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Analysis of CO₂ Emissions in the Whole Production Process of Coal-Fired Power Plant

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Abstract: The linear programming (LP) model has been used to identify a cost-effective strategy for reducing CO₂ emissions in power plants considering coal washing, pollutant removal, and carbon capture processes, thus CO₂ emissions in different production processes can be obtained. The direct emissions (combustion emissions and desulfurization emissions) and indirect emissions (pollutant removal, coal washing, and carbon capture) of CO₂ were all considered in the LP model. Three planning periods were set with different CO₂ emission control desirability to simulate CO₂ emissions of the different reduction requirements. The results can reflect the CO₂ emissions across the whole production process of a coal-fired power plant overall. The simulation results showed that for a coal-fired power plant containing two 1000 MW ultra super-critical sets, when the desirability was 0.9, the CO₂ total emissions were 2.15, 1.84, and 1.59 million tons for the three planning periods. The research results suggest that the methodology of LP combined with fuzzy desirability function is applicable to represent the whole production process of industry sectors such as coal-fired power plants. The government policy makers could predict CO₂ emissions by this method and use the results as a reference to conduct effective industrial and energy structure adjustment.



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1. Introduction

The report of the *AR6 Climate Change 2021: The Physical Science Basis* has recently been released by the Intergovernmental Panel on Climate Change (IPCC), which suggests that the global surface temperature has increased with a speed much faster than over the 50 years before 1970, which are mainly due to the greenhouse gases released by humans. Due to the experienced high economic growth over the past 40 years, China has become the country with the largest carbon emissions in the world [1–5]. Thus, China is facing tremendous domestic and international pressure to reduce carbon dioxide emissions [5]. In September 2020, China proposed at the United Nations General Assembly that it would strive to achieve a “carbon neutral” goal by 2060. This displayed China’s determination to reduce the carbon emissions to the global community. Coal power plants occupy a major position in the Chinese power sector, where 66% of its total electricity production originates from coal, according to International Energy Agency. The contributions of electricity and heat supply sectors to the total CO₂ emissions of China is significant at 4896 Mt while the coal-fired power generation sector contributed to 97% or 4747.9 Mt of carbon dioxide emissions [6]. Thus, it is vital that focus is applied to coal-fired power plants and the

resulting CO₂ emissions in order to mitigate global warming problems and provide plans for future adjustments in China's electrical power sectors [7].

The optimization model has been widely used to identify cheaper strategies for the reduction in greenhouse gas emissions in electrical power systems [8]. Most of the optimization models are based on bottom-up data, which considers the technology characteristics and parameters, and can be accordingly called bottom-up optimization models. The advantages of bottom-up models are that they usually include a great deal of detail about the model system, and are able to identify the very specific technological options and the associated investments and costs, so the model has been applied in many sectors [9–12]. Huang et al. [13] investigated energy savings and CO₂ abatement using an extended energy conservation supply curve (ECSC) and an extended marginal abatement cost curve to help the cement industry find cost-effective measures with the greatest potential in terms of energy savings and CO₂ abatement. Bhadbhade et al. [14] and Zuberi [15] investigated the current potential for energy efficiency (EE) improvement and CO₂ abatement in the Swiss metals and cement sectors, respectively, by means of bottom-up cost curves. These results could help overcome the techno-economic barriers in implementing best practices in the high value-added metals and cement sectors. Wang et al. [16,17] developed optimization models for multi-regional electricity generation and trading strategies in China based on power-plant constraints and renewable portfolio standard targets. However, a shortcoming of bottom-up models is that they single out the system of the study from the entire economic system and fail to take the indirect characteristics into consideration. In order to overcome these shortcomings, more and more researchers are using the linear programming (LP) method to solve optimization problems in the field of building management, particularly in problems related to environmental impact, energy consumption, or energy efficiency.

There have been a wide range in studies for bottom-up emission power plant models [18–23]. Coal-fired power plants are still the major source of electrical power in China, thus understanding CO₂ emissions in future scenarios is necessary for formulating carbon reduction goals and planning adjustments to the power industries. There are several model issues that need to be resolved in order to analyze CO₂ emissions across the whole production process for coal-fired plants: (1) How to set the model to make sure the stability and rationalization of the generating system at the same time reducing CO₂; (2) how to reflect the direct and indirect CO₂ emissions in the whole production process of a coal-fired power plant; and (3) how to reflect CO₂ emissions in different scenarios and limits.

Thus, in order to address these issues, linear programming was applied to the models in this study to solve the up-bottom issues. This would allow CO₂ emissions predictions and different desirability levels and also investigate the direct and indirect emissions in the whole production process of a coal-fired power plant. Scenarios analysis has been applied as a complementary method to obtain more decision-making plans under different desirability levels. The developed model was applied in a simulated power plant to provide effective support for the formulation of carbon emissions reduction measures. These simulated methods and their results allow for the prediction of CO₂ emissions potentials and provide a foundation for future adjustments in the energy sector to reduce CO₂ emissions overall.

2. Methodology

2.1. The Whole Coal-Fired Power Generation Process Analysis

Many steps in coal-fired power generation lead to either direct or indirect CO₂ emissions, whether that be coal washing, coal fired, or pollution removal processes. Thus, in order to analyze the generating cost and pollution emissions integrally, the programming model should be established based on the whole coal-fired power generation process. The whole coal-fired power generation process was defined (see Figure 1) in this study. As seen in Figure 1, the key steps in the whole coal-fired power generation process has been selected to construct the programming model, which includes coal washing, coal-fired

generation, pollution removal, and CO₂ capture processes. The CO₂ emissions and other air pollution were calculated under the objective function of generating cost.

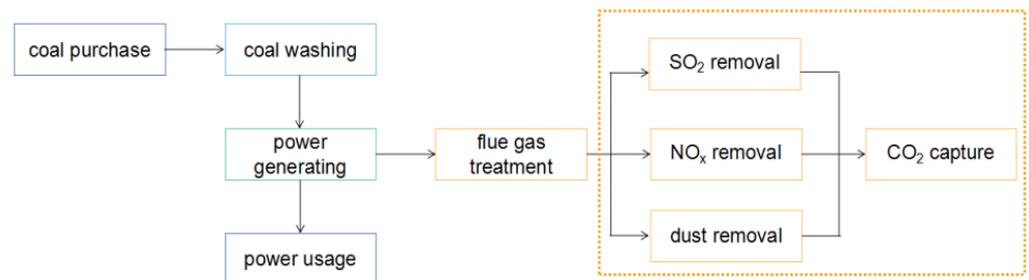


Figure 1. The process of coal-fired power generation.

2.2. Linear Programming

Consider a linear programming model as follows:

$$\min f = CX \quad (1)$$

Subject to:

$$AX \leq B, \quad (2)$$

$$X \geq 0. \quad (3)$$

where

$$C = (c_1, c_2, \hat{c}_n), \quad (4)$$

$$A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix} = (A_1, A_2, \hat{A}_m)^T,$$

$$A_i = (a_{i1}, a_{i2}, \hat{a}_{in}), \forall i, \quad (5)$$

$$B = (b_1, b_2, \hat{b}_n)^T, \quad (6)$$

$$X = (x_1, x_2, \hat{x}_n)^T \quad (7)$$

where f refers to an objective function; A is the parameter matrix; B and C are the parameter vectors; and X is the independent vector.

2.3. Fuzzy Linear Programming

Consider an interval fuzzy programming model as follows:

$$\max f = CX \quad (8)$$

subject to:

$$AX \lesseqgtr B \quad (9)$$

$$X \geq 0 \quad (10)$$

where $=$ and \lesseqgtr represent fuzzy equality and fuzzy inequality, respectively. Based on the principle of fuzzy flexible programming, a connection between the value of λ and membership function would be established. Specifically, the flexibility of constraint conditions and the fuzziness of the system objective would be denoted by the fuzzy number set. $[\lambda]$ as the degree of membership associated with the degree of satisfaction represents the "fuzzy constraint" or "fuzzy object". $\lambda = \min\{\mu_G, \mu_{C_1}, \mu_{C_2}, \hat{\mu}_{C_m}\}$ denotes the membership level. Therefore, the fuzzy linear programming model would be converted as follows:

$$\max \lambda$$

subject to:

$$\lambda \leq \mu_G, \mu_{C_i} (i = 1, 2, \dots, m) \quad (11)$$

$$0 \leq \lambda \leq 1 \quad (12)$$

$$\mu_G = \begin{cases} 1, & CX \geq f'' \\ \frac{CX - f'}{f'' - f'}, & f' \leq CX \leq f'' \\ 0, & CX \leq f' \end{cases} \quad (13)$$

$$\mu_{C_i} = \begin{cases} 1, & (AX)_i \leq b_i \\ 1 - \frac{(AX)_i - b_i}{P_i}, & b_i \leq (AX)_i \leq b_i + P_i \\ 0, & (AX)_i \geq b_i + P_i \end{cases}, \forall i \quad (14)$$

(Take Equations (13) and (14) into (11), and simplify the formulation, and then:

$$\max \lambda$$

subject to:

$$CX \geq \lambda f'' + (1 - \lambda) f' \quad (15)$$

$$(AX)_i \leq b_i + (1 - \lambda) P_i, \forall i \quad (16)$$

$$X \geq 0 \quad (17)$$

$$0 \leq \lambda \leq 1 \quad (18)$$

3. Case Study

3.1. Overview of the Simulation Case

A hypothetical coal-fired power plant was used in this study as the simulation case. The coal-fired power plant contained two 1000 MW ultra super-critical sets, and the coal consumption was about five million tons a year, and the generating capacity was about 10,000 GW·h. Due to the site limitations, the power plant could store the coal for 20 days, thus the power plant would purchase one or more kinds of the coal with the calorific values of 18.8, 20.9, 23.0, and 24.3 MJ/kg (named C1, C2, C3, and C4), respectively, for each month according to the coal price and generating cost. After being purchased, the coal was sent to the coal washing plant. The coal was washed by the coal jigging method until the mass fraction of sulfur in coal was lower than 1%, and then the coal was sent to the power plant to be blended. The coal-fired power plant was also equipped with wet flue gas desulfurization (WFGD), selective catalytic reduction (SCR), and bag-hose precipitation devices. Absorptive methods were also used to capture CO₂ in the power plant.

In order to study the air pollution and carbon dioxide removal effect under the objective function of the minimum generating cost in different scenarios, three planning periods were considered, each planning period lasting for three years. The parameters of generating cost, power consumption of coal washing, air pollution emission amounts, and power consumption of air pollution removal for each planning period are shown in Tables 1–4, respectively.

Table 1. Generating cost for the three planning periods.

	Coal Type	Planning Periods		
		t = 1	t = 2	t = 3
Generating cost (CNY/KW·h)	C1	0.135	0.135	0.135
	C2	0.125	0.125	0.125
	C3	0.115	0.115	0.115
	C4	0.11	0.11	0.11

Table 2. Power consumption of coal washing for the three planning periods.

	Coal Type	Planning Periods		
		t = 1	t = 2	t = 3
Power consumption of coal washing (KW·h/t)	C1	7	7	7
	C2	8	8	8
	C3	9	9	9
	C4	10	10	10

Table 3. Air pollution emission amounts for the three planning periods.

	Pollutants	Planning Periods		
		t = 1	t = 2	t = 3
Air pollution emission amounts (g/KW·h)	SO ₂	8.03	7.87	7.63
	NO _x	6.90	6.76	6.56
	Particulate matter (PM)	3.35	3.28	3.18

Table 4. Power consumption of air pollution removal for the three planning periods.

	Pollutants	Planning Periods		
		t = 1	t = 2	t = 3
Power consumption of air pollution removal (KW·h/t)	SO ₂	1500	1500	1500
	NO _x	3000	3000	3000
	Particulate matter (PM)	110	110	110

Desirability function was combined with the fuzzy linear programming of this study to analyze the solution of the programming model at the different desirability level, and the desirability degree was set as 0.9, 0.7, 0.5, 0.3, and 0.1 in each planning period.

3.2. Model Building

A linear programming model was established with the minimum generating cost as the objective function. In this study, the generating cost of the coal-fired power plant was considered to be composed of the resource cost, coal washing cost, power generating cost, air pollution treatment cost, and CO₂ capture cost. The model is shown below concretely.

3.2.1. Objective Function

Considering the coal consumption and pollutant treatment costs, the minimum generating cost of the power plant was set as the objective function in this study.

$$\begin{aligned} \min f = & \sum_{t=1}^T \sum_{i=1}^I Z_{it} \cdot CZ_{it} + \sum_{t=1}^T \sum_{i=1}^I Z_{it} \cdot CZS_{it} + \sum_{t=1}^T \sum_{i=1}^I Z_{it} \cdot W_{it} \cdot PV_{it} \\ & + \sum_{t=1}^T \sum_{i=1}^I \sum_{k=1}^K Z_{it} \cdot W_{it} \cdot PP_{tk} \cdot TP_{tk} \cdot CP_{tk} + \sum_{t=1}^T \sum_{i=1}^I Z_{it} \cdot FC_{it} \cdot TC_t \cdot CC_t \cdot M_t \cdot \frac{44}{12} \end{aligned} \quad (19)$$

Resource cost:

$$\sum_{t=1}^T \sum_{i=1}^I Z_{it} \cdot CZ_{it} \quad (20)$$

where Z_{it} is the coal mass of i kind in the period of t ; CZ_{it} is the coal price of i kind in the period of t , CNY/t; t represents the planning period; and $i = 1, 2, 3, 4$, represent C1, C2, C3, and C4, respectively.

Coal washing cost:

$$\sum_{t=1}^T \sum_{i=1}^I Z_{it} \cdot CZS_{it} \quad (21)$$

where CZS_{it} is the unit coal washing price of i kind in the period of t , CNY/t.

Power generating cost:

$$\sum_{t=1}^T \sum_{i=1}^I Z_{it} \cdot W_{it} \cdot PV_{it} \quad (22)$$

where W_{it} is the power generating of the i kind coal per unit mass in the period of t , KW·h/t. PV_{it} is the cost of power generating per unit, KW·h/t.

Air pollution removal cost:

$$\sum_{t=1}^T \sum_{i=1}^I \sum_{k=1}^K Z_{it} \cdot W_{it} \cdot PP_{tk} \cdot TP_{tk} \cdot CP_{tk} \quad (23)$$

where PP_{tk} is the emission amount of k kind air pollutant per unite power generating, t/KW·h; and $k = 1, 2, 3$, represent SO₂, NO_x, and particulate matter, respectively. TP_{tk} is the removal efficiency of k kind air pollution in the period of t , %. CP_{tk} is the air pollution removal cost per unit mass of k kind air pollution, CNY/t.

CO₂ capture cost:

$$\sum_{t=1}^T \sum_{i=1}^I Z_{it} \cdot FC_{it} \cdot TC_t \cdot CC_t \cdot M_t \cdot \frac{44}{12} \quad (24)$$

where FC_{it} is the combustion efficiency of i kind coal in the period of t ; TC_t is the CO₂ capture efficiency in the period of t ; CC_t is unit CO₂ capture cost in the period of CNY/t; and M_t is the usage rate of CO₂ capture technology in the period of t , %.

3.2.2. Constraint Conditions

(1) Electric quantity constraint:

$$\sum_{i=1}^I Z_{it} \cdot W_{it} \geq DE_t, \forall t \quad (25)$$

where DE_t is the power generating demand of the different planning periods, KW·h.

Resource constraint:

$$Z_{it} \leq MZ_{it}, \forall i, t \quad (26)$$

where MZ_{it} is the maximum supply quantity of i kind coal in the period of t , t.

Air pollutant emission constraint:

$$\sum_{i=1}^I Z_{it} \cdot W_{it} \cdot (1 - TP_{tk}) \cdot PP_{tk} \leq EP_{tk}, \forall t, k \quad (27)$$

where EP_{tk} is the emission amount of k kind air pollution in the period of t , t.

Fuzzy desirability function:

It is regulated that CO₂ emissions at each planning period decrease by 20% compared to the last planning period.

$$\frac{S_0 - \frac{F_t}{\sum_{i=1}^I (Z_{it} \cdot W_{it})}}{S_0 - S_0(1 - 20\%)^t} \geq \lambda \quad (28)$$

$$F_t = \sum_{i=1}^I Z_{it} \cdot FC_{it} \cdot (1 - TC_t) \cdot M_t \cdot \frac{44}{12} + \sum_{i=1}^I Z_{it} \cdot FC_{it} \cdot (1 - M_t) \cdot \frac{44}{12} + \sum_{i=1}^I Z_{it} \cdot CK_{it} \cdot S_t + \sum_{i=1}^I \sum_{k=1}^K Z_{it} \cdot W_{it} \cdot PP_{tk} \cdot TP_{tk} \cdot FP_{tk} \cdot S_t + \sum_{i=1}^I Z_{it} \cdot FC_{it} \cdot TC_t \cdot M_t \cdot EC_t \cdot S_t \cdot \frac{44}{12} + \sum_{i=1}^I \sum_{k=1}^K Z_{it} \cdot W_{it} \cdot PP_{t,k=1} \cdot TP_{t,k=1} \cdot \frac{44}{64} \quad (29)$$

where F_t is the total emission amounts of CO₂ in the period of t including direct emissions and indirect emissions, t. S_0 represents the CO₂ emissions of per unit power generating at the base year of the planning period, t/KW·h. CK_{it} is the coal washing power

consumption per unit i kind coal in t period, $\text{KW}\cdot\text{h}/\text{t}$. FP_{tk} is the air pollution removal power consumption per unit i kind coal in t period, $\text{KW}\cdot\text{h}/\text{t}$. EC_t is the power consumption of unit CO_2 capture, $\text{KW}\cdot\text{h}/\text{t}$. S_t is CO_2 emission amounts per unit electricity consumption, $\text{t}/\text{KW}\cdot\text{h}$. λ is the desirability, which were 0.1, 0.3, 0.5, 0.7, and 0.9, respectively, in each planning period.

4. Results Analysis and Discussion

4.1. System Cost Analysis

The model in this study used the minimum total generating cost as the objective function, after calculation, the model operation results of the total and unit power generating costs at the different desirability values are shown in Figure 2. It can be seen in Figure 2 that the total and unit power generating costs all increased with the increase in the desirability values. The total power generating cost of the three planning periods were 7.13, 7.40, 7.70, 7.97, and 8.28 billion CNY, and the unit power generating cost were 0.73, 0.76, 0.79, 0.82, and 0.85 CNY/ $\text{KW}\cdot\text{h}$ at the desirability values of 0.1, 0.3, 0.5, 0.7, and 0.9, respectively. It can be seen that the unit power generating cost had a linear relation with the desirability values, and the results suggest that when the desirability value increases by 0.2, the unit power generating cost will increase 0.03 CNY/ $\text{KW}\cdot\text{h}$.

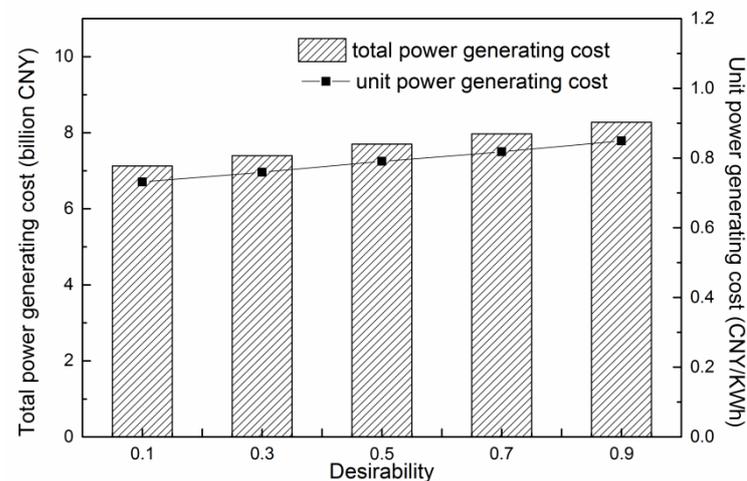


Figure 2. Power generating cost at different desirability.

The power generating production and coal consumption of the three planning period are shown in Table 5. It is expected that China's energy needs will continue to increase, thus in this study, the power generating production increased with the planning period time, and the total power generating production of the three planning periods were 97,436.56 $\text{GW}\cdot\text{h}$. Four kinds of coal (C1, C2, C3, and C4) were considered in this study with the calorific values of 18.8, 20.9, 23.0, and 24.3 MJ/kg, which were 64.3%, 71.4%, 78.6%, and 82.6% of the calorific values of the standard coal. As shown in Table 1, the generating cost of using C1, C2, C3, and C4 were 0.135, 0.125, 0.115, and 0.11, respectively. Thus, during the whole planning period, using C4 for generating was the most economical, and after the calculation, the coal consumption for generating was 382 g/ $\text{KW}\cdot\text{h}$ C4 in this case.

Table 5. Power generating production and coal consumption of the three planning periods.

		Power Generating Production ($\text{GW}\cdot\text{h}$)	Coal Consumption (Million Tons)			
			i = 1	i = 2	i = 3	i = 4
Planning period	t = 1	31,011.00	0	0	0	1.21
	t = 2	32,561.55	0	0	0	1.25
	t = 3	33,864.01	0	0	0	1.26
Total		97,436.56	0	0	0	3.72

4.2. Air Pollutant Emissions Analysis

The air pollutant emissions of the three planning periods are shown in Figure 3, and the emission intensities of the three air pollutants are shown in Table 6. It can be seen in Figure 3 that the air pollutant emissions increased with the power generating amounts, which were 17,431, 17,937, and 18,095 tons for SO₂; 13,327, 13,812, and 14,178 tons for NO_x; 1039, 1069, and 1078 tons for PM in the three planning periods, respectively. Though the emissions increased with the power generating amounts, the air pollutant emissions intensity of SO₂, NO_x, and PM all decreased with the increase in the planning period, which increases with the planning period due to the pollutants' removal level.

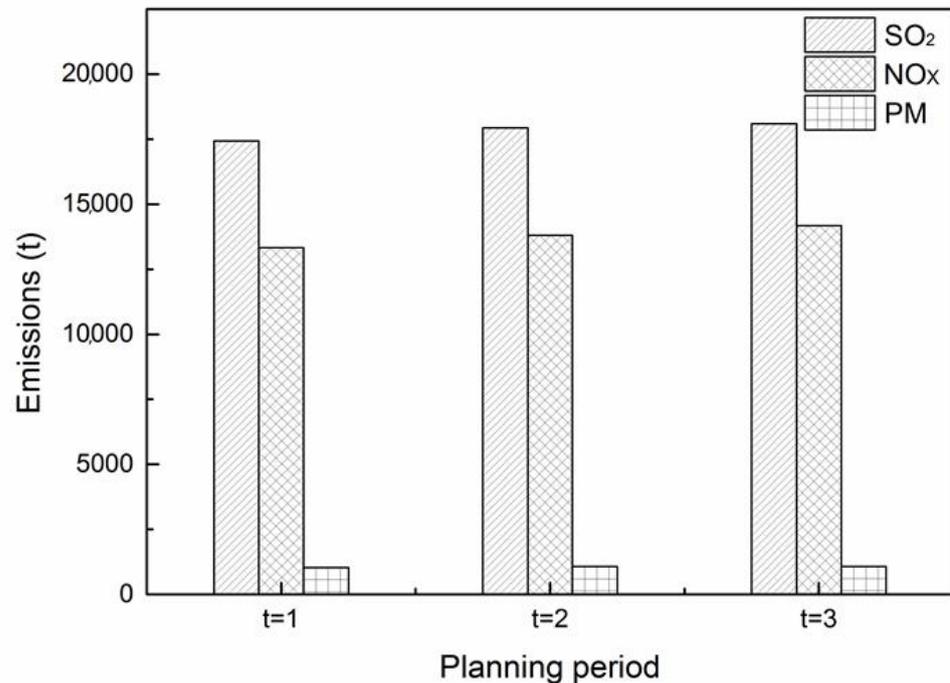


Figure 3. Air pollutant emissions at different planning periods.

Table 6. Air pollutant emissions intensity.

Planning Period (t)	Emission Intensity (g/kW·h)		
	SO ₂	NO _x	PM
1	0.56	0.43	0.034
2	0.55	0.42	0.033
3	0.53	0.42	0.032

4.3. Carbon Dioxide Emissions Analysis

4.3.1. The Utilization Rate of Carbon Capture Technology

Applying carbon capture technology is an effective method to help coal-fired power plants to reduce carbon dioxide emissions. The extravagant cost, however, provides major constraints in the actual application of carbon capture technologies for coal-fired plants. Thus, it is important to analyze the utilization rate of carbon capture technology at different planning periods with different desirability.

The utilization rate of carbon capture technology for different planning periods with different desirability is shown in Figure 4. It can be seen from the figure that with the increase in the desirability, the utilization rates of carbon capture technology all increased in the three planning periods. In order to obtain good carbon dioxide removal effects, the power plants must increase the investment in the application of carbon capture technology. It is noteworthy that when the desirability was 0.1, the utilization rate of carbon capture

technology in the first and second planning period was similar, however, that of the third planning period showed a small decrease. It can be suggested that the power generation technology improved in the third planning period, which caused the carbon emission of unit electric energy production to decrease. Lower desirability also means more carbon dioxide permissible discharge. Thus, the utilization rate of carbon capture technology in the third planning period decreased slightly and the rate of the utilization of carbon capture technology increased with the increase in the desirability value.

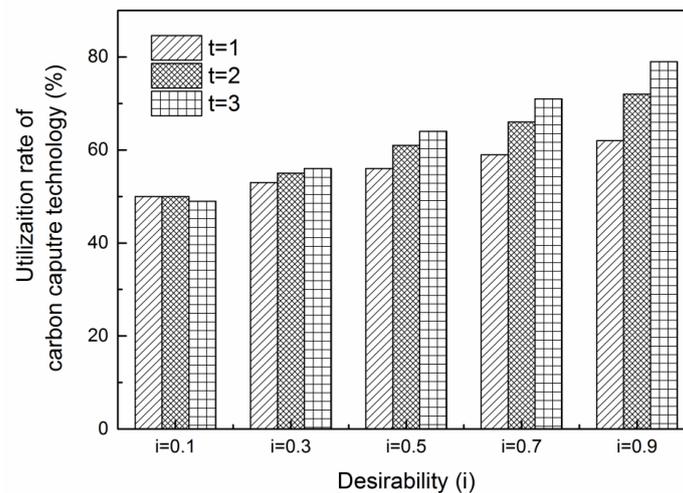


Figure 4. The utilization rate of carbon capture technology at different planning periods with different desirability.

4.3.2. Carbon Dioxide Emissions

In this model, the total emissions of carbon dioxide are composed by the direct and indirect emissions parts. Specifically, carbon dioxide emissions released by the combustion process and limestone-gypsum handling system resulted in direct emissions; while energy consumption of the carbon dioxide capture system, pollutant removal system, and coal washing process led to direct emissions. These simulated results are discussed below.

Analysis of the Total Carbon Dioxide Emissions

Total emissions of carbon dioxide at different planning periods with different desirability are shown in Figure 5A. As seen in this figure, total emissions of carbon dioxide decreased with the increase in the desirability value for each planning period. When the desirability was 0.1, total emissions of carbon dioxide increased with the increase in the planning period due to the moderate emission limits, which were 2.58, 2.65, and 2.70 million tons, respectively. However, when the desirability was above 0.3, total emissions of carbon dioxide all decreased with the increase in the desirability, and the emission reductions increased with the increase in the desirability value. When the desirability value was 0.9, which was the strictest emission limit of the simulation, the total emissions of carbon dioxide were 2.15, 1.88, and 1.59 million tons for the three planning periods, respectively.

Carbon dioxide emission intensities at different planning periods with different desirability are shown in Figure 5B. As seen in Figure 5B, the carbon dioxide emission intensity decreased with the increase in the desirability for each planning period, and the carbon dioxide emission intensities were 0.83, 0.80, 0.76, 0.73, and 0.69 kg/kW·h in the first planning period, and 0.80, 0.72, 0.63, 0.56, and 0.47 kg/kW·h in the third planning period. With the increase in application of carbon dioxide capture technology, carbon dioxide emissions intensities all decreased with the increase in the planning periods under each desirability, and the decrease rates of the third planning period to the first planning period were 4%, 9%, 17%, 23%, and 32% under the desirability values of 0.1, 0.3, 0.5, 0.7, and 0.9, respectively. When the desirability was 0.1, the total emissions of carbon dioxide increased, and the emission intensities decreased with the increase in the planning period, suggesting that

under the moderate emission limits, the increase level in the application of carbon dioxide capture technology was lower than the energy production increase level in order to pursue a higher economic benefit.

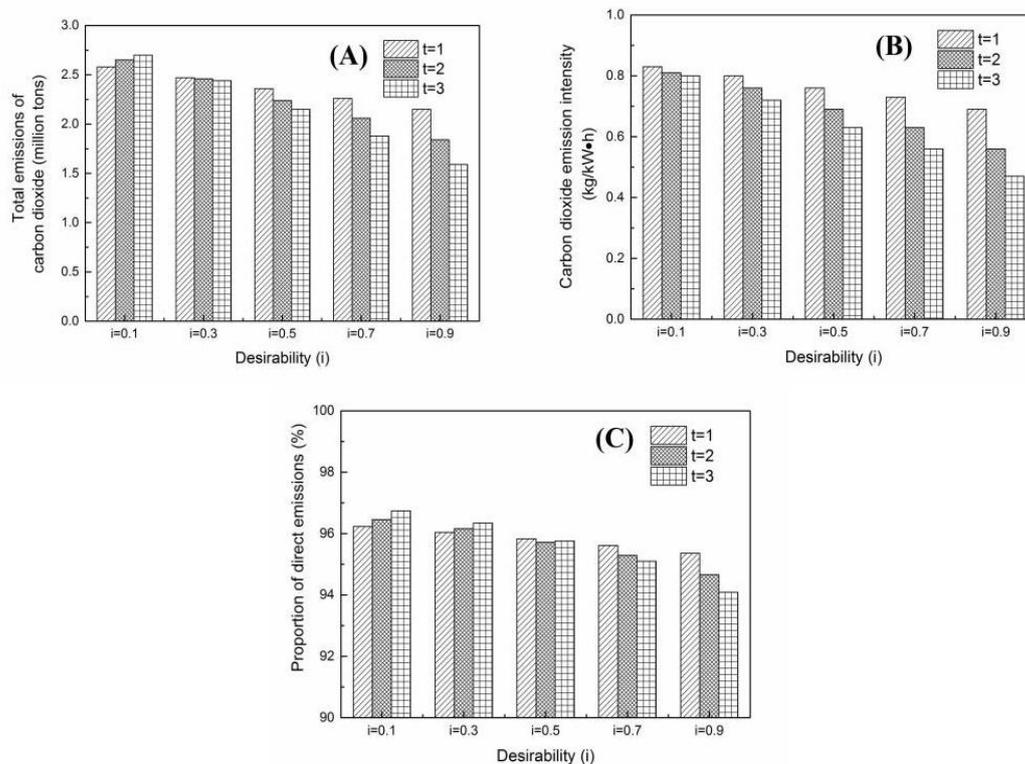


Figure 5. Total emissions (A), carbon dioxide emission intensity (B), and proportion of CO₂ direct emissions (C) of carbon dioxide at different planning periods with different desirability.

Figure 6 clearly shows that CO₂ direct emissions decrease, while the indirect emissions of CO₂ increase with the increase in the desirability for each planning period. As seen in Figure 6A, when the desirability was 0.1, CO₂ direct emissions increased with the increase in the planning period that had the same tendency with the total CO₂ emissions, which is because in the simulation, when the desirability is low, the upper limit of CO₂ emissions is high enough, so more energy production leads to the high direct CO₂ emissions for pursuing high economic benefit. With the increase in the desirability value, the upper limit of CO₂ emissions becomes lower gradually, and CO₂ reductions were enhanced with the increase in the planning period when the desirability was above 0.3. Indirect emissions of CO₂ decreased with the increase in the planning period under each desirability, and increased with the increase in desirability for each planning period due to the stricter requirements of CO₂ capture and pollution removal.

Proportions of CO₂ direct emissions at different planning periods with different desirability are shown in Figure 5C. As seen in Figure 5C, the proportions of CO₂ direct emissions increased with the increase in the planning period at the low desirability level (0.1 and 0.3), and decreased with an increase in the planning period at higher desirability levels (0.7 and 0.9). When the desirability was 0.5, the proportions of CO₂ direct emissions of the three planning periods were similar. The results suggest that with the enhancing of CO₂ reduction, CO₂ indirect emissions increase due to the increase in the power consumption of the technology improvement. Thus, CO₂ indirect emissions due to the powered devices should be considered when the power plant pursues low direct emissions of CO₂.

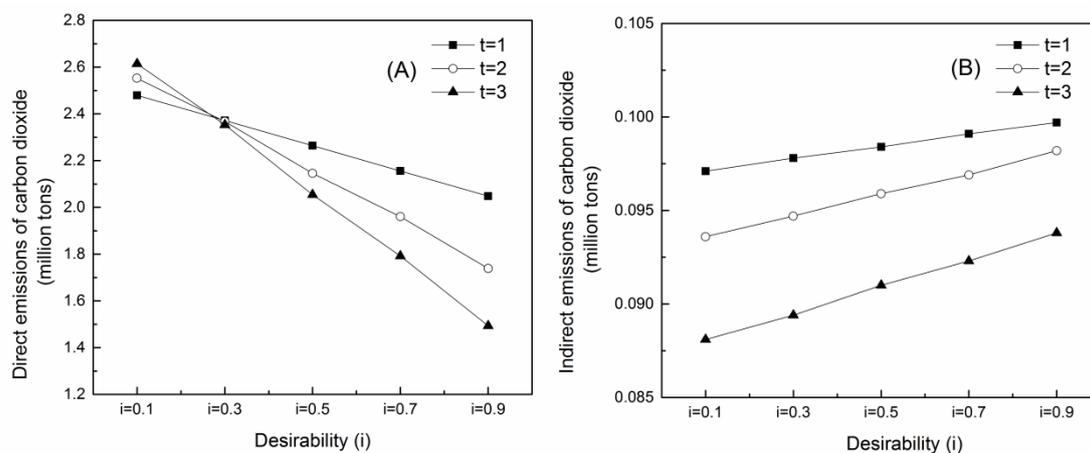


Figure 6. Direct (A) and indirect (B) emissions of carbon dioxide.

Analysis of the Direct and Indirect Carbon Dioxide Emissions

Proportions of direct CO₂ emissions by combustion in the total direct emissions are shown in Table 7. It can be suggested that the combustion is the major source (97%) of direct CO₂ emissions. When the desirability is 0.1, CO₂ combustion emissions are similar at different planning periods. When the desirability is above 0.3, CO₂ combustion emissions decrease with the increase in the planning period, suggesting that CO₂ emission controls produce effects. Proportions of indirect CO₂ emissions by CO₂ capture technology in total indirect emissions are shown in Table 8. As seen in Table 8, proportions of indirect CO₂ emissions by CO₂ capture technology increased with the increase in the planning periods and desirability. CO₂ emission control technology application can be reflected by the indirect CO₂ emissions by CO₂ capture technology, thus the greater the CO₂ capture technology applications, the higher indirect emission proportions of the technology and less direct CO₂ emissions by combustion.

Table 7. Proportions of direct CO₂ emissions by combustion in total direct emissions.

Desirability	t = 1	t = 2	t = 3
i = 0.1	98.07%	98.07%	98.10%
i = 0.3	97.99%	97.92%	97.89%
i = 0.5	97.89%	97.71%	97.59%
i = 0.7	97.78%	97.49%	97.23%
i = 0.9	97.67%	97.17%	96.68%

Table 8. Proportions of indirect CO₂ emissions by CO₂ capture technology in total indirect emissions.

Desirability	t = 1	t = 2	t = 3
i = 0.1	11.26%	10.97%	10.62%
i = 0.3	11.85%	11.93%	11.96%
i = 0.5	12.44%	13.06%	13.43%
i = 0.7	13.02%	13.98%	14.69%
i = 0.9	13.59%	15.06%	16.08%

5. Conclusions

The methodology of linear programming combined with fuzzy desirability function was applicable to reflect the CO₂ emissions across the whole production process of coal-fired power plants. CO₂ emissions were roundly predicted by the simulated model under five desirability values of the three planning periods, respectively. The main simulation results can be concluded below:

- (1) The simulation results showed that for the coal-fired power plant contained two 1000 MW ultra super-critical sets, the total power generating cost of the three planning periods were 7.13, 7.40, 7.70, 7.97, and 8.28 billion CNY, and the unit power generating costs were 0.73, 0.76, 0.79, 0.82, and 0.85 CNY/KW·h at the desirability values of 0.1, 0.3, 0.5, 0.7, and 0.9, respectively.
- (2) With the increase in the planning periods, the air pollutant emissions increased with the power generating amounts while the air pollutant emission intensity of SO₂, NO_x, and PM all decreased with the increase in the planning period.
- (3) When the desirability value was 0.1, the CO₂ emissions limit was moderate and the CO₂ emissions increased with the planning periods to obtain more economic benefits. When the desirability was above 0.3, the total emissions of CO₂ all decreased with an increase in the desirability values. The total emissions of carbon dioxide were 2.15, 1.88, and 1.59 million tons for the three planning periods, respectively, at the desirability value of 0.9.
- (4) CO₂ direct emissions accounted for above 96% of total CO₂ emissions. High desirability values meant high requirements for reducing CO₂ and air pollutant removal, thus with the increase in the desirability for each planning period, CO₂ direct emissions decreased, while the indirect emissions of CO₂ increased due to more applications of carbon capture and pollutant removal devices.
- (5) Combustion emissions and desulfurization emissions composed the direct emissions of CO₂. CO₂ combustion emissions were similar at different planning periods at the desirability of 0.1. When the desirability was above 0.3, CO₂ combustion emissions decreased with the increase in the planning period. Proportions of indirect CO₂ emissions by CO₂ capture technology increased with the increase in the planning periods and desirability, and more CO₂ capture technology application, higher indirect emission proportions of the technology, and less direct CO₂ emissions by combustion.

Thus, from the conclusions above, for the power plants to the whole energy production processes, reinforcing the investment of carbon capture technology is an effective way to reduce CO₂ emissions. However, with the increased requirements for emissions reduction, the indirect emissions of CO₂ will increase. Thus, on one hand, it is meaningful to find the balanced energy producing strategy to keep the direct and indirect CO₂ emissions at a low level on the premise of ensuring economic benefits; on the other hand, it is also important to find an energy conservation and pollution reduction approach to control CO₂ emissions for coal fired power plants in the future.

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