancing with increasing amounts of intermittent RESs. A bilateral transaction model based on the Bayesian game is proposed in (Kong et al., 2021) to calculate the Bayes-Nash equilibrium point of the electricity spot market with high penetration RESs, thereby evaluating the smoothness of the participation of RESs in the electricity spot market. It is pointed out in (Gu, 2020) that the government-authorized contract system needs to be adopted collaboratively to ensure the accommodation of RES generation and external electricity, and the percentage of electricity covered by government-authorized contracts should be adjusted reasonably to achieve a smooth transition to the competitive market is proposed. A transition mechanism for the market participation of RESs based generation units, which can gradually guide the transition of RES generation units to the electricity spot market by adjusting the percentage of electricity covered by medium- and long-term contracts is proposed in (Dai and Chen, 2020). However, the specific method for determining the percentage of electricity is not given.

In addition, none of the above publications considers the impacts of TGC transactions on the spot electricity market participation of RES generation units. Since 2017, a green certificate voluntary subscription market has been in operation in China, laying the foundation for the implementation of the mandatory quota system. As an important measure to support the quota system, TGC trading has a positive effect on reducing the burdens on government financial subsidies and restructuring the revenue of renewable energy generation. A system dynamic model is established, and case studies are conducted in (Zhang et al., 2021; Zhu et al., 2022), pointing out that compared with fixed price purchases, the development efficiency and economic benefits of RES generation units under the quota system are higher. A study that analyzes the overall framework, core elements and supporting measures of the quota system in recent years together with China’s current national conditions, and puts forward a dual-track system of “fixed electricity price + quota system” is established in (Jiang et al., 2020). Another study that shows the role of the penalty mechanism of the quota system and points out that setting a reasonable penalty can support the market mechanism to improve the income of RES generation units, thereby reducing the financial burden of subsidies on the government is proposed in (Zhang et al., 2017).

In order for the sustainable development of renewable energy generation and the relief of the government financial burdens, it is of great significance to explore and study the transition mechanism for RES units to participate in the electricity spot market. To fill this knowledge gap, the main contributions of this paper are summarized as follows:

1) A new transition mechanism is proposed to reduce the financial burden of governmental subsidies and guide the participation of RES units in the electricity spot market.
2) A multi-agent three-layer optimal decision-making model is established to calculate the contract coverage ratio (CCR). CCR denotes the percentage of electricity traded via CFDs over the total on-grid electricity and is an important parameter in the transition mechanism. The method for determining the CCR of RES units is proposed to maximize the income of generators while taking into
account the clearing results of the electricity spot market and the green certificate market.

3) Extensive numerical experiments using practical electricity market data are conducted to demonstrate the feasibility and efficiency of the proposed transition mechanism and solving methods. Both short-term and medium-long-term simulations are carried out in the experiments.

The rest of paper is organized as follows: Section 2 introduces the established transition mechanism for RES units to participate in the electricity spot market. Section 3 presents the method for determining the CCR. Extensive numerical experiments are employed to illustrate the proposed model in Section 4, and the simulation results are also analyzed. Finally, the paper is concluded in Section 5.

2 TRANSITION MECHANISM FOR THE PARTICIPATION OF RENEWABLE ENERGY GENERATORS IN THE ELECTRICITY SPOT MARKET

How to balance the financial burdens of government subsidies and the incentives for renewable energy development is an important issue in the electricity market reform. In the early stage of market development, RES generators face difficulties due to their weak market competitiveness, immature technologies, and recovery of investment costs, thus government financial subsidies are needed (Upton and Snyder, 2017). During this period, the government can sign CFDs with the RES generators to ensure their income and encourage their continued development. With the continuous progress of the electricity market reform, the number of RES generation companies is increasing, and the total amount of government subsidies will be increasing accordingly. In order to avoid such a situation, it is necessary to gradually reduce the CFDs for RES units, and guide RES units to participate in the electricity spot market. Meanwhile, through cooperation with the TGC market, the subsidies for RES units can be gradually shifted from government financial subsidies to TGC transaction subsidies (Ma et al., 2017), so as to encourage RES companies to carry out technological innovation while ensuring the consumption of green electricity. The market transaction framework of the proposed transition mechanism is shown in Figure 1.

In the above framework, the income of RES units is divided into three parts, namely the income from CFDs, the income of selling electricity in the spot market, and the income of selling TGCs. According to existing policy (National Development and Reform Commission of China, 2020), depending on the number...
TABLE 1 | Comparison of the two types of RES units.

| RES units | Project registered time | CFD price                                      | Number of TGCs issued |
|-----------|--------------------------|-----------------------------------------------|-----------------------|
| Type I    | Before 2021              | Including the governmental premium subsidy, which is higher than the local benchmark price of coal-fired power generation | Issued for electricity traded in the spot market |
| Type II   | In and after 2021        | Local benchmark price of coal-fired power generation | Issued for all of the RES output |

of issued TGCs and the price of CFDs, the calculation rules for RES units with different registered times are different, as shown in Table 1. Therefore, this paper categorizes RES units into two types.

The RES units in type I are those which were put into operation before 2021, and their CFD price is higher than the local benchmark price of coal-fired power generation. For these RES units, the government will subsidize for the generation output covered in the CFDs, and TGCs will be issued only for the electricity traded in the spot market. It means that only the RES generations traded in the spot market are compensated through the TGC market.

The RES units in type II are those to be put into operation after 2021, whose contract purchase price is specified as the benchmark price of local coal-fired power generation, and the CFDs of such RES units do not receive governmental subsidies. For these RES units, TGCs are issued for the electricity in both the CFDs and the spot market, thus the RES generations traded in both the CFDs and the spot market are compensated.

The CCR of RES units in the proposed transition mechanism is defined as the percentage of electricity traded via CFDs over the total on-grid electricity, as shown in Eq. 1.

$$\gamma_F = \frac{Q_F}{Q_{sp} + Q_F}$$

where $\gamma_F$ is the CCR (%); $Q_F$ is the electricity covered by CFDs (MWh); $Q_{sp}$ is the amount of electricity allowed to be traded in the electricity spot market (MWh).

The CCR stipulates the share of RES generations entering the spot market, and also determines the amount of government financial subsidies. The CCR is a key parameter and should be determined carefully considering the system reliability requirements, the reform process of the power market and other factors. The determination of CFD coverage is also related to the revenue of RES units and the number of TGCs issued, and may affect the market strategies of fossil-fueled generation units in the spot market and TGC market. Therefore, designing an innovative method for determining the contract coverage is the key to a smooth transition of the reform.

It should be noted that the continuous development of renewable energy generation will greatly increase the uncertainty of the power system. Correspondingly, the demand for ancillary market will also increase. This will inevitably have an impact on the clearing results of the electricity spot market and the market strategies of fossil-fueled generation units. But the situation discussed in this paper is in the early stage of RES development. In other words, the impact of participation of RESs in the electricity spot market on ancillary market is not fatal. Notably, the focus of this paper is on the participation of RES in electricity energy markets, and the extension of the proposed mechanism to the ancillary market will be systematically studied in our future work.

3 MATHEMATICAL FORMULATIONS OF PROPOSED TRANSITION MECHANISM

The ratio of RES electricity generations covered by CFD contracts directly affects the amount of government financial subsidies and the income of RES units, and it is closely related to the clearing results of the spot market and the TGC market. In the transition mechanism, a three-layer optimization model is established, where the objective is to maximize the revenues for both RES units and fossil-fueled units while taking into account the outcomes of the spot market and the TGC market. The method for solving the optimal CCR is also presented.

Figure 2 illustrates the structure of decision-making model in the proposed transition mechanism. Specifically, the input parameters of the three-layer model include the predicted generation output, typical load curve, elasticity of load demand, generation constraints, network constraints, fuel costs of fossil-fueled units, and assessment weights of the quota system. In the upper-level model, the research object is the policymaker, the decision variable is the contract coverage of RES units, and the objective function is to minimize government financial subsidies and to maximize the fairness of subsidies. The contract coverage obtained in the upper model is passed to the middle-level model. The middle-level model includes the clearing models of both the electricity spot market and the green certificate market, which are managed by the Independent System Operator (ISO). The middle-level model feeds back the market clearing results to the upper model via simulations of the electricity spot market and the green certificate market. The lower-level model formulates the bidding strategies of RES and fossil-fueled units aiming at maximizing their income, and the optimal bidding results will be fed back to the middle-level model to complete the simulation of market operation.

3.1 Upper-Level: Decision-Making Model for Policymaker

The decision-making model for policymaker is to solve the CCR. The CCR not only determines the amount of financial subsidies, but also affects the fairness among RES units. Therefore, the upper-level model adopts a multi-objective optimization method
to minimize government financial subsidies while maximizing the fairness. In this section, photovoltaic (PV) generators and wind turbines are taken as examples to elaborate details of the upper-level model.

3.1.1 Objective Function

3.1.1.1 Government Financial Subsidies

Alleviating the financial burdens of governmental subsidies is the first objective of this model, which is measured by the premium CFDs for RES units. It is modeled here that the difference between the CFD price and the spot market price is the premium value of government subsidies. Therefore, the objective function based on the amount of government financial subsidies can be expressed as follows.

$$\min f_{p1} = \sum_{t=1}^{T} \left( P_{p1}^s - P_{p1}^{Cp} \right) \cdot Q_{p1,t}^{en} \cdot \gamma_p$$

(2)

where \( f_{p1} \) is the amount of government financial subsidy to the photovoltaic generation; \( P_{p1}^s \) is the spot market price during the trading period \( t \); \( P_{p1}^{Cp} \) and \( Q_{p1,t}^{en} \) are the CFD price of the type I photovoltaic generators and the predicted output during the trading period \( t \), respectively; \( \gamma_p \) is the photovoltaic contract coverage in the decision-making period \( T \). As mentioned earlier, based on existing policies, the government only subsidizes the type-I RES generation units. For the type-II units, the government financial subsidy is 0.

3.1.1.2 Fairness in the Benefits Among Renewable Energy Sources Generation Units

The fairness of the governmental subsidy is crucial for the policymaker. While maximizing the overall benefit of RES units, it is necessary to ensure that each unit can benefit from the transition mechanism as much as possible. Therefore, another goal of the upper-level model is to maximize the fairness of income among the same type of units. The concept of unit power generation profit as defined in Eq. 3 is introduced to measure the fairness.

$$K_{pi,t_d} = \frac{\sum_{t=1}^{T_d} \left[ P_{pi}^s Q_{pi,t}^{en} + P_{pi}^{Cp} Q_{pi,t}^{en} (1 - \gamma_p) + P_{pi}^{gr} Q_{pi,t}^{gr} \right] - \varpi_{pi} C_p}{\sum_{t=1}^{T_d} Q_{pi,t}^{en}}$$

(3)

where \( K_{pi,t_d} \) is the unit power generation profit of the \( i^{th} \) photovoltaic generator in trading period \( t_d \); \( t_d \) is the time period for calculating the unit power generation profit, which can avoid the problem that the denominator of Eq. 3 is 0 when there is no photovoltaic output at night, usually takes \( t_d = 24 \); \( P_{pi}^s \) is the CFD price of the \( i^{th} \) photovoltaic generator; \( Q_{pi,t}^{en} \) and \( Q_{pi,t}^{gr} \) are the predicted output of the \( i^{th} \) photovoltaic generator and the amount of electricity traded in the TGC market in the trading period \( t \), respectively; \( C_p \) is the levelized cost of photovoltaic generation; \( \varpi_{pi} \) is the efficiency coefficient of the \( i^{th} \) photovoltaic generator, which is assessed by professional organizations. When the power generation efficiency is higher than the average level, \( \varpi_{pi} > 1 \).

To ensure that the RES units of the same type have the smallest variance in the unit power generation profit in all trading periods, the objective function to measure the fairness of the RES profit can be expressed by Eq. 4.

$$\min f_{p2} = \sum_{p=1}^{N_p} \sum_{t_d=1}^{T} \left( K_{pi,t_d} - \frac{1}{N_p} \sum_{p=1}^{N_p} K_{pi,t_d} \right)^2$$

(4)
where $f_{p2}$ is the fairness coefficient of photovoltaic generators, i.e., the sum of squared difference of the unit power generation profit of each unit in each trading period; $N_p$ is the number of photovoltaic generators. The unit power generation profit obtained by the same type of RES units in a certain trading period through CFDs, the spot market, and the TGC market is used as an indicator to measure the fairness, which reflects the equal treatment of the same type of units.

Similarly, the objective functions for wind turbines can be formulated by Eqs. 5, 6.

$$\min f_{w1} = \sum_{i=1}^{T} (P_{w1}^i - P_{1}^{p}) \cdot Q_{w1}^{\text{est}} \cdot y_w$$

$$\min f_{w2} = \sum_{i=1}^{T} \sum_{j=1}^{N_w} \left( K_{w,t,d} - \frac{1}{N_w} \sum_{i=1}^{N_w} K_{w,t,d} \right)^2$$

where $f_{w1}$ and $f_{w2}$ are the amount of government financial subsidy and fairness coefficient of the wind turbines, respectively; $P_{w1}^i$ and $Q_{w1}^{\text{est}}$ are the CFD price and the predicted output of wind turbines during the trading period $t$; $y_w$ is the contract coverage of the wind turbines in the decision-making period $T$; $N_w$ is the number of wind turbines; $K_{w,t,d}$ is the unit power generation profit of the $i$th wind turbine in trading period $t$, as calculated in Eq. 7.

$$K_{w,t,d} = \sum_{i=1}^{N_w} \left( P_{w1}^i \cdot Q_{w1}^{\text{m}} \cdot y_w + P_{1}^{p} \cdot Q_{w1}^{\text{est}} (1 - y_w) + P_{1}^{p} \cdot Q_{w1}^{\text{gr}} \right) - \omega_{w} C_w$$

where $P_{w1}^i$ is the CFD price of the $i$th wind turbine; $Q_{w1}^{\text{est}}$ and $Q_{w1}^{\text{gr}}$ are the predicted output of the $i$th wind turbine and the amount of electricity traded in the TGC market in the trading period $t$, respectively; $C_w$ is the leveled cost of wind power generation; $\omega_{w}$ is the efficiency coefficient of the $i$th wind turbine.

### 3.1.2 Constraints and Solution

The output constraints of RES generators are included in the upper-level decision-making model, as shown in Eqs. 8, 9.

$$y_{p,\text{min}} \leq y_p \leq y_{p,\text{max}}$$

$$y_{w,\text{min}} \leq y_w \leq y_{w,\text{max}}$$

where $y_{p,\text{min}}$ and $y_{p,\text{max}}$ are the upper and lower limits of the CCR of photovoltaic generators; $y_{w,\text{min}}$ and $y_{w,\text{max}}$ are the upper and lower limits of the CCR of wind turbines.

With the above objective functions and constraints, a multi-objective optimization model is obtained for solving the CCR of photovoltaic generators and wind turbines. The original objective functions are first normalized to [0,1] to get rid of the influence of dimension and order. Then the weighting factors are added as in Eq. 10.

$$\min f(y_p, y_w) = \sum_{i=1}^{3} \left( a_p \cdot \frac{f_{pi} - f_{pi,\text{min}}}{f_{pi,\text{max}} - f_{pi,\text{min}}} + a_w \cdot \frac{f_{wi} - f_{wi,\text{min}}}{f_{wi,\text{max}} - f_{wi,\text{min}}} \right)$$

where $a_p$ and $a_w$ are the weighting factors of each part of the objective function. They are determined according to factors such as power supply reliability, environmental benefits, policy support for RES development, financial subsidy burdens during this period, etc.

### 3.2 Middle-Level: Market Clearing Model

#### 3.2.1 Electricity Spot Market Clearing Model

The clearing results of the electricity spot market are required when calculating the CCR. Therefore, the electricity spot market clearing model is the first middle-level model to provide market data for decision-making in the upper-level model. Participants in the spot market can be divided into fossil-fueled generation units and renewable energy units. For a trading session $t$, the spot market clearing model is given as follows.

$$\min \sum_{i=1}^{N_p} r_{F,i}^t P_{F,i}^t + \sum_{j=1}^{N_f} r_{G,j}^t P_{G,j}^t$$

s.t. $\sum_{i=1}^{N_p} P_{F,i}^t + \sum_{j=1}^{N_f} P_{G,j}^t = \sum_{k=1}^{N_k} p_{D,k,t}$

$$P_{F,i,t} \leq P_{F,i,t}^\text{max}$$

$$P_{G,j,t} \leq P_{G,j,t}^\text{max}$$

$$\sum_{i=1}^{N_p} P_{F,i,t} + \sum_{j=1}^{N_f} P_{G,j,t} + \sum_{k=1}^{N_k} p_{D,k,t} \leq P_{F,t}^\text{max}$$

$$\sum_{j=1}^{N_f} r_{j}^\text{reserve,} \geq R_{j}^\text{reserve,}$$

$$\sum_{j=1}^{N_f} r_{j}^\text{reserve,} \geq R_{j}^\text{reserve,}$$

where $p_{D,k,t}$ is the power demand of the $k$th load in clearing period $t$; $r_{F,i}^t$ and $r_{G,j}^t$ are the market quotations of the $i$th RES unit and the $j$th fossil-fueled generation unit, respectively; $P_{F,i}^t$ and $P_{G,j}^t$ are the bid-winning power of the $i$th RES unit and the $j$th fossil-fueled generation unit, respectively; $p_{D,k,t}$ is the power transmission distribution factors of the RES unit $i$, fossil-fueled generation unit $j$, and load $k$, respectively; $P_{F,t}^\text{max}$ represents the transmission capacity of line $i$; $N_p$, $N_f$, and $N_f$ are number of RES units, fossil-fueled generation units, and lines, respectively; $r_{j}^\text{reserve,}$ and $r_{j}^\text{reserve,}$ represent the positive spinning reserve and the negative spinning reserve provided by the $j$th fossil-fueled generation unit, respectively, $R_{j}^\text{reserve,}$ and $R_{j}^\text{reserve,}$ represent the positive spinning reserve requirement and the negative spinning reserve requirement of the system, respectively; $P_{F,j}^\text{max}$ and $P_{G,j}^\text{min}$ are the upper and lower limits of the bid-winning power of the fossil-fueled generation unit; $P_{F,i,t}^\text{max}$ and $P_{F,i,t}^\text{min}$ are the upper and lower limits of the bid-winning power of the RES units.

For both the type I and type II RES units, $P_{F,i,t}^\text{max}$ is the generation output minus the power covered by CFDs, as shown in Eq. 18.

$$P_{F,i,t}^\text{max} = \frac{Q_{F,i,t}^m (1 - \gamma)}{t}$$
where $Q_{Fi,t}^r$ is the predicted output of the RES unit in trading period $t$; $y$ is the CCR corresponding to the RES unit, $y = [y_F, y_w]$.

### 3.2.2 Green Certificate Market Clearing Model

The TGC market clearing model is the other middle-level model. During a decision cycle, the TGC market clearing process is repeated. After a market clearing, the winning participants can conduct transactions, and the participants who have not reached a transaction will requote and enter the next market clearing process. Since the clearing results of the spot market and the TGC market affect each other, it is assumed that the clearing cycle of the TGC market is consistent with the spot market. For a trading session $t$, the TGC market clearing model is shown in Eqs. 19–22.

$$\text{max} \sum_{i=1}^{N_F} r_{Fi,t}^r q_{Fi,t}^r - \sum_{j=1}^{N_G} q_{Gj,t}^r G_{Gj,t} \quad (19)$$

$$\text{s.t.} \quad \sum_{i=1}^{N_F} q_{Fi,t}^r = \sum_{j=1}^{N_G} q_{Gj,t}^r \quad (20)$$

$$0 \leq q_{Fi,t}^r \leq q_{Fi,t}^r_{max} \quad \forall i \in [1, N_F] \quad (21)$$

$$0 \leq q_{Gj,t}^r \leq q_{Gj,t}^r_{max} \quad \forall j \in [1, N_G] \quad (22)$$

where $r_{Fi,t}^r / r_{Fi,t}^r$ is the market quotation of the $i$th RES/ $j$th fossil-fueled generation unit; $q_{Fi,t}^r$ and $q_{Fi,t}^r_{max}$ are the number and the upper limit of TGCs offered by RES generation unit, respectively; $q_{Gj,t}^r$ and $q_{Gj,t}^r_{max}$ are the number and the upper limit of TGCs bid by the fossil-fueled generation unit, respectively. $q_{Fi,t}^r$ and $q_{Fi,t}^r_{max}$ corresponds to the output of RES unit $Q_{Fi,t}^r$ and $Q_{Fi,t}^r_{max}$. For type I and type II RES units, the upper limits on electricity corresponding to the number of winning TGCs are different, as shown in Eq. 23.

$$Q_{Fi,t}^r_{max} = Q_{Fi,t}^r \cdot (1 - y)$$

$$Q_{Fi,t}^r_{max} = Q_{Fi,t}^r \quad (23)$$

where $Q_{Fi,t}^r_{max}$ and $Q_{Fi,t}^r_{max}$ represent the upper limits on electricity corresponding to the number of winning TGCs for the type I and type II RES units, respectively; $Q_{Fi,t}^r$ and $Q_{Fi,t}^r$ represent the predicted output of the type I and type II RES units, respectively. Eq. 23 corresponds to the definitions of the two types of RES units.

### 3.3 Lower-Level: Bidding Model of Participants

#### 3.3.1 Bidding Model for Renewable Energy Units

RES units participate in both the electricity spot market and the TGC market, so their income is made up of two parts as shown in Eq. 24.

$$\text{max} E_{Fi,t} = E_{Fi,t}^P (r_{Fi,t}^P) + E_{Fi,t}^r (r_{Fi,t}^r) \quad (24)$$

where $E_{Fi,t}$ is the total revenue of the $i$th RES unit at time $t$; $E_{Fi,t}^P$ and $E_{Fi,t}^r$, are the spot market revenue and TGC market revenue of the $i$th RES unit at time $t$, respectively.

The constraints in the bidding model of RES generation units are shown in Eqs. 25–27.

$$r_{Fi,t}^{Pmin} \leq r_{Fi,t}^P \leq r_{Fi,t}^{Pmax} \quad (25)$$

$$P_{Fi,t}^{min} \leq P_{Fi,t}^P \leq P_{Fi,t}^{max} \quad (26)$$

where Eqs. 25 and 26 are bidding price constraints, and Eq. 27 is the bidding output constraint. $r_{Fi,t}^{Pmax}$ / $r_{Fi,t}^{Pmin}$ and $r_{Fi,t}^{rmax}$ / $r_{Fi,t}^{rmin}$ are the maximum/minimum bidding prices of the $i$th RES unit at time $t$ in the electricity spot market and TGC market; $P_{Fi,t}^{max}$ and $P_{Fi,t}^{min}$ are the maximum and minimum output of the $i$th RES unit at time $t$, respectively.

Notably, RESs are usually coupled with energy storage resources (ESRs). Since this paper discusses the situation in the early stage of RES participation in electricity markets, the penalty mechanism for the generation output deviations of RES units is not considered in this paper. The forecasting results of RES generation outputs could be more accurate if ESRs are modeled, but the attained conclusions will remain unchanged. It is implicitly assumed that the predicted generation outputs of RES units could be accurate enough, even without ESRs installed.

#### 3.3.2 Bidding Model for Fossil-Fueled Generation Units

When fossil-fueled generation units participate in the spot market, they need to purchase TGCs in the TGC market, so their income is also divided into two parts, as shown in Eq. 28.

$$\text{max} E_{Gj,t} = E_{Gj,t}^P (r_{Gj,t}^P) + E_{Gj,t}^r (r_{Gj,t}^r) \quad (28)$$

where $E_{Gj,t}$ is the total revenue of the $j$th fossil-fueled generation unit at time $t$; $E_{Gj,t}^P$ and $E_{Gj,t}^r$ are the spot market revenue and TGC market revenue of the $j$th fossil-fueled generation unit at time $t$, respectively. $E_{Gj,t}^r$ is usually negative.

The constraints of the bidding model for fossil-fueled generation units are shown in Eqs. 29–31. Among them, Eqs. 29, 30 are the bidding constraints, and Eq. 31 is the output constraints.

$$r_{Gj,t}^{Pmin} \leq r_{Gj,t}^P \leq r_{Gj,t}^{Pmax} \quad (29)$$

$$r_{Gj,t}^{rmin} \leq r_{Gj,t}^r \leq r_{Gj,t}^{rmax} \quad (30)$$

$$P_{Gj,t}^{min} \leq P_{Gj,t}^P \leq P_{Gj,t}^{max} \quad (31)$$

where Eqs. 29 and 30 are the bidding price constraints, and Eq. 31 is the bidding output constraint. $r_{Gj,t}^{Pmax}$ / $r_{Gj,t}^{Pmin}$ and $r_{Gj,t}^{rmax}$ / $r_{Gj,t}^{rmin}$ are the maximum/minimum bidding outputs of the $j$th fossil-fueled generation unit at time $t$, respectively.
are the maximum/minimum bidding prices of the \( j \)th fossil-fueled generation unit at time \( t \) in the electricity spot market and TGC market; \( p_{Gj,t}^{\max} \) and \( p_{Gj,t}^{\min} \) are the maximum and minimum output of the \( j \)th fossil-fueled generation unit at time \( t \), respectively.

In the spot market, the market strategies of fossil-fueled generation units not only depends on its production conditions, fuel prices and other factors, but also is affected by the TGC market clearing results. When the TGC market clearing price is high and fossil-fueled units cannot buy enough TGCs, their risk of being punished for not meeting the quota will increase, so they will tend to reduce the value of offering prices.

In the TGC market, the market strategies of fossil-fueled generation units (i.e., TGC buyers) is affected by the clearing results of the spot market. If the clearing electricity in the spot market is low, or the clearing price is low, the incentive of purchasing TGCs will drop. In the long run, the market strategies of fossil-fueled units is also affected by factors such as the approaching assessment date and the overall trend of TGC prices throughout the assessment cycle.

3.4 Reformulation and Solution

Mathematical reformulations are conducted to solve the proposed three-layer model. Since the lower-level model is a linear programming problem, the Karush-Kuhn-Tucker (KKT) condition is a necessary and sufficient condition of optimality. Therefore, the two bidding models in the lower layer can be replaced with their KKT conditions and incorporated into the middle-level model, thereby transforming the original three-layer model into a two-layer optimization model. After substituting Eqs. 34–37 into Eqs. 11–17 and Eqs. 19–22, the Lagrangian functions of the spot market clearing model and the TGC market clearing model are constructed as below.

\[
\Gamma^p = \sum_{t=1}^{T} \left( \sum_{i=1}^{N_T} \left( \frac{\partial p_{Fi,t}^p}{\partial r_{Fi,t}^p} r_{Fi,t}^p + \frac{\partial p_{Fi,t}^p}{\partial q_{Gj,t}} q_{Gj,t} \right) \right) + \sum_{t=1}^{T} \sum_{i=1}^{N_T} \pi_{Fi,t}^p
\]

\[
\Gamma^g = \sum_{t=1}^{T} \left( \sum_{j=1}^{N_G} \left( \frac{\partial q_{Gj,t}^g}{\partial p_{Fi,t}^g} r_{Fi,t}^p + \frac{\partial q_{Gj,t}^g}{\partial q_{Gj,t}} q_{Gj,t} \right) \right) + \sum_{t=1}^{T} \sum_{j=1}^{N_G} \pi_{Gj,t}^g
\]

Thus, the KKT conditions of the lower-level model are derived as Eqs. 34–37.

\[
\frac{\partial \Gamma_{Fi,t}}{\partial r_{Fi,t}^p} = \pi_{Fi,t}^p
\]
\[
\frac{\partial \Gamma_{Fi,t}}{\partial q_{Gj,t}} = \pi_{Gj,t}^g
\]
\[
\frac{\partial \Gamma_{Gj,t}}{\partial p_{Fi,t}^g} = \pi_{Fi,t}^p
\]
\[
\frac{\partial \Gamma_{Gj,t}}{\partial q_{Gj,t}} = \pi_{Gj,t}^g
\]

\[\frac{\partial \Gamma_{Fi,t}}{\partial q_{Gj,t}} = \pi_{Gj,t}^g\]

Based on the above KKT conditions, the lower-level model is transformed into linear constraints and incorporated into the middle-level optimization model.

Similarly, since the two market clearing models in the middle layer are both linear programming problems, the middle-level model can also be replaced with its KKT condition, thereby transforming the original problem into a single-level optimization model.
3.4.2 Inequality Constraints

\[ p_{i,j,t}^p \leq p_{i,\text{max}} \quad \forall i \in [1, N_F], \forall t \in [1, T] \]  \hspace{1cm} (47)

\[ p_{g,j,t}^p \leq p_{g,\text{max}} \quad \forall j \in [1, N_G], \forall t \in [1, T] \]  \hspace{1cm} (48)

\[ \sum_{i=1}^{N_F} p_{i,j,t}^p + \sum_{j=1}^{N_G} p_{g,j,t}^p \leq P_{l,t}^\text{max} \quad \forall l \in [1, N_l], \forall t \in [1, T] \]  \hspace{1cm} (49)

\[ 0 \leq q_{i,j,t}^p \leq q_{i,j,t}^\text{max} \quad \forall i \in [1, N_F], \forall t \in [1, T] \]  \hspace{1cm} (50)

\[ 0 \leq q_{g,j,t}^p \leq q_{g,j,t}^\text{max} \quad \forall j \in [1, N_G], \forall t \in [1, T] \]  \hspace{1cm} (51)

3.4.3 Dual Complementary Constraints

\[ 0 \leq \mu_{i,j,t}^p \perp (p_{i,j,t}^p - p_{i,j,t}^\text{min}) \geq 0 \quad \forall i \in [1, N_F], \forall t \in [1, T] \]  \hspace{1cm} (52)

\[ 0 \leq \mu_{i,j,t}^p \perp (p_{i,j,t}^p - p_{i,j,t}^\text{max}) \geq 0 \quad \forall i \in [1, N_F], \forall t \in [1, T] \]  \hspace{1cm} (53)

\[ 0 \leq \mu_{g,j,t}^p \perp (p_{g,j,t}^p - p_{g,j,t}^\text{min}) \geq 0 \quad \forall j \in [1, N_G], \forall t \in [1, T] \]  \hspace{1cm} (54)

\[ 0 \leq \mu_{g,j,t}^p \perp (p_{g,j,t}^p - p_{g,j,t}^\text{max}) \geq 0 \quad \forall j \in [1, N_G], \forall t \in [1, T] \]  \hspace{1cm} (55)

\[ 0 \leq v_{i,j,t} \perp \left( P_{l,t}^\text{max} - \sum_{i=1}^{N_F} p_{i,j,t}^p - \sum_{j=1}^{N_G} p_{g,j,t}^p \right) \geq 0 \quad \forall l \in [1, N_l], \forall t \in [1, T] \]  \hspace{1cm} (56)

\[ 0 \leq \mu_{g,j,t}^q \perp (q_{g,j,t}^q - q_{g,j,t}^\text{min}) \geq 0 \quad \forall j \in [1, N_G], \forall t \in [1, T] \]  \hspace{1cm} (57)

\[ 0 \leq \mu_{g,j,t}^q \perp (q_{g,j,t}^q - q_{g,j,t}^\text{max}) \geq 0 \quad \forall j \in [1, N_G], \forall t \in [1, T] \]  \hspace{1cm} (58)

\[ 0 \leq \mu_{g,j,t}^q \perp (q_{g,j,t}^q - q_{g,j,t}^\text{min}) \geq 0 \quad \forall j \in [1, N_G], \forall t \in [1, T] \]  \hspace{1cm} (59)

\[ 0 \leq \mu_{g,j,t}^q \perp (q_{g,j,t}^q - q_{g,j,t}^\text{max}) \geq 0 \quad \forall j \in [1, N_G], \forall t \in [1, T] \]  \hspace{1cm} (60)

where \( 0 \leq x \perp y \geq 0 \) means that at most one of the scalars \( x \) and \( y \) can be strictly greater than 0.

For the dual complementary constraints in Eqs. 52–60, the big-M method can be used to convert them into linear constraints. For example, a binary variable \( r \) is introduced to transform Eq. 52 into Eq. 61 and Eq. 62.

\[ 0 \leq \mu_{i,j,t}^p \leq Mr_{i,j,t}^p \quad \forall i \in [1, N_F], \forall t \in [1, T] \]  \hspace{1cm} (61)

\[ 0 \leq (p_{i,j,t}^p - p_{i,j,t}^\text{min}) \leq M(1 - r_{i,j,t}^p) \quad \forall i \in [1, N_F], \forall t \in [1, T] \]  \hspace{1cm} (62)

where \( M \) is a sufficiently large positive number.

With the above KKT conditions, the middle- and lower-level models are transformed into linear constraints and incorporated into the upper-level optimization model. At this point, the three-layer optimization model to determine the contract coverage of RES units has been transformed into a Mixed-Integer Linear Programming (MILP) problem, which can be solved by the GUROBI solver in MATLAB, and finally obtain the optimal CCR for RES generation units.

4 NUMERICAL EXAMPLES AND RESULTS

4.1 Data Specifications

Numerical experiments are performed using the real-world data of the electricity market in a province of eastern China. It is assumed that the RES units begin to participate in the electricity spot market in 2020. The fossil-fueled generation units are coal-fired units, and the RES units include photovoltaics and wind turbines. The parameters of each unit in 2020 are shown in Table 2. The RES units in 2020 all belonged to type I; the number of type-II units started increasing in 2021, and the annual growth rate is set according to governmental policies (The State Council Information Office of the People’s Republic of China, 2020). The CFD price for the type-II RES units is the same as for coal-fired units. The RES output is predicted based on the historical data, and the annual load demand curve is obtained based on the load curve of that province in 2020. The responsibility weight of the quota system for coal-fired units is set according to governmental policies (National Development and Reform Commission and National Energy Administration Commission, 2021). The lower and upper limits of the clearing price in the spot market are set to -200 yuan/MWh and 1000 yuan/MWh respectively, and the lower and upper limits of the clearing price in the TGC market are set to 0 yuan/MWh and 800 yuan/MWh respectively.

In this experiment, the CCR in the peak period, flat period, and valley period of each day for photovoltaic generators and wind turbines will be determined, respectively. Furthermore, the comprehensive CCR is introduced to reflect the overall CCR of RES units, which is a weighted average of the contract coverage at different times, as shown in Eq. 63.

\[ y_c = \frac{y_p Q_{p,F}^m + y_a Q_{a,F}^m + y_g Q_{g,F}^m}{Q_{p,F}^m + Q_{a,F}^m + Q_{g,F}^m} \]  \hspace{1cm} (63)

where \( y_c \) is the comprehensive CCR; \( y_p, y_a, \) and \( y_g \) are the CCR of RES generation units in the peak, flat, and valley periods, respectively; \( Q_{p,F}^m, Q_{a,F}^m, \) and \( Q_{g,F}^m \) are the generation outputs of RES generation units during the peak, flat and valley periods, respectively.

4.2 Analysis of Short-Term Simulation Results

The changes in CCR in 2020 are simulated by adjusting the weighting factors of the decision-making model, as shown in Figure 3 and Figure 4. Changes in policy, load supply, and demand are simulated by changing the ratio of \( a_{p2}/a_{p1} \) and \( a_{a2}/a_{a1} \). The decrease of \( a_{p2}/a_{p1} \) and \( a_{a2}/a_{a1} \) indicates that the determination of CCR is more inclined to ease the financial burden of the government, and the increase indicates that it is more inclined to the fairness of the RES units. Affected by the epidemic, the load demand is at a low level from January to April in 2020. At this time, the clearing price of the spot market is relatively low. In order to protect the interests of RES units and promote their development, \( a_{p2}/a_{p1} \) and \( a_{a2}/a_{a1} \) should increase. From June to September, the load demand is at a high level, the spot market clearing price is relatively high. In addition, due to
seasonal factors, the output of RES units increases, meaning that RES units can earn more from the spot market and the TGC market, so $\frac{a_{w2}}{a_{w1}}$ and $\frac{a_{w2}}{a_{w1}}$ should be reduced to ease the government’s financial burden.

It can be seen in Figure 3 and Figure 4 that the comprehensive CCR of photovoltaics and wind turbines is consistent with the trend of the weighting factors. This indicates that the CCR can achieve the expected effect by adjusting the weighting factors. Specifically, at the beginning of the year, the contract coverage of RES units remains at a high level since the winter is the peak period, and then it decreases significantly in the summer. This allows RES units to maintain a high level of income through high-coverage CFDs in winter. When the spot market price is higher in summer, the CCR of RES generations is low and thus the remaining RES outputs can opt to participate in the spot market. This is consistent with the intention of setting the weighting factors. The output of photovoltaics mainly concentrates in the middle of the load, so it needs to be maintained at a high level to ensure the income. The output of wind turbines in each load period fluctuates greatly due to the uncertainty of wind power, so its CCR varies more. The contract coverage of photovoltaics in the load valley period is extreme, because the output of photovoltaics in the valley period is extremely small, whose impact on the objective function is limited.

The results of income and fairness coefficients for RES units are compared between proposed method and the fixed contract coverage method, as shown in Table 3. For the income analysis, fixing the CCR will increase the government’s financial subsidy but reduce the income of RES units in the spot market and the TGC market. For the fairness analysis, the fairness coefficient of fixing the contract coverage is relatively large, meaning that the variance of unit power generation income among the RES units is too high. Further analysis shows that fixing the annual CCR is a solution to the proposed model, but not the optimal solution. The decision-making cycle of optimal contract coverage should be determined according to the changes of generation output and load demand, while considering the policy formulation and the implementation of RES units.

4.3 Analysis of Medium and Long-term Results
Based on the “14th Five-Year Plan”, the “carbon peaking” goal in 2030, and the outline of the long-term goal in 2035 (The State Council Information Office of the People’s Republic of China, 2020; National Development and Reform Commission of China, 2021), the simulation of RES unit participation in the spot market from 2020 to 2035 is carried out, as shown in Figure 5.
As can be seen from Figure 5, during the transition of electricity market reform, due to the continuous development of renewable energy generation technology, government subsidies decrease, and the CCR of RES units and the proportion of CFD revenue also decrease. The proportion of spot market revenue and TGC market revenue increase, indicating that RES units gradually adapt to the spot market during the transition, and the subsidy for RES units changes smoothly from the financial subsidy to TGC subsidy. As the number of type II RES units increases year by year, the impact of contract coverage on the spot market and TGC market gradually decreases. In this example, to 2035, the revenue composition of RES units is relatively fixed, the change in contract coverage tends to be stable, and the curve only fluctuates with changes in load demand and generation output. When RES units have completed the transition phase to participate in the spot market, the determination of the contract coverage can be further adjusted according to the development and policies of renewable energy.

### 4.4 Sensitivity Analysis of Contract Coverage

Sensitivity analysis is carried out for the contract coverage against market boundary conditions including the assessment weights of the quota system and system load demand.

Firstly, the changes in CCR for RES units under different assessment weights of the quota system for fossil-fueled units are studied. Taking the simulation data in July 2025 as an example, the results are shown in Figure 6 and Figure 7. It can be found that with the increasing of the assessment weights, the CCR of RES units shows a downward trend. The reason is that when the assessment

| Unit type  | CFD income (yuan) | Spot market income (yuan) | TGC market income (yuan) | Fairness coefficient |
|------------|-------------------|---------------------------|--------------------------|---------------------|
|            | Proposed method   | Fixed contract coverage   | Proposed method          | Fixed contract coverage |
| Photovoltaics | 5148318 k         | 568219 k                | 89089 k                  | 57820 k             |
| Wind turbines | 672275 k         | 765772 k                | 126694 k                  | 76212 k             |
|             | 61488 k           | 58382 k                  | 184.7                    | 418.1               |
|             | 68649 k           | 55911 k                  | 340.3                    | 610.4               |

**FIGURE 5** | Changes in contract coverage and revenue share of RES generation units from 2020 to 2035.

**FIGURE 6** | Changes in CCR for photovoltaics under different assessment weights of the quota system.

**FIGURE 7** | Changes in CCR of wind turbines under different assessment weights of the quota system.
weights of the quota system for fossil-fueled units increase, their demand for TGC increases, which then leads to an increase in the clearing price of the TGC market. Under the same weighting factors, the reduction in government subsidy burdens caused by the decrease in contract coverage does not come at the expense of a sharp decline in the RES generator revenue, because the number of TGCs sold by type I RES units increases, making the RES generator revenue from TGC market increase. As fossil-fueled units have a greater demand for TGC during peak load periods, the contract coverage decreases the most during peak periods. Changing the market strategies of fossil-fueled units in the TGC market can effectively make up for the difference in the degree of decline of RES units in peak, flat and valley periods. Due to the output uncertainties of RES units and the approaching deadline for assessment, this difference cannot be eliminated. However, it is worth noting that due to the small output of photovoltaics during the valley period, the CCR does not change when adjusting the assessment weights.

Secondly, the changes in CCR for RES units under different load levels are examined. Taking the simulation data in July 2025 as an example, the results are shown in Figure 8 and Figure 9. The results show that the contract coverage shows a downward trend as the load demand increases. The reason is that the increase in load will lead to an increase in the spot market price. Under the same weighting factors, the reduction in government subsidy caused by the decrease in contract coverage does not come at the expense of a sharp drop in the revenue of RES units, because the spot market revenue of RES units increases. Similar to when adjusting the assessment weights, adjusting the load demand does not change the CCR for photovoltaics during valley periods.

Figures 6–9 shows that the changes in the assessment weights of the quota system and the system load level change the supply-demand relationship in the TGC market and the spot market, respectively. When the market demand exceeds the supply, the contract coverage for RES unit tends to decline.

5 CONCLUSION

How to realize the smooth transition from a fixed-price transaction mode to a full participation in the electricity spot market is an important issue for RES generation in the electricity market reform procedure of China. In this paper, a transition mechanism for RES units to participate in the electricity spot market considering the TGC market is presented. In this transition mechanism, the adjustment of CFD coverage for RES units is a key part of the design. By modeling the decision-making of various agents participating in the electricity spot market, a three-layer optimization model for solving the CCR is proposed, which provides a scientific basis for adjusting the CCR during the transition period. Extensive numerical experiments are conducted to manifest the feasibility and efficiency of proposed models and algorithms. The impact of the transition mechanism on the RES development is analyzed from the perspectives of short-term and medium-/long-term. The sensitivity analysis of CCR against the assessment weights and system load level shows that when the market demand exceeds supply, the CCR of RES units tends to decline.

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

KW, HD, and CX conceptualized the study. HD, JY, and CX performed the analysis. KW, JY, CX, and DQ
performed investigations. HD and ZZ acquired resources. HD, CX, and ZZ acquired funding. KW wrote the original draft. JY, CX, and FW reviewed and edited the manuscript. All authors agreed to be accountable for the content of the work.

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