



## A Race Between Economic Growth and Carbon Emissions: How Will the CO<sub>2</sub> Emission Reach the Peak in Transportation Industry?

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Promoting carbon emissions of the transportation industry to reach the peak as soon as possible is an effective way to control carbon emissions in China. Combined with the logarithmic mean Divisia index (LMDI) model and the Tapio decoupling model, this paper tries to predict and analyze the time, path, and quality of carbon peaking in the transportation industry in China and its eastern, central, and western regions. The research shows the following: 1) Under the "benchmark scenario," the transportation industry in China and its three regions will not achieve the international commitment of achieving peak carbon emissions by 2030. 2) Under the "radical scenario," the peak time of China's transportation industry can be reached in 2027, which is 5 years earlier, 10% peak value lower than that in the "benchmark scenario." In that scenario, in transportation industry, the eastern region will reach the carbon peak in 2025 and the central and western regions will both reach the carbon peak in 2028. By then, the peak quality of carbon emissions will be better than that under the "benchmark scenario" in transportation industry. This paper can provide theoretical support for promoting carbon emission reductions and carbon peaking in China's transportation industry.

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## INTRODUCTION

To combat climate change, greenhouse gases need to be reduced mainly from  $CO_2$  emissions in China (Sun et al., 2021). As the world's largest economy emitting greenhouse gases (Chen et al., 2020; Zhao et al., 2021), China has set a two-stage goal of peaking the carbon emissions by 2030 and achieving carbon neutrality by 2060 to actively shoulder the international responsibility for carbon emission reductions (CPG, 2020a). In recent years, the transportation industry's carbon emissions accounted for an increasing proportion of the total national carbon emissions, from 9% in 2003 to 13% in 2019 (CESY, 2021). As the foundation of national economic and social development, the transportation industry's carbon emissions have maintained a rapid growth trend, with an average annual growth rate of as high as 8%. As of 2019, carbon emissions have still not reached the peak or exhibited a slowing trend (IEA, 2021). The growth trend of carbon emissions in the transportation industry may slow down the overall process of China's carbon peak. Therefore, research on carbon emission reductions and carbon peaking in the transportation industry is of great significance for China to achieve the 2030 goal.

Due to China's vast territory, there are great differences in economic development, population distribution, and geographical conditions in the eastern, central, and western regions, resulting in

TABLE 1	Comparison of	°CO <sub>2</sub> emis	sion prediction	i models in the	transportation industry.

Method	Advantages	Disadvantages	Documents Chen et al. (2020) Zhang and Su (2020) Yang et al. (2020)	
Econometric analysis (EA)	Simple technology; predictions can be based on any available data	Over-reliance on historical trends; impact factors have strong subjectivity		
Long-range energy alternatives planning (LEAP)	Flexible modeling structure; fewer data requirements; a simple and comprehensive accounting framework	Highly dependent on personal judgment; official data on technical parameters are lacking	Zhang et al. (2020) Liu J. et al. (2018) Fan et al. (2017)	
National energy technology model (NET)	Predicting the minimum cost of energy consumption and pollution emissions	Low execution efficiency; lack of technical parameters	Zhao et al. (2021) Li and Yu (2019)	
KAYA forecast	Identifying intrinsic driving factors; easy of analysis from the perspective of economics	Limited settings for intrinsic driving factors	Wang D. et al. (2019) Wang and He (2018)	

different energy consumption structures and carbon emission efficiency of the transportation industry in its eastern, central, and western regions. This represents the overall peak of the transportation industry does not mean that the eastern, central, and western regions can reach the peak at the same time. Therefore, to promote China's transportation industry to achieve the carbon peak "quickly and well" and follow the strategic deployment of places with conditions to reach the peak first, it is necessary to conduct an in-depth analysis of the peak time and quality of the eastern, central, and western regions based on predicting the carbon peak of China's transportation industry.

Based on the above realistic background, this paper intends to study the following issues: first, analyzing the relationship between transportation industry carbon emissions and economic growth and exploring the socio-economic factors of carbon emissions from the transportation industry in China and its eastern, central, and western regions; second, predicting the peak time and peak value of the transportation industry in China and its three regions; and, third, evaluating the carbon peak quality.

Reaching the carbon emission peak does not mean an improvement in carbon emission efficiency. If the carbon peak is achieved at a higher total carbon emission, it will not be conducive to the total control of carbon emissions and will directly affect the time and difficulty of achieving carbon neutrality. Based on the prediction of the peak time and the peak value of the transportation industry, this paper puts forward the judgment criteria of the carbon peak quality and development quality of the transportation industry. This method of judgment is not only scientific and reasonable but also convenient for policy-makers to understand and operate, which is the main innovation of this paper.

The contributions of this paper are as follows: First, different from only studying carbon peak values and time (Liu J. et al., 2018; Chen et al., 2020; Zhao et al., 2021), this paper focuses on peak quality and puts forward the judgment criteria of peak quality. Second, different from the study of individual countries and regions (Fan et al., 2017; Zhang et al., 2020; Zhang and Su, 2020), from a regional perspective, the differences of the peak time, the peak value, and the peak quality in China and its eastern, central, and western regions are compared, which is conducive to the Chinese government to set the carbon peak goal at the regional level, so as to help China realize its carbon peak commitment in 2030 on time.

## LITERATURE REVIEW

With the increasing prominence of climate change caused by excessive greenhouse gas emission, in recent years, the influencing factors of carbon emission and its future evolution trend have gradually attracted wide attention in academic circles. Experts and scholars used different models to study the peak time and the peak value of carbon emissions in China. This paper made a comparative analysis of existing prediction models, as shown in **Table 1**.

At present, many experts and scholars study the peaking of carbon emissions from the macro level and the meso level based on different models and scenarios. Regarding the macro-level carbon peak research, the focus of scholars' research is whether China can fulfill its promise of peaking before 2030. It is expected that China will reach the carbon peak in 2030, which has a greater sense of responsibility for independent contribution than the United States, Japan, and the EU countries (Pan et al., 2017). Wang and He (2018) compared with the United States, Japan and EU countries in predicting the carbon peak of China's transportation industry, they believe that Chinese peak path is much better than that of developed countries. Similarly, Fragkos and Kouvaritakis (2018) predicted that China will reach the peak in 2030, later than OECD developed economies and earlier than other developing countries such as India. However, the volume of CO<sub>2</sub> emissions in the selected economies will continue to increase in the long run if the economic productivity of the OECD panel continues to grow (Sun et al., 2020b), which means that it is difficult to achieve the carbon peak, and the energy structure and energy efficiency are particularly important for emission reduction. In China, some scholars think it will reach the carbon peak in 2030 (Liu and Xiao, 2018; Mi et al., 2017) by optimizing the energy structure, adjusting the industrial structure (Yu et al., 2018), and improving the energy efficiency (Xu et al., 2021; Ding et al., 2019). On the contrary, other scholars believe

that it is difficult to realize the adjustment of energy consumption structure in the process of economic development (Duan et al., 2018), such that China's commitment to the international community cannot be fulfilled (Cui et al., 2019; Wang Z. et al., 2019; Chen et al., 2020). In addition, due to the vast region in China, the national carbon peak target should be divided into regional and provincial levels in accordance with the principles of efficiency and equity (Li et al., 2020), and some scholars focus on the peak time and value of carbon emissions of the area (Cheng et al., 2021) and provinces (Fang et al., 2019; Weng et al., 2019).

For the research on CO<sub>2</sub> emission reduction potential at the middle level, domestic and foreign scholars mainly focus on high energy consumption industries such as power transportation. Some scholars believe that the carbon emissions of China's power industry (Tang et al., 2018; Zhao et al., 2020), construction industry (Li et al., 2021), and heavy chemical industry (Lu et al., 2020) may reach the carbon peak by 2030. However, the economic and social development will lead to more traffic demand and boost the growth of traffic energy consumption (Sun et al., 2020a; Yu et al., 2020). Different scholars consider that the transportation industry may peak carbon emissions before 2030 (Liu J. et al., 2018; Liu L. et al., 2018; Liu et al., 2021; Zhao et al., 2021) and 2050 (Gambhir et al., 2015; Pan et al., 2018; Wang and He, 2018; Zhou et al., 2019; Chen et al., 2020) and after 2050 (Yang et al., 2020; Zhang et al., 2020) in China. Due to differences in the research design of different scholars, such as databases, methods, basic assumptions, and policy objectives, there will be greater flexibility in the forecast peak and time of carbon emission peak in the transportation industry (Fang et al., 2019; Wang D. et al., 2019). In order to analyze the fundamental source of carbon emissions from the perspective of economics (Zhang et al., 2020; Wang et al., 2021), this paper intends to use the LMDI model based on the Kaya model to predict the carbon peak of transportation industry in China.

To sum up, the academia about whether China can achieve commitment on time has not yet reached agreement on carbon emissions to peak, and the industry level of carbon emissions and its peak forecasting are still inadequate; especially, the transportation industry in our country's carbon emissions still lacks systematic analysis and forecasting of peak values and the corresponding peak policy analysis and recommendations. As a major emitter, it is of global significance to achieve a high-quality carbon peak for China. Therefore, this paper attempts to put forward the standard to evaluate the quality of carbon peak and discuss the peak time and how to achieve a high-quality peak in transportation industry in China.

#### METHODS AND DATA SOURCES

## Carbon Emission Measurement: Inventory Method

To measure the carbon emissions of the transportation industry in eastern, central, and western regions of China, this paper adopts the inventory method introduced in the 2006 IPCC Guidelines for National Greenhouse Gas Inventories (IPCC, 2006). The specific calculations are as follows:

$$CO_2 = \sum_{i=8}^{8} E_i \times K_i \times CEF_i \times \frac{44}{12},$$
(1)

where  $E_i$  indicates the consumption of the *i* terminal energy in the transportation industry,  $K_i$  is the conversion factor of energy standard coal of the *i* energy, and  $CEF_i$  indicates the carbon emission coefficient of the *i* energy (**Supplementary Appendix SA**). It is found that the transportation industry mainly consumes coal, gasoline, kerosene, diesel, fuel oil, liquefied petroleum gas, natural gas, and liquefied natural gas. Thus, this paper mainly considers the CO<sub>2</sub> emissions brought by the eight energies.

### The Relationship Between Economic Growth and Carbon Emission: Tapio Decoupling

To study the relationship between carbon emissions in the transportation industry and economic development, this paper introduces a decoupling model. Decoupling models include the OECD decoupling index, Velma decoupling index, and Tapio decoupling index (Wu et al., 2019). The Tapio decoupling model further subdivides the decoupling state, and its results are more accurate than those of the previous two methods. Therefore, this study chooses the Tapio decoupling model for research. The calculations are as follows:

$$\varepsilon = \frac{\Delta CO_2}{CO_2} \div \frac{\Delta GDP}{GDP},\tag{2}$$

where  $\varepsilon$  is the decoupling index,  $\Delta CO_2$  is the change of carbon emissions in the transportation industry, and  $\Delta GDP$  is the change of gross domestic product (GDP). To avoid overinterpreting slight changes as significant changes, Tapio (2005) still considers the interval floating at an elastic value of 20% at 1 as a coupling. The decoupling results are divided into eight categories: when  $\Delta GDP$ >0, it is divided into strong decoupling ( $\varepsilon < 0$ ), weak decoupling ( $0 < \varepsilon < 0.8$ ), expansive coupling ( $0.8 < \varepsilon < 1.2$ ), and expansive negative decoupling ( $\varepsilon > 1.2$ ); when  $\Delta GDP < 0$ , it is divided into recessive decoupling ( $\varepsilon > 1.2$ ), recessive coupling ( $0.8 < \varepsilon < 1.2$ ), weak negative decoupling ( $0 < \varepsilon < 0.8$ ), and strong negative decoupling ( $\varepsilon < 0$ ).

In order to measure the peak quality of carbon emissions in the transportation industry, based on the Tapio decoupling model, this paper puts forward the quality evaluation criteria of carbon peak in the transportation industry: take the peak (historical highest value) to 1 year after the peak as the research period to measure the peak quality. If the decoupling index is smaller, then the peak quality is higher. Similarly, the development quality of the transportation sector is also measured by the annual decoupling index. The reasons for that approach are as follows: if the decline is faster within 1 year, it proves that the pollution prevention and control measures taken are more effective, and the relationship between carbon emissions and economic growth is weaker, which implies that the peak quality is higher.

## Influencing Factors of Carbon Emission's Growth: LMDI Decomposition Method

To measure the main socio-economic factors that affect the transportation industry's carbon emissions, this paper selects

the logarithmic mean Divisia index (LMDI) model, which has no residuals after decomposition, and can deal with zero-value problems to quantitatively evaluate the action mechanism and contribution efficiency of each factor. The formula is as follows:

$$CO_2 = \sum_{i=1}^{8} \frac{CO_{2i}}{E_i} \times \frac{E_i}{E} \times \frac{E}{GDP} \times \frac{GDP}{P} \times P, \qquad (3)$$

where *i* is the type of energy. The ratio of carbon dioxide emitted by *i* energy to energy consumption of type *i* ( $E_i$ ) represents the carbon emission coefficient (EF); the ratio of energy consumption of type *i* ( $E_i$ ) to total energy consumption of transportation industry represents the energy structure (ES); the ratio of total energy consumption to gross domestic product (GDP) in the transportation industry represents the energy intensity (EI); the ratio of GDP to total population (P) represents economic growth (EG). To simplify **Eq. 3**, we rewrite it as follows:

$$CO_2 = \sum_{i=1}^{8} EF_i \times ES_i \times EI \times EG \times P.$$
(4)

According to model calculations, the change in carbon emissions from the transportation industry from the base year (t) to the target year (t + n) can be decomposed into five factors, as follows:

$$\Delta CO_2 = \Delta EF + \Delta ES + \Delta EI + \Delta EG + \Delta P.$$
 (5)

Considering that if the technology does not change, under normal circumstances, the carbon emission coefficient of each energy will remain constant, and the effect of the emission coefficient on carbon emissions ( $\Delta EF$ ) will be zero. The decomposition results of the other four factors are as follows:

$$\Delta ES = \sum_{i} \frac{CO_{2,t+n} - CO_{2,t}}{lnCO_{2,t+n} - lnCO_{2,t}} \times ln\left(\frac{ES_{t+n}}{ES_t}\right), \tag{6}$$

$$\Delta EI = \sum_{i} \frac{CO_{2,t+n} - CO_{2,t}}{lnCO_{2,t+n} - lnCO_{2,t}} \times ln\left(\frac{EI_{t+n}}{EI_{t}}\right),$$
(7)

$$\Delta EG = \sum_{i} \frac{CO_{2,t+n} - CO_{2,t}}{lnCO_{2,t+n} - lnCO_{2,t}} \times ln\left(\frac{EG_{t+n}}{EG_t}\right), \quad (8)$$

$$\Delta P = \sum_{i} \frac{CO_{2,t+n} - CO_{2,t}}{lnCO_{2,t+n} - lnCO_{2,t}} \times ln\left(\frac{P_{t+n}}{P_t}\right),\tag{9}$$

where  $\Delta ES$  represents the energy structure effect;  $\Delta EI$  represents the energy efficient effect;  $\Delta EG$  represents the economic growth effect; and  $\Delta P$  represents the population effect.

#### **Data Sources**

This paper studies the transportation industry in China and the eastern, central, and western regions from 2003 to 2019. Energy consumption by each region is extracted from the China Energy Statistical Yearbook; other data come from China's National Bureau of Statistics (http://www.stats.gov.cn/tjsj/). To offset the effects of inflation, GDP in each region was adjusted to 2010 price levels. Due to the unavailability of data, this paper excludes Tibet, Taiwan, Hong Kong, and Macau. According to the classification

of the National Bureau of Statistics of China, this paper divides the remaining 30 provinces into the eastern, central, and western regions (**Supplementary Appendix SB**).

### THE DRIVING FACTORS FOR CARBON EMISSIONS' GROWTH IN THE TRANSPORTATION INDUSTRY Regional Carbon Emission Estimation: 2003–2019

In the national analysis, the carbon emission of China's transportation industry has not yet peaked from 2003 to 2019, and it is showing a rapid growth trend, with an average annual growth rate of 8% (**Figure 1**). In regional analysis, carbon emissions in the eastern, central, and western regions have not peaked, and the average annual growth rate in the central and western regions is about 8.5%, which is higher than that in the eastern region (7.4%).

#### The Decoupling Relationship Between Carbon Emission and Economic Growth

Through Tapio elastic analysis, this paper measures the decoupling indicators between carbon emissions and economic growth in the transportation industry from 2003 to 2018. And the historical decoupling status is studied in the medium term (5 years). This period is stable and can intuitively reflect the trend of change. The decoupling indicators in the range of 0-0.8 mean weak decoupling, and those lower than 0 refer to strong decoupling. The lower the value is, the better the decoupling state is. As shown in **Figure 2**, the decoupling relationship between carbon emissions and economic growth in the transportation industry showed an overall trend of optimization in recent decades.

At the national level, from 2003 to 2008, the decoupling index was 1.38, indicating that the growth rate of carbon emissions from the transportation industry was significantly higher than the growth rate of economic development, that is, the development of the transportation industry at the expense of the environment. From 2008 to 2013, the decoupling index decreased to 0.45, indicating a slowdown in the growth of  $CO_2$  emissions while GDP grew. The decoupling index from 2013 to 2018 was basically the same as that from 2008 to 2013, which shows that the quality of carbon emissions in the transportation industry has stagnated recently.

At the regional level, from 2003 to 2013, the decoupling index of carbon emissions and economic growth in the central region has been in a state of expansive coupling, indicating that the decline in carbon emissions in the central region is not obvious; the eastern and western regions have changed from an expansive negative decoupling state to a weak decoupling state. From 2013 to 2018, the three major regions all showed a weak decoupling state.

Although the decoupling status of China and its three regions is weak decoupling, it is still a long way from strong decoupling. This means that, with the economic growth, the  $CO_2$  of China's





transportation industry is still growing slowly, and the trend of emission inflection point is not obvious. It can be seen that exploring the emergence of emission inflection point has become an important topic.

# Socio-Economic Factors Affecting Carbon Emissions' Growth

To predict the trend of carbon emissions reasonably, this paper employs the LMDI model to analyze the main socio-economic factors that lead to the growth of carbon emissions.

From 2003 to 2019, the total  $CO_2$  emissions of China's transportation industry increased by 510 million tons (Mt). Among them, economic growth and population size have a

positive effect on carbon emissions, while the energy structure and energy intensity have a negative effect on carbon emissions (**Figure 3**). The energy intensity reduces carbon emissions by 100 Mt, and the contribution rate to the growth of  $CO_2$  emissions is -20%. The energy intensity of the transportation industry can be improved *via* developing and utilizing the energy-saving and emission reduction technologies, and the carbon emissions will be suppressed effectively.

The eastern region is the most developed region with highquality human capital. The energy-saving and emission reduction technologies in the eastern region are better than those in the central and western regions because of more investment, and the negative contribution rate of energy intensity to the growth of carbon emissions is -23%. The central region is under great



economic growth pressure and has undertaken some high-energy industries transferred from the eastern region. Therefore, the contribution rate of energy intensity effect is only -14%, while that of the western region is -19%.

The energy structure resulted in a 19.45 Mt reduction in transportation carbon emissions, with a contribution rate of -4%. The improvement of the energy structure is attributed to the increase in the proportion of clean energy due to the implementation of energy conservation and emission reduction policies. The contribution of energy structure in the eastern, central, and western regions was -4%, -2%, and -5%, respectively.

The economic growth effect and population growth effect are the main drivers of carbon emissions' growth. From 2003 to 2019,  $CO_2$  emissions increased by 574.79 Mt due to economic growth, and the contribution rate is 113%;  $CO_2$  emissions increased by 55.34 Mt due to population growth, and the contribution rate is 11%.

# CARBON PEAK: TIME, VALUE, AND QUALITY

## Scenario Design for the Development of Transportation Industry

From the LMDI model, optimizing the energy structure and reducing the energy intensity are the primary ways to reduce carbon emissions in the transportation industry. Therefore, according to **Eq. 4**, the predicted  $CO_2$  emission of the transportation sector can be described as

$$FCO_{2t} = EF_{t-1} \times (ES_{t-1} - S) \times (EI_{t-1} - I) \times (GDP_{t-1} + g) \times (P_{t-1} + p),$$
(10)

where  $FCO_{2t}$  represents the predicted value of  $CO_2$  in the transportation industry; S represents the average annual decline in the proportion of oil and gas in the transportation industry; I represents the average annual

**TABLE 2** | Benchmark scenario of the transportation industry's development in2030 and 2035.

Factors	2030	2035		
	750/	700/		
Energy structure Energy intensity		40% lower than that in 2015		
Per capita GDP Population	_	100% more than that in 2020 14.34 billion		
opulation				

Note: Energy structure represents the percentage of oil and gas in the transportation industry.

TABLE 3   Mil	d and r	adical	scenarios	of the	transportation	industry's
development.						

Scenario	Energy s	structure	Energy intensity	
	2030 (%)	2035 (%)	2035	
Scenario 1 (mild A)	70	65	40% lower than that in 2015	
Scenario 2 (mild B)	75	72	45% lower than that in 2015	
Scenario 3 (radical)	70	65	45% lower than that in 2015	

Note: See Supplementary Appendix SC for the background of scenario setting.

decline in the energy intensity of the transportation industry; g represents the average annual growth of GDP; and p represents the average annual increase of population.

Combined with the important speech of General Secretary Xi Jinping at the Climate Ambition Summit and China's policies or objectives in 2030 and 2035 (CPG, 2020a; CPG, 2020b; CF40, 2020), the benchmark scenario will be designed to predict the carbon peak time and peak value of the transportation industry. In **Table 2**, the benchmark targets for 2030 and 2035 are listed.

Based on the benchmark scenario, according to the changes in the two major factors of energy structure and energy intensity, three scenarios are set in **Table 3**. In scenario 1 (mild scenario A), it is assumed that the energy



structure is cleaner than that in the benchmark scenario in 2030 and 2035, i.e., the oil and gas proportion will be reduced to 70% in 2030 and 65% in 2035. In scenario 2 (mild scenario B), it is assumed that the energy intensity will be lower than that in the benchmark scenario by 2035, i.e., the energy intensity will decrease by 5%. In scenario 3 (radical scenario), it is assumed that, in 2030, the energy structure and energy intensity are improved at the same time, i.e., scenario 1 and scenario 2 are combined. It should be specifically noted that although GDP growth and population growth are the main factors affecting carbon emissions, we do not consider sacrificing economic interests and demographic dividends to achieve carbon emission reduction.

Therefore, the scenario changes of economic growth and population growth are not considered in the prediction of this paper. The economic growth rate and population growth rate of mild scenario and radical scenario are consistent with those in the benchmark scenario, and the energy structure and energy efficiency will change at a constant rate.

#### **Carbon Peak Forecast: Time and Value**

At the national level, under the benchmark scenario, the transportation industry will achieve the carbon peak in the year 2032 with 880.86 Mt carbon emissions (**Figure 4**). In mild scenario

Scenario	Points	China	Eastern	Central	Western
Benchmark	Decoupling index	-0.05	-0.03	-0.05	-0.03
	Peaking time (T)	(2032)	(2031)	(2032)	(2033)
Scenario 1 (mild A)	Decoupling index	-0.05	-0.02	-0.10	-0.05
× ,	Peaking time (T)	(2029)	(2027)	(2030)	(2030)
Scenario 2 (mild B)	Decoupling index	-0.03	-0.05	-0.10	-0.05
	Peaking time (T)	(2029)	(2028)	(2030)	(2030)
Scenario 3 (radical)	Decoupling index	-0.06	0.00	-0.05	-0.05
	Peaking time (T)	(2027)	(2025)	(2028)	(2028)

TABLE 4 | Comparison of the peak quality in four scenarios.

Note: (T, T + 1) is the calculated period of decoupling.

A, the peak time is 2029, 3 years earlier than that in the benchmark scenario, and the peak value is 821.95 Mt. In mild scenario B, the peak time is the same as that of mild scenario A, but the peak value increases by 14.53 Mt compared with that in mild scenario A. In the radical scenario, the carbon peak of the transportation industry will be advanced to the year 2027, reaching a peak of 792.31 Mt, which is 88.55 Mt less than that in the benchmark scenario.

At the regional level, in the benchmark scenario, the eastern region will reach the peak in the year 2031, with 420.49 Mt carbon emissions, and the central region and western region will peak in the years 2032 and 2033, respectively, with carbon emissions of 238.77 Mt and 222.28 Mt, respectively. Reasons for these results are that the eastern region pays more attention to the quality of economic development and considers environmental constraints at the same time, while the central and western regions focus on the speed of economic growth, and the rapid economic growth effect offsets the energy structure effect and energy intensity effect, resulting in the increase of carbon emissions year by year, and the peak time is relatively late.

In mild scenario A, the eastern region is expected to reach the peak in the year 2027, 2 years earlier than the country, indicating that the eastern region's emission peak will help drive China's transportation carbon emissions to peak. The central region and the western region will reach the peak simultaneously in the year 2031. In mild scenario B, the eastern region will peak in the year 2029, and the central region and the western region will reach the peak simultaneously in the year 2030. Comparing the results of mild scenarios A and B, it can be found that optimizing the energy structure is more conducive to the transportation industry in the eastern region to achieve the carbon peak and reducing the energy intensity is more conducive to the transportation industry in the central and western regions to achieve the carbon peak. In the radical scenario, the eastern region will reach the carbon peak at 384.22 Mt in 2025, while the central region and western region will reach the carbon peak at 212.61 Mt and 196.89 Mt in the year 2028.

To sum up, China's transportation industry cannot reach the carbon peak by 2030 under the current policy. With a small improvement in the energy structure or energy intensity, the peak time can be reached in the year 2029 under the mild scenario and in the year 2027 under the radical scenario. However, reaching the peak does not mean a high carbon peak quality. Under various policy scenarios, the carbon peak quality will be analyzed in combination with the Tapio decoupling index.

TABLE 5 | Decoupling indices in 2030.

China	Eastern	Central	Western
0.06	0.02	0.06	0.12
-0.11	-0.14	-0.10	-0.05
-0.11 -0.27	-0.14 -0.3	-0.10 -0.26	-0.05 -0.21
	0.06 -0.11 -0.11 -0.27	China         Eastern           0.06         0.02           -0.11         -0.14           -0.11         -0.14           -0.27         -0.3	China         Eastern         Central           0.06         0.02         0.06           -0.11         -0.14         -0.10           -0.11         -0.14         -0.10           -0.27         -0.3         -0.26

Note: (2030, 2031) is the calculated period of decoupling.

#### **Peak Quality Analysis**

First, the decoupling index between carbon emissions and economic growth is used to reflect the quality of carbon peak when reaching the peak (**Table 4**).

The smaller decoupling index means the weaker correlation between carbon emissions and economic growth and the better quality of carbon peak. From a regional perspective, mild scenario B is a better scenario. When the carbon emission reaches the peak, under mild scenario B, the decoupling index is -0.05 in the eastern region, the decoupling index is -0.1 in the central region, and the decoupling index is -0.05 in the western region, which is better than that in the other three scenarios. Unfortunately, the peak year in this scenario is later than that in the radical scenario, which is mainly influenced by eastern China. In eastern China, reaching the peak in advance will not obtain higher peak quality because of the rapid economic development and adjusting the energy structure and reducing the energy intensity cannot effectively offset the carbon emissions caused by the economy. Therefore, sometimes, reaching the peak in advance does not mean high quality.

From a national perspective, the radical scenario is a better scenario. When carbon reaches the peak, the decoupling index of the radical scenario is -0.06, which is better than that of the benchmark scenario (-0.05) and mild scenarios A (-0.05) and B (-0.03), indicating that the "two-pronged" approach of optimizing the energy structure and improving the energy efficiency can not only shorten the peaking time and reduce the peak value effectively but also achieve a better peaking quality.

Then, the decoupling index is used to reflect the quality of transportation industry development in 2030 (**Table 5**). In each policy scenario, the peak time is discrete. The decoupling index of different peak time lacks a uniform benchmark to compare with. China has a commitment to reach the peak in the year 2030 to the international community. By examining the relationship between

carbon emissions and economic growth in 2030, we can compare the development quality under different policy scenarios in a unified time dimension. In 2030, the benchmark scenario has the worst decoupling state, that is, weak decoupling; both mild and radical scenarios achieve strong decoupling. The decoupling index of radical scenario is the best, and mild scenarios are second only to the radical scenario. This means that the development quality of China's transportation industry will be relatively good in the mild scenario and radical scenario B and radical scenario will be excellent scenarios to promote carbon emissions to peak.

### CONCLUSIONS AND SUGGESTIONS

This paper analyzes the relationship between transportation industry carbon emissions and economic growth and explores the socio-economic factors of carbon emissions. On this basis, it predicts the peak time, peak value, and peak quality of carbon emissions in China and the eastern region, central region, and western region. The main conclusions are as follows.

From a regional perspective, mild scenario B may be the best scenario for carbon emission reduction. In the benchmark scenario, the transportation industry in eastern, central, and western China cannot achieve the goal of reaching the carbon peak by 2030. In mild scenario B, carbon emissions will reach the peak by 2028 in the eastern region, carbon emissions will reach the carbon peak by 2030 in the central and western regions, and the peak quality is better than that in the benchmark scenario. Therefore, mild scenario B is considered to be one of the best scenarios.

From a national perspective, the radical scenario is the best scenario for carbon emission reduction. Under the benchmark scenario, the transportation industry is expected to peak in 2032. Under the radical scenario, the peak time of China's transportation industry may be in 2027. And the peak quality of carbon emissions will be better than that under the benchmark scenario in transportation industry. Therefore, the radical scheme is considered to be one of the best choices to advance the peak time and ensure the peak quality.

To sum up, to promote the transportation industry to achieve the carbon peak "quickly and well," the carbon emission quality can be improved from the following two aspects:

First, the transportation energy structure should be improved and transportation energy should be transformed to green energy. For example, through cutting tax, adding subsidies, and publishing policies to increase support for new energy vehicles,

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the promotion of new energy vehicles is accelerated to replace traditional fuel vehicles.

Second, technical progress should be promoted in the transportation industry. Reduction in energy intensity is a key factor in peaking. The government should provide policy support to encourage enterprises to actively reduce automobile energy consumption by optimizing automobile engine technology and other transportation energy-saving technologies.

At the regional level, the eastern region should pay attention to the optimization of transportation energy structure due to its high traffic density. And the central region and western region should concentrate on improving the transportation energy efficiency due to having the potential for technological progress.

### DATA AVAILABILITY STATEMENT

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

## **AUTHOR CONTRIBUTIONS**

YW conceptualized the research idea and was involved in formal analysis and funding acquisition. YW and YZ performed the methodology. YZ ran the software. YW and YL were involved in data validation. YZ and JL curated the data. YL and JL visualized the data. YW, YZ, and YL wrote and edited the paper. All authors contributed to the article and approved the submitted version.

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## SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fenrg.2021.778757/full#supplementary-material

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