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Impact of Chinese fiscal decentralization on industrial green transformation: From the perspective of environmental fiscal policy

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To accomplish the goals of carbon mitigation, industrial green transformation is an inevitable requirement for achieving high-quality economic development. Based on the data of 30 provinces in China from 2007 to 2017, this paper uses projection pursuit model and entropy method to calculate the industrial green transformation and Chinese fiscal decentralization. It further employs static and dynamic spatial Durbin model to explore the impact of Chinese fiscal decentralization on industrial green transformation by the environmental fiscal policy. The study finds that: 1) China's industrial green transformation presents an unbalanced state with high in the east and low in the west, which has a positive spatial spillover among provinces. 2) Fiscal decentralization is significantly conducive to the industrial green transformation, while the spatial spillover of fiscal decentralization is negative. Moreover, fiscal decentralization affects the industrial green transformation by environmental fiscal policy, in of which environmental fiscal expenditure is the main path. 3) From the perspective of heterogeneity of fiscal decentralization, the impact of fiscal expenditure decentralization in promoting industrial green transformation is significantly greater than that of fiscal revenue decentralization. In terms of heterogeneity of regional location, fiscal decentralization in the eastern and central regions plays a role in accelerating industrial green transformation, while it has an obstacle impact in the western region. In terms of market-based heterogeneity, Fiscal decentralization in highmarket areas facilitates the green transformation of industry, while in low-market areas is the opposite.

KEYWORDS

Chinese fiscal decentralization, industrial green transformation, environmental fiscal policy, heterogeneity analysis, static and dynamic spatial Durbin model

1 Introduction

In recent years, global warming and climate change have gradually threatened the safety of human survival. Rising temperatures are leading to a rapid depletion of freshwater supplies (Bhardwaj A. et al., 2022). At the same time, the frequent occurrence of hot weather leads to shortages of food crops (Bhardwaj M. et al., 2022). The main reason for global warming is the increasing carbon dioxide emission. Referring to a report published by the IEA (2011), more than 60% of greenhouse gas levels are due to carbon dioxide (Dagar et al., 2022). To combat the effects of climate change, leaders of more than 170 countries signed the Paris Agreement, committing to limit global temperature in 2016. China is the country with the highest carbon emission and largest energy consumption in the world. Therefore, China expressed its determination to reduce carbon emissions at the 75th United Nations General Assembly, striving to achieve the goal of peaking carbon dioxide emissions by 2030 and carbon neutrality by 2060. This target is also in line with the expectations of the Sustainable Development Goals. China's industrial GDP dominates economic development with an average annual growth rate of 11.5%. However, at the same time, 40.1% of industrial GDP consumes 67.9% of China's energy and emits 83.1% of China's CO2 (Wang and Shao, 2019; Cai et al., 2019). The industrial carbon dioxide emissions of 30 provinces in China from 2007 to 2017 shown in Figure 1. In order to achieve the important goal of sustainable development, the industrial green transformation is an inevitable choice to achieve highquality economic development in China (Li et al., 2011).

The industry sector has the fastest growth, the greatest reforming effort and the highest opening scale in China for a long time. Accelerating industrialization has great contributions on China's sustainable economic growth. However, China's economic growth and industrialization development mainly rely on the resource-based growth route, which is characterized by "high input, high consumption, high pollution, low quality, low efficiency, low output, pollution first and treatment later." China extensive economic development model stems from the government behavior under the Chinese fiscal decentralization (Cai et al., 2008). China has established a system of fiscal decentralization since the implementation of the tax sharing reform in 1994. This system has greatly motivated local governments in China (Tian and Wang, 2018). Compared to other countries such as the United States, the European Union, India and Russia, China has implemented fiscal decentralization in a different political environment, where the typical vertical management system at all levels of government means that local government officials are primarily accountable to their superiors, resulting in the so-called Chinese style of fiscal decentralization (Yan, 2012; Fu, 2010). Fiscal decentralization gives local governments a priority on fiscal expenditure and resource allocation. To attract external investment and expand employment, local governments will compete to lower environmental standards, resulting in the decline of environmental quality (Zhang et al., 2011). However, with the intensity of environmental problems in recent years, the central government continuously polish the institute for evaluating and assessing the officials' performance. Due pollution control will directly affect the promotion of Chinese officials, local governments may increase the supervision on local enterprises' pollution discharge behavior and constantly encourage enterprises to reduce the pollution.

Industrial green transformation derives from green economy. The term "green economy" comes from the book "green



economy blueprint" published by Pearce, a British environmental economist, in 1989 (Pearce et al., 1989). Subsequently, some studies defined "green economy" as "a resource-saving and low-carbon economic development mode, which should be conducive to the protection and improvement of natural resources, promoting sustainable consumption and production" (Morgera and Savaresi, 2013). So far, there is no unified definition of green economy, and most of the relevant definitions emphasize "jointly improving economic and environmental benefits, achieving sustainable growth through economic behaviors that are beneficial to the environment or that are not antagonistic to the environment" (Sustainable Development Strategy Study Group Chinese Academy of Sciences, 2010). The structural changes and transformation of economies (Shahzad et al., 2022), the sustainable use of material resources (Xie et al., 2022), the development of renewable energy sources and the constraints of non-renewable resources (Islam et al., 2022), and the use of wastewater treatment (Bhardwaj A. et al., 2022) will all have a significant positive impact on sustainable growth. Research group of institute of industrial economics CASS (Research group of institute of industrial economics CASS, 2011) defined the industrial green transformation as taking the intensive resources utilization and environmental friendliness as the guidance, green innovation as the core, following on the new road of industrialization, realizing the green and sustainable development in the whole process of industrial production, and achieving a win-win of economic and environmental benefits. Wang and Zhang (2019) regarded green transformation as the transformation of development mode to sustainable development led by the ecological civilization construction. It relied on circular economy, green management, environmental resource conservation, friendliness, ecological balance and harmonious development between human, nature and society. The theory of fiscal decentralization arose from the classic article "the pure theory of local expenditures" published by Tiebout in the Journal of political economy in 1956 (Tiebout, 1956). This article extended the research from local government expenditure to fiscal decentralization theory deeply. Colm and Musgrave (1960) and Oates (1972) developed the relevant theories from the perspective of local expansion effect of fiscal policy and fiscal decentralization under the federal system. The industrial green transformation has the dual attributes of a market economy and an environmental public good. The multiple attributes make the impact of fiscal decentralization on industrial green transformation uncertain. Large numbers of studies have argued that fiscal decentralization has a negative impact on industrial green innovation with both economic and environmental attributes (You and Ouyang, 2020; Dong et al., 2022). It is argued that under the "GDP tournament" assessment approach, local governments focus more on pursuing high economic growth and prefer to invest in projects with quick results in the short term at the expense of environmental protection (Zhang, 2016; Wu, 2017; Hou et al., 2018a). However, after the Chinese government proposed and improved the "green GDP" performance assessment, local governments are not obsessed with chasing economic growth and may invest more in the area of environmental protection (Kuai et al., 2019). This may make a positive change in the impact of fiscal decentralization on industrial green transformation, and therefore, the impact of fiscal decentralization on industrial green transformation needs to be further explored.

First, Industrial green transformation has a rich connotation, and it is difficult to accurately measure the nature of industrial green transformation by a single indicator, so it needs to be measured by multidimensional indicators. Studies have been conducted on the measurement of industrial green transformation mainly using the green total factor productivity method (Hou et al., 2018b; Chen and Golley, 2014), the elastic decoupling method (Tapio, 2002) and the comprehensive indicator method (Chen et al., 2016; Deng and Yang, 2019). Comparatively speaking, the composite indicator method can measure the connotation of indicators more comprehensively. And will continue to evolve with the development, the connotation and requirements of industrial green transformation are also enriched, we should continue to enrich the corresponding indicators. Second, the existing studies currently adopt single indicator of expenditure or revenue (Davoodi and Zou, 1998; Lin and Liu, 2000; Jin and Zou, 2005). In fact, fiscal decentralization is a comprehensive system includes both fiscal expenditure decentralization and fiscal revenue decentralization (Kuai et al., 2019). We should measure fiscal decentralization from the multidimensional perspective. Third, most scholars analyze the industrial green transformation from the aspects of environmental regulation and technology innovation. Magat and Viscusi (1990) took the pulp and paper products industry in Quebec, Canada as the research object, and used the least square method to demonstrate that environmental regulation can promote enterprises to reduce emissions by 20%. Telle and Larsson (2007) used enterprise panel data to identify that there is a significant positive correlation between the environmental regulation and green total factor productivity. Peng and Li (2016) employed the dynamic panel model and found out the nonlinear impact of various environmental regulations on China's industrial green transformation. Yue et al. (2017) studied the effects of independent innovation, technology import and government support on the industrial green transformation. Shen et al. (2018) explored the influence of different environmental regulation modes on industrial environmental efficiency. Fahad et al. (2022) argue that environmental regulation and industrial policy are effective in promoting technology spillovers, which have a greater impact on industrial structure. But there is a lack of exploration of fiscal policy on the green transformation of industry. Environmental fiscal policy has the dual attributes of environmental policy and

fiscal policy. The way in which it operates under a system of fiscal decentralization merits further discussion.

Based on this, this paper aims to answer three questions: 1) Given Chinese-style fiscal decentralization system, how does fiscal decentralization affect the industrial green transformation 2) How is the moderating impact of environmental fiscal policy on industrial green transformation. 3) Whether there is heterogeneity impact in different environment fiscal policy and. The resolution of these questions can help provide reference for similar economies to formulate a reasonable fiscal decentralization and fiscal policy in the context of economic transformation, it has a great practical significance for achieving sustainable development. Therefore, based on the definition of industrial green transformation and fiscal decentralization, this paper proposes to adopt the projection tracing method and entropy method to measure the industrial green transformation and fiscal decentralization, respectively. Meanwhile, we use the static and dynamic spatial Durbin model to comprehensively analyze the influence mechanism and spatial effect of fiscal decentralization on industrial green transformation from the perspective of environmental fiscal policy. It also further explores the analysis of the heterogeneity of fiscal decentralization on industrial green transformation from multiple perspectives, providing a novel way of thinking for us to analyze the above issues.

Compared with the established literature, the main contributions of this paper are as follows: Firstly, scientific industrial green transformation evaluation index system and measurement method are essential to identify the weak links in industrial transformation. Based on the definition, a complete set of industrial green transformation index is constructed. The system is based on a green growth strategy framework. It covers seven aspects, including pollution emissions, pollution control, resource intensity, green innovation, structural optimization, production and operation, and sustainable development. The weights are determined by the projection tracing method and the evaluation results are more robust. This provides a valuable reference for the industrial green development index system. Secondly, Different from the single fiscal decentralization indicator in most studies, this paper calculates the fiscal decentralization index from the dual dimension of fiscal revenue and expenditure decentralization. We consider information on on-budget and off-budget revenues and expenditures and central transfers and eliminate the confounding of population and economic size factors. Thirdly, based on the perspective of environmental fiscal policy, the influence mechanism of fiscal decentralization on Industrial green transformation is explored. it is argued that environmental fiscal policy is an important way for fiscal decentralization to influence industrial green transformation. Fourth, based on heterogeneity, we further explore the differences in the impact of fiscal decentralization on the green transformation of industry from multiple perspectives, it provides insight into the contribution of different fiscal decentralization and different regions to transformation.

The paper is arranged as follows. Section 2 presents the research hypotheses. Section 3 explains the models, data and econometric methods used in this paper. Section 4 presents the results of the empirical study. Section 5 presents a discussion between our main results with previous studies. Section 6 concludes the study as well as gives some policy recommendations.

2 Research hypothesis

2.1 Impact mechanism of fiscal decentralization on industrial green transformation

As an important institutional arrangement, fiscal decentralization is the key to adjust the financial power between the central government and local governments and further divide the responsibility of fiscal expenditure. Compared with the central government, local governments have more information advantages and better understands the preferences and needs of residents in their jurisdiction (Tiebout, 1956). With the continuous strengthening of the green-performance assessment system, fiscal decentralization encourages local governments to focus on their information advantages and abolish policies serving for local development. Kuai et al. (2019) believed that fiscal decentralization is expected to give local governments greater financial resources and power (Cheng et al., 2020), so that local governments have more rights to deal with environmental problems. In terms of resource intensity, fiscal decentralization accelerates renewable energy consumption and reduces the use of non-renewable energy (Su et al., 2021; Zhang et al., 2022). Renewable energy is considered to be one of the most important ways to protect the environment. The decentralization has led to greater autonomy for local authorities. Local governments will increase subsidies for renewable energy in order to mitigate environmental problems, reducing the cost of sustainable energy use and thus contributing to an increase in the share of renewable energy consumption in energy consumption. In a word, fiscal decentralization gives full play to the local information dominance and reduce the cost of enterprises, which inspires enterprises to perform green innovation and realize the industrial green transformation. We made the following hypotheses:

Hypothesis 1: Fiscal decentralization benefits industrial green transformation.

2.2 Moderating mechanism of environmental fiscal policy

On basis of fiscal decentralization, local governments often control regional pollution by environmental regulations. Environmental fiscal policy is a finance policy tool to advocate ecological protection and environment governance, mainly including environmental fiscal expenditure and environmental tax revenue. Environmental fiscal expenditure has the dual attributes of social investment and environmental regulation. It has the induced effect of general fiscal expenditure, which leads to more social capital and then affects the regional industrial structure, playing a regulatory role in pollution reduction (Jiang, 2018). It is also an input-oriented environmental regulation means, which encourages enterprises to implement green innovation. Environmental tax revenue internalizes the environmental cost (Pigou, 1932). By using the means of price, it adjusts the production cost and product price flexibly. Meanwhile, it effectively motivates the enterprises' initiative to reach the optimal solution of pollution reduction, which promotes their green transformation (Bovenberg and Goulder, 1997).

With the improvement of fiscal decentralization, the division of fiscal expenditure responsibilities is clearer, which is conducive to bringing into full play to the advantages of fiscal decentralization and reaching the optimal allocation of financial resources. At the same time, make sure to continuously implement the green strategy, local governments increase financial investment in environmental expenditure, reduce industrial environmental protection cost, encourage local industrial enterprises to develop green technology, and push the green transformation of the overall industry. However, from the actual situation of China's environmental tax, the special environmental tax settings sector is lacking. Scattered and small sewage charges provide local governments with limited financial support to perform the function of environmental pollution control. Thus, the following hypothesis is made:

Hypothesis 2: Fiscal decentralization has a positive impact on industrial green transformation through environmental fiscal policy. Moreover, Under the Chinese fiscal decentralization system, local governments are more inclined to adopt environmental fiscal expenditure policy to speed the industrial green transformation.

2.3 Spatial mechanism of fiscal decentralization on industrial green transformation by environmental fiscal policy

To get the goal of short-term economic growth, there is differentiated strategy interaction in fiscal decentralization among regions. When the local fiscal decentralization increases, local industrial green transformation improves. Then, there is a spillover to the industrial green technology and industry of neighbor regions. Neighbor regions choose a substitution strategy. They do not increase the expenditures of environmental protection to obtain revenues. This type of "free riding" behavior can easily lead to bottom-by-bottom competition, which is detrimental to the industrial green transformation of neighbor regions. Moreover, the regional environmental fiscal policy is not only affected by local factors, but also by that of neighbor regions. Some scholars believe that increasing environmental fiscal expenditure in a region will often lead to imitation by neighbor regions (Allers and Elhorst, 2005). By contrary, the environmental taxes are relatively scattered. It is difficult to generate fiscal revenue. The neighbor regions' governments are more motivated to take "competition to the end" measures, such as tax incentives and reducing environmental thresholds, to make concessions for economic growth (He et al., 2016). Thus, the following hypothesis is made:

Hypothesis 3: The spatial spillover effect of fiscal decentralization on industrial green transformation is negative. Moreover, the spatial spillover of environmental fiscal expenditure and environmental tax revenues may be heterogeneous.

In summary, the theoretical model of hypotheses constructed in this paper is shown in Figure 2.

3 Research design

3.1 Data sources

This paper takes the data of 30 provinces in China (excluding Hong Kong, Macao, Taiwan and Tibet) from 2007 to 2017 as samples. The data of industrial green transformation and other model's control variables are mainly from the China Statistical Yearbook, China Environmental Statistical Yearbook, China Energy Statistical yearbook, China Industrial Economic Statistical Yearbook, China Science and Technology Statistical Yearbook, Industrial Enterprise Science and Technology Yearbook, China High-Tech Industry Statistical Yearbook China Labor Statistics Yearbook, EPS database, Wind database and Guotai'an database. The model's core explanatory variables of fiscal decentralization and environmental fiscal policy are mainly from China Tax Yearbook, China Fiscal Yearbook and forward-looking database.

3.2 Variable Selection and descriptive statistics

(1) The Explained Variable. This paper draws on the definition of industrial green transformation by the research group of institute of industrial economics CASS (Research group of institute of industrial economics CASS, 2011). Combined with the main indicators of the industrial green development plan (2016–2020) issued by the Ministry of industry and



information technology in 2016, and the "14th Five-year plan" of China's industrial development strategy, we build an industrial green transformation index from seven aspects: pollution emission, pollution treatment, resource intensification, green innovation, structural optimization, production efficiency and sustainable development. The industrial green transformation index system is shown in Table 1.

- (2) Core Explanatory Variables. Fiscal decentralization (FD) is divided into fiscal revenue decentralization (FDR) (He et al., 2016) and fiscal expenditure decentralization (FDE) (Kuai et al., 2019). Fiscal revenue decentralization refers to the distribution relationship between the central and local governments in fiscal revenue. The higher the fiscal revenue decentralization is, the greater the power of local governments to control fiscal revenue is. Fiscal expenditure decentralization refers to the distribution relationship between the central and local governments in fiscal expenditure. The higher fiscal expenditure decentralization means that local governments have greater autonomy in fiscal expenditure. In this paper, referring to Chu et al. (2018), the four indicators of fiscal revenue autonomy rate, fiscal revenue proportion, fiscal expenditure determining rate by itself and fiscal expenditure proportion are used to calculate the fiscal decentralization by using the entropy method. Fiscal decentralization index system is shown in Table 2.
- (3) Moderating Variables. Environmental fiscal policy (EFT) is a moderating variable, including environmental fiscal expenditure (EF) and environmental tax revenue (ET). Environmental fiscal expenditure refers to Zhu and Lu

(2017). It is measured by the proportion of environmental fiscal expenditure in GDP. Environmental tax revenue turns to Wang and Li (2018). It is measured by the proportion of resource tax, cultivated land occupation tax, consumption tax, vehicle purchase tax, urban maintenance and construction tax, vehicle and vessel tax, and urban land use tax in the sum of total taxes and sewage charges.

(4)Control Variables. Referring to Zhang et al. (2016) and Chen and Ma (2011), This paper selects the following indicators as the control variables (X): Population density (PD) is measured by the ratio of total population to administrative area at the end of the year. The higher the population density, the greater the demand for the social public environment. It forces local governments to improve environmental benefits. Natural conditions (NC) are measured by per capita forest area. Areas with good natural conditions have a good rate of forest greenery, which facilitates the adsorption of pollutants. Local governments may be inclined to reduce funding for environmental protection. Human capital (HC) is measured by the average number of education years. An increase in the tertiary educated working population will give the region a strong R&D capability and provide a growth engine for industrial transformation. Descriptive statistics of variables are shown in Table 3.

3.3 Method and model

3.3.1 Projection pursuit method

This paper uses projection pursuit method to determine the weight. The basic idea of projection pursuit method is to project

TABLE 1 Industrial green transformation index system.

Target Layer	Criterion Layer
Industrial Green Transformation (UP _{grade})	Pollution Emission
	Pollution Treatment
	Resource Intensification
	Green Innovation
	Structural Optimization
	Production Efficiency
	Sustainable Development

TABLE 2 Fiscal decentralization index system

Target Layer	Criterion Layer	Index Layer			
Fiscal decentralization	Fiscal Revenue Decentralization (FDR)	Fiscal Revenue Autonomy Rate (RA)			
(FD)		Fiscal Revenue Proportion (RR)			
	Fiscal Expenditure Decentralization (FDE)	Fiscal expenditure determining rate by itself (EA)			
		Fiscal Expenditure Proportion(ER)			

high-dimensional data onto low-dimensional subspace through some combination, use projection objective function to measure the possibility of projection configuration exposing a certain data structure, set constraints, find out the projection direction vector that maximizes the projection objective function, and then analyze the structural characteristics of high-dimensional data in low-dimensional subspace according to the projection direction vector. The specific modeling process is as follows:

Step 1: Standardize original value. Set the sample as $\{x^*(i, j)|i = 1, 2, 3, \dots, n, j = 1, 2, 3, \dots, p\}$. Where $x^*(i, j)$ is the original data of the evaluation indicator j in the sample i; n is

the number of samples and p is the number of evaluation indicators. Most of the evaluation indicators have dimensional differences, and the original value of each evaluation indicator has different variation ranges. It is lack of comparability, so that it must be standardized. The following method can be used to standardize the original data of evaluation indicators:

Positive indicators:

$$x(i, j) = \left[x^{*}(i, j) - x_{\min}(j)\right] / \left[x_{\max}(j) - x_{\min}(j)\right]$$
(1)

Negative indicators:

$$x(i, j) = \left[x_{\max}(j) - x^{*}(i, j) \right] / \left[x_{\max}(j) - x_{\min}(j) \right]$$
(2)

Eqs 1, 2, $x_{max}(j)$, $x_{min}(j)$ are the maximum and minimum values of original data of evaluation indicator j; x(i, j) is standardized value of evaluation indicator, $0 \le x(i, j) \le 1$.

Step 2: Construct the projection objective function. Set $a = \{a(1), a(2), \dots, a(p)\}$ as the projection direction vector, and the one-dimensional projection value of the sample in this direction is:

$$Z(i) = \sum_{j=1}^{p} a(j) x(i, j) \ (i = 1, 2, \dots, n)$$
(3)

When optimizing the one-dimensional projection value, the projection value Z(i) is required to reflect the dispersion characteristics of small concentration (the projection points are condensed locally as closely as possible to form point clusters) and large dispersion (the projection point clusters are dispersed as far as possible). For this purpose, the projection objective function is constructed with the following formula:

$$Q(a) = S_Z D_Z \tag{4}$$

Eq. 4, S_Z , D_Z , respectively is the standard deviation and local density of the projection value respectively, namely:

$$S_{Z} = \sqrt{\sum_{i=1}^{n} (Z(i) - E(z))^{2} / (n-1)}$$
(5)

Variable	Symbol	Obs	Mean	S.D.	Minimum	Maximum
Industrial Green Transformation (log)	LnUPgrade	330	1.066	0.277	0.177	1.498
Fiscal Decentralization (log)	LnFD	330	-3.408	0.113	-3.605	-3.080
Fiscal Revenue Decentralization (log)	LnFDR	330	-0.748	0.230	-1.181	-0.219
fiscal expenditure decentralization (log)	LnFDE	330	-0.204	0.070	-0.428	-0.068
Environmental fiscal expenditure (log)	LnEF	330	-4.943	0.588	-6.727	-3.308
Environmental tax revenue (log)	LnET	330	-6.138	0.976	-10.367	-3.707
Population Density	PD	330	0.058	0.069	0.004	0.383
Natural Condition	NC	330	0.190	0.202	0.001	0.996
Human capital	HC	330	1.259	0.201	0.768	1.686

$$D_{Z} = \sum_{i=1}^{n} \sum_{j=1}^{p} (R - r(i, j)) u(R - (ri, j))$$
(6)

Eqs 5, 6, E(z) is the average value of the sequence $\{Z(i)|i = 1, 2, 3, \dots, n\}$ and R is the window radius of the local density. r(i, j) represents the distance between samples, r(i, j) = |Z(i) - Z(j)|. The symbolic function u(R - r(i, j)) is a unit step function. If $R \ge r(i, j)$, the function value is 1; If R < r(i, j), the function value is 0.

Step 3: Optimize the projection objective function. After the sample set is determined, the change of projection direction vector a determines the change of projection objective functionQ(a). Different data structure features are represented by different projection direction vectors, and the projection direction vector that most likely reflects a certain structural feature of high-dimensional data is the best projection direction vector. By setting constraints, solve the maximized projection objective function and calculate the best projection direction vector.

$$\max Q(a) = S_Z D_Z \tag{7}$$

$$S.T.\sum_{j=1}^{p} a^{2}(j) = 1$$
(8)

This is a complex linear optimization problem with optimization variables $\{a_j = | j = 1, 2, 3, \dots, p\}$, which is difficult to deal with by conventional optimization methods. Therefore, this study uses the genetic algorithm to solve, which can realize the simplicity and effectiveness of high-dimensional global optimization.

Step 4: Rank. After obtaining the best projection direction vector, take the best projection direction vector as the weight, multiply it by the standardized value of the corresponding evaluation indicator, and add all the evaluation index to obtain the projection value. According to the projection value in the sample, the superior and inferior of all the samples can be calculated.

3.3.2 Spatial Durbin model

To explore the spatial effect of fiscal decentralization on industrial green transformation, this paper first constructs the following static spatial Durbin model:

$$LnUPgrade_{it} = \alpha + \rho WLnUPgrade_{it} + \varphi_1 LnFD_{it} + \tau_1 WLnFD_{it} + \theta X_{it} + \eta WX_{it} + \mu_i + \nu_t + \varepsilon_{it}$$
(9)

Eq. 9, *i* denotes province, *t* denotes time, LnUPgrade represents the industrial green transformation, LnFD represents the fiscal decentralization, W is the spatial weight matrix, X are other control variables, μ_i and ν_t represents the fixed effect of individual and time, ε_{it} is the residual disturbance term.

Due to the spatial spillover and time lagging of industrial green transformation, this paper refers to Luo and Wang (2017),

and adds the industrial green transformation of the previous period into the model to form a dynamic spatial Durbin model:

$$LnUPgrade_{it} = \alpha + \delta LnUPgrade_{it-1} + \rho WLnUPgrade_{it} + \varphi_1 LnFD_{it} + \tau_1 WLnFD_{it} + \theta X_{it} + \eta WX_{it} + \mu_i + \nu_t + \varepsilon_{it}$$
(10)

Eq. 10, $LnUPgrade_{it-1}$ refers to the industrial green transformation of province *i* in the period t-1, Other symbols have the same meaning as Eq. 9.

In order to test the moderating mechanism of environmental fiscal policy on the relationship between fiscal decentralization and industrial green transformation, the interaction and spatial lagged term of fiscal decentralization and environmental fiscal policy are added to Eq. 10. The specific model is as follows:

$$LnUPgrade_{it} = \alpha + \delta LnUPgrade_{it-1} + \rho WLnUPgrade_{it} + \varphi_1 LnFD_{it} + \varphi_2 LnFD_{it} \times LnEFT_{it} + \tau_1 WLnFD_{it}$$
(11)
+ $\tau_2 WLnFD_{it} \times LnEFT_{it} + \theta X_{it} + \eta WX_{it} + \mu_i + \nu_t + \varepsilon_{it}$

3.3.3 Spatial weight matrix

The setting of spatial weight matrix is the key to the analysis of dynamic spatial panel model. Based on Zhang and Zhu (2010), this paper sets up a nested spatial weight matrix including geographical and economic factors. The formula is as follows:

$$W_n(\phi) = (1 - \phi)W_d + \phi W_e \tag{12}$$

Eq. 12, W_d is the geographical distance weight matrix, W_e is the economic distance weight matrix, then W_n is between 0 and 1. ϕ is closing to 0, which indicates that the spatial weight is more related to the geographical distance factor. ϕ is closing to 1, which indicates that the spatial weight is more related to the economic distance factor.

4 Empirical analysis

4.1 Analysis of the industrial green transformation

We use projection pursuit method to calculate the industrial green transformation of 30 provinces in China from 2007 to 2017. The results are shown in Table 4, Figure 3. It is the development trend of China's industrial green transformation from 2007 to 2017 on basis of Supplementary Appendix Table S4. It shows that the industrial green transformation is rising over the whole period.

To analyze the change trend of industrial green transformation and its regional differences, this paper divides 30 provinces and province-level municipality into eastern, central and western region according to the national "Seventh Five Year Plan" economic and social development plan. This paper uses Dagum Gini coefficient decomposition to decompose the regional differences of China's industrial green

Province	2007	2017	Ranking	Province	2007	2017	Ranking
Beijing	4.001	4.258	3	Hubei	3.143	3.537	11
Tianjin	3.879	4.040	4	Hunan	2.581	3.543	15
Heibei	2.926	2.733	19	Guangdong	4.129	4.328	1
Shanxi	2.367	2.711	25	Guangxi	2.229	2.742	23
Neimengu	2.625	2.343	22	Hainan	2.927	2.313	21
Liaoning	2.971	2.734	18	Chonqing	2.931	3.961	8
Jilin	2.864	3.074	17	Sichuan	3.118	3.534	13
Heilongjiang	2.927	2.366	20	Guizhou	1.880	2.917	26
Shanghai	3.864	4.295	5	Yunnan	2.560	2.560	24
Jiangsu	4.090	4.080	2	Shaanxi	2.925	3.572	10
Zhejiang	3.721	3.830	6	Gansu	2.229	1.896	27
Anhui	2.926	3.541	14	Qinghai	1.895	1.321	29
Fujian	3.281	3.411	9	Ningxia	1.499	1.706	30
Jiangxi	2.906	3.064	16	Xingjiang	2.182	2.320	28
Shandong	3.720	3.543	7				
Heinan	3.155	3.527	12				

TABLE 4 Evaluation results of industrial green transformation index in 30 provinces (2007-2017).



transformation. The results are shown in Table 5. The results show that: 1) China's industrial green transformation gap increased from 0.1308 to 0.1435 during the period of 2007–2017. 2) The gap between the eastern, central and western regions of industrial green transformation expands year by year. 3) The gap between the eastern and the western regions is the largest. 4) The contribution rate of inter-regional gap to the industrial green transformation widening gap in China is always the highest. Therefore, narrowing the gap between regions is the key to solve the regional imbalance of industrial green development in China.

4.2 Spatial correlation test

4.2.1 Global Moran's I index

Table 6 reports the global Moran's I index from 2007 to 2017. The results show that except for FD, which failed to pass the

Year	Gini	Intraregional differences			Regional diffe	rences	Contribution rate (%)			
		East	Central	West	East-central	East-West	Central-West	Within region	Inter regional	Hyper variable density
2007	0.131	0.090	0.048	0.126	0.122	0.202	0.088	0.031	0.088	0.012
2008	0.144	0.104	0.051	0.138	0.137	0.218	0.097	0.035	0.092	0.018
2009	0.139	0.094	0.042	0.142	0.118	0.217	0.108	0.032	0.091	0.016
2010	0.132	0.084	0.038	0.150	0.106	0.213	0.107	0.030	0.090	0.012
2011	0.120	0.077	0.041	0.148	0.091	0.185	0.103	0.029	0.075	0.016
2012	0.144	0.098	0.050	0.167	0.115	0.219	0.118	0.035	0.089	0.020
2013	0.151	0.100	0.064	0.191	0.123	0.221	0.124	0.039	0.087	0.025
2014	0.155	0.107	0.070	0.199	0.125	0.220	0.130	0.041	0.083	0.031
2015	0.146	0.094	0.074	0.185	0.120	0.210	0.121	0.038	0.085	0.024
2016	0.158	0.111	0.091	0.204	0.134	0.211	0.127	0.044	0.077	0.036
2017	0.144	0.110	0.084	0.186	0.121	0.187	0.115	0.042	0.063	0.038

TABLE 5 Dagum Gini coefficient and its decomposition results of China's industrial green transformation.

significance test in part years, other variables are significantly greater than zero. It indicates that there is an obvious spatial dependence between variables. Therefore, it is necessary to employ the spatial model to further investigate the relationship between fiscal decentralization and industrial green transformation.

4.2.2 Local spatial autocorrelation

Figure 4 is the scatter chart of Moran's I index of industrial green transformation in 2007 and 2017. In the whole, the number of provinces in the low-low agglomeration mode of industrial green transformation has decreased, and the number of provinces in the high-high agglomeration has increased.

4.3 Model estimation results

Before estimating the parameters of the model, we need to determine the specific form of the spatial model at first. In this paper, LM Test, LR test and Hausman test are carried out in turn. The test results are shown in Table 7. LM statistics are significantly positive, indicating that there is significant spatial autocorrelation. So it is appropriate to use spatial model. Further, in order to determine the form of spatial panel model, Wald is also used to test whether the spatial Durbin model can be weakened into the spatial lagged model or spatial error model. It can be seen that models have passed LR test, meaning that the model should be set as spatial Durbin model. Then the Hausman test is used to determine whether it is a fixed effect or a

TABLE 6 Results of global Moran's I from 2007 to 2017.

Variables	LnUPgrade	LnFD	LnGT	LnTAX
2007	0.055***(2.650)	0.049**(2.470)	0.074***(3.194)	0.031*(1.940)
2008	0.053**(2.577)	0.041**(2.236)	0.069***(3.039)	0.032**(1.982)
2009	0.049**(2.439)	0.039**(2.179)	0.081***(3.391)	0.027*(1.826)
2010	0.054***(2.617)	0.044**(2.312)	0.106***(4.146)	0.035**(2.064)
2011	0.045**(2.352)	0.050**(2.516)	0.087***(3.574)	0.035**(2.060)
2012	0.046**(2.394)	0.030*(1.904)	0.075***(3.216)	0.028*(1.883)
2013	0.048**(2.450)	0.026*(1.781)	0.066***(2.993)	0.043**(2.317)
2014	0.044**(2.310)	0.025*(1.764)	0.073***(3.148)	0.034**(2.056)
2015	0.052**(2.550)	-0.004 (0.900)	0.059***(2.771)	0.046**(2.432)
2016	0.052**(2.564)	0.008 (1.257)	0.034**(2.035)	0.036**(2.127)
2017	0.051**(2.542)	0.027 (1.847)	0.030*(1.907)	0.050**(2.524)

Note: *, ** and *** donate significance at 10%, 5% and 1% levels, respectively. Z-values are given in parentheses.



TABLE 7 Estimation results of spatial Durbin model.

Variables	Benchmark regre	ssion	Mechanism test		
	(1)	(2)	(3)	(4)	
LnUPgrade _{t-1}		0.973***	0.843***	0.830***	
		(19.840)	(17.210)	(16.370)	
LnFD	0.266***	0.577***	1.081***	0.963***	
	(2.680)	(6.840)	(12.300)	(10.280)	
LnFD×LnEF			0.095***		
			(16.230)		
LnFD×LnET				0.047***	
				(11.700)	
PD	0.518*	0.275	0.813***	0.647**	
	(1.740)	(1.090)	(3.210)	(2.560)	
NC	-0.358	1.036***	0.351	1.171***	
	(-1.300)	(3.960)	(1.330)	(4.490)	
HC	0.245***	0.600***	0.793***	0.650***	
	(3.270)	(8.660)	(11.400)	(9.010)	
LM-Lag test	872.348***		354.441***	926.863***	
Robust LM-Lag test	286.131***		130.962***	352.707***	
LM-Error test	616.044***		294.046***	590.135***	
Robust LM-Error test	29.827***		70.567***	15.979***	
LR Test (SAR)	11.18**		16.180***	16.980**	
LR Test (SEM)	10.560***		13.330***	20.800**	
Hausman test	13.020**		38.900***	16.280**	
LogL	-410.806	-593.619	-1088.330	-410.806	
Р	0.460***	0.311***	0.992***	0.403***	
R2	0.016	0.166	0.099	0.176	
Ν	330	300	300	300	

Note: *, ** and *** donate significance at 10%, 5% and 1% levels, respectively. T values are given in parentheses.

Variables	Effect	(1)	(2)	(3)
LnFD	SR Direct effect	0.618***	1.212***	1.035***
	SR Indirect effect	-2.841***	-2.199***	-3.549***
	LR Direct effect	-20.919	8.912	9.573
	LR Indirect effect	10.243	-10.644	-16.131**
LnFD×LnEF	SR Direct effect		0.075***	
	SR Indirect effect		0.327***	
	LR Direct effect		0.559	
	LR Indirect effect		0.145	
LnFD×LnET	SR Direct effect			0.048***
	SR Indirect effect			-0.011
	LR Direct effect			0.350
	LR Indirect effect			-0.256

TABLE 8 Results of the direct and indirect effects.

Note: *, ** and *** donate significance at 10%, 5% and 1% levels, respectively.

random effect. It can be found that the Hausman test statistics significantly reject the null hypothesis, so the double fixed effect model is selected for estimation.

4.3.1 Impact of fiscal decentralization on industrial green transformation

In Table 7, models (1)–(2) are the regression results of fiscal decentralization affecting industrial green transformation using static fixed-effect spatial model and dynamic fixed-effect spatial model. The results show that the coefficient of fiscal decentralization is significantly positive, which reveals that fiscal decentralization is helpful to industrial green transformation. Due to the improvement of fiscal decentralization, local governments make better use of local information to optimize resource allocation and accelerate the upgrade of industrial green transformation. The spatial lagged coefficient in is significantly positive, which proves that the industrial green transformation has obvious spatial agglomeration characteristics. Therefore, the Hypothesis 1 is verified.

4.3.2 Moderate of environmental fiscal policy

To test the moderating mechanism of environmental fiscal policy, the interactive term of environmental fiscal expenditure and fiscal decentralization (LnFD×LnEF) is introduced into the column (3) of Table 7, it can be found that the interaction coefficient is significantly positive. It indicates that fiscal decentralization can indeed affect the industrial green transformation through environmental fiscal expenditure. In column (4) of Table 7, the interaction term of environmental tax revenue and fiscal decentralization (LnFD×LnET) is introduced, we can get that the coefficient of the interaction term is significantly positive. It means that fiscal decentralization can indeed affect the industrial green

transformation through environmental tax revenue. By comparing the coefficients of the interaction terms in column (3) and column (4), the coefficients of LnFD×LnEF is greater than that of LnFD×LnET. It shows that under the fiscal decentralization system, it is more likely for local governments to adopt environmental expenditure policies to push forward the industrial green transformation. Hypothesis 2 is verified.

4.3.3 Spatial effect decomposition

Referring to Anselin (2007), this paper decomposes the spatial effects into direct and indirect effects. The direct and indirect effect is further divided into long-term (LR) and short-term (SR) effect. Columns (1)–(3) of Table 8 give the decomposition results based on the columns (2)–(4) of Table 7.

The results show that, no matter direct or indirect, the shortterm effect of fiscal decentralization on industrial green transformation is more significant. Among that, the shortterm direct effect is significantly positive and short-term indirect effect is significantly negative. It indicates that fiscal decentralization enhances local industrial green transformation, but it has an adverse impact on its neighbors' industrial green transformation. The increase of fiscal decentralization in local intensifies the neighbors' fiscal competition. Then the neighbors compete to relax fiscal policies, which is detrimental to the local industrial green transformation. Column (2) of Table 8 shows that the short-term effect of the interaction coefficient between fiscal decentralization and environmental fiscal expenditure is significantly positive, the long-term effect is not significant, and the coefficient of the short-term indirect effect is greater than the short-term direct effect. The environmental fiscal expenditure policy has a model effect, which prompts neighbors to strengthen the environmental fiscal expenditure. Thus, it positively regulates the adverse impact of fiscal decentralization on the neighbors' industrial green transformation. Column (3) of Table 8 shows that the short-term direct effect of the interaction coefficient between fiscal decentralization and environmental tax revenue is significantly positive, while the short-term indirect effect is significantly negative. There is bottom-by-bottom competition in environmental tax among regions, which negatively regulates the impact of fiscal decentralization on industrial green transformation, but this effect is relatively small. It means that environmental tax policy restricts the industrial enterprises' emission by imposing a fee on the pollutants. The low charging standard with poor supervision on China's emission collecting institute inhibits its incentive role. The above analysis verifies Hypothesis 3.

4.4 Robustness test

The above comprehensive fiscal decentralization index calculated by entropy method is used to inspect the relationship between fiscal decentralization and industrial

Variables	Robustnes	Robustness test 1			Robustness test 2			Instrumental variable test		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	
LnUPgrade _{t-1}	1.115***	1.070***	1.261***	0.952***	0.935***	0.852***				
	(23.210)	(22.320)	(25.780)	(19.720)	(19.390)	(17.150)				
LnFD	0.319***	1.099***	0.250***	0.101**	0.085*	0.106*	7.539***	2.938***	2.652*	
	(6.500)	(9.810)	(3.750)	(2.060)	(1.720)	(1.920)	(3.800)	(8.070)	(1.690)	
LnFD×LnEF		0.205***			0.570**			0.103***		
		(9.060)			(2.380)			(9.370)		
LnFD×LnET			0.010			0.003			1.252**	
			(0.360)			(0.920)			(2.590)	
LM-Lag test	418.925***	232.899***	279.978***	167.820***	127.477***	148.187***				
Robust LM-Lag test	196.847***	89.126***	108.283***	57.235***	61.198***	63.810***				
LM-Error test	269.969***	209.394***	248.633***	110.610***	66.861***	84.486***				
Robust LM-Error test	47.891***	65.622***	76.938***	0.024	0.581	0.109				
Control Variables	YES	YES	YES	YES	YES	YES	YES	YES	YES	
LogL	-982.166	-1195.691	-1093.935	377.093	373.819	402.070				
ρ	0.488***	0.554***	0.272**	0.663***	0.660***	0.272**				
R^2	0.094	0.079	0.089	0.069	0.038	0.050	0.46	0.962	0.342	
Ν	300	300	300	300	300	300	330	330	330	

TABLE 9 The estimated results of robustness test.

Note: *, ** and *** donate significance at 10%, 5% and 1% levels, respectively. T values are given in parentheses.

green transformation. Table 9 in Columns (1)-(3) use fiscal autonomy (fiscal revenue/total fiscal expenditure in the provincial budget) as an alternative variable for robustness test, and columns (1)-(3) are estimated by dynamic spatial Durbin model. The estimated parameter symbols and significance of fiscal decentralization and its interaction term with environmental fiscal policy are basically consistent with the above estimated results. This shows that the impact of fiscal decentralization on industrial green transformation is reliable and stable.

Columns (4)–(6) in Table 9 report the results of the dynamic spatial regressions under the economic weight matrix setting, the common practice of constructing economic weight matrices based on the inverse of the absolute difference of an economic indicator that produces a spatial effect. This paper uses the inverse of the gap in the level of industrial green transformation between the two provinces: $W = 1/|UPgrade_i - UPgrade_j|$, otherwise it takes the value of 0. The results are generally consistent with the analysis above.

Furthermore, instrumental variable test can solve the endogenous problem to a certain extent. This paper draws on E Qu and Liu (2021) and uses the average value of the neighbors' fiscal decentralization in the same year as the instrumental variable (fd_iv) of local fiscal decentralization. Only the regression results of the second stage are reported in Table 9 in columns (7)–(9). It can be found that the coefficients are basically consistent with the previous conclusions.

4.5 Heterogeneity analysis

4.5.1 Decentralization heterogeneity

Due to the heterogeneity of fiscal decentralization, this paper divides fiscal decentralization into revenue decentralization and expenditure decentralization. Table 10 gives the estimated results. Column (1) of Table 10 shows the impact of revenue decentralization on industrial green transformation. The coefficient of revenue decentralization is significantly positive, which indicates that revenue decentralization is beneficial to the industrial green transformation. Column (2) of Table 10 introduces the interaction term between revenue decentralization and environmental fiscal expenditure (InFDR×LnEF), and the coefficient of the interaction term and spatial lagged term of revenue decentralization and environmental fiscal expenditure is significantly positive. Column (3) of Table10 introduces the interaction term of revenue decentralization and environmental tax revenue (InFDR×LnET), the coefficient of the interaction term is positive, which indicates that the environmental fiscal policy plays a positive moderating role between revenue decentralization and industrial green transformation. By comparing the interaction term coefficient of revenue decentralization and environmental fiscal expenditure, we can find that this coefficient is greater than that of revenue decentralization and environmental tax revenue, which is consistent with the previous analysis. That is, under the fiscal

Variables	(1)	(2)	(3)	(4)	(5)	(6)
LnUPgrade _{t-1}	0.863***	0.956***	0.663***	0.744***	0.915***	0.735***
	(17.700)	(19.860)	(13.270)	(15.320)	(18.880)	(14.320)
LnFDR	0.857***	1.612**	1.736***			
	(13.100)	(12.780)	(13.990)			
LnFDE				2.655***	7.380***	4.013***
				(8.730)	(14.310)	(8.600)
LnFDR×LnEF		0.252***				
		(10.680)				
LnFDR×LnET			0.222***			
			(11.900)			
LnFDE×LnEF					1.268***	
					(13.990)	
LnFDE×LnET						0.446***
						(6.140)
LM-Lag test	460.229***	227.821***	225.874***	630.324***	561.619***	566.376***
Robust LM-Lag test	236.013***	91.401***	109.114***	164.249***	224.761***	207.390***
LM-Error test	255.602***	192.158***	169.173***	533.363***	385.026***	426.819***
Robust LM-Error test	31.385***	55.739***	52.413***	67.287***	48.169***	67.833***
LR Test (SAR)	10.390**	14.170**	15.500***	14.420***	15.240***	27.590***
Test (SEM)	8.160*	11.220***	16.080***	13.250**	13.740**	24.410***
Control Variables	YES	YES	YES	YES	YES	YES
LogL	-627.046	-872.690	-586.466	-426.279	-1024.291	-377.222
Р	0.218***	0.517***	0.460***	0.907***	0.832***	0.591***
R2	0.182	0.288	0.161	0.098	0.102	0.120
Ν	300	300	300	300	300	300

TABLE 10 Estimated results of heterogeneity analysis of fiscal decentralization.

Note: *, ** and *** donate significance at 10%, 5% and 1% levels, respectively. T values are given in parentheses.

decentralization, local governments are more inclined to adopt environmental fiscal expenditure policies to boost the industrial green transformation.

Besides, Column (4) of Table 10 shows the impact of decentralization on industrial expenditure green transformation. The coefficient of expenditure decentralization is significantly positive, indicating that expenditure decentralization is useful to industrial green transformation. By comparing the coefficients of revenue decentralization and expenditure decentralization, we find that expenditure decentralization plays a greater role in advancing industrial green transformation. The reason is that local governments have greater control over the use of funds by the expenditure decentralization, which can better optimize the efficiency of resource allocation and promote the enhancement of industrial green transformation. Column (5) introduces the interaction item between expenditure decentralization and environmental fiscal expenditure (LnFDE×LnEF), the interaction coefficient is significantly positive. Column (6) introduces the between interaction expenditure decentralization environmental and tax revenue

(LnFDE×LnET), the interaction coefficient is significantly positive. It can also be found that the coefficient of LnFDE×LnEF is significantly greater than that of LnFDR×LnEF, which shows that we can give full play to environmental fiscal policy on expenditure decentralization to drive industrial green transformation.

4.5.2 Region heterogeneity

Due to geographical locations, natural resources and economic development difference, this paper divides the national sample into three regions: eastern, central and western region. It further examines the impact of environmental fiscal policies industrial on green transformation under fiscal decentralization in different regions. See Table 11 for the specific results. The coefficient of fiscal decentralization in the eastern and central regions is positive, and the coefficient of fiscal decentralization in the western regions is negative. This shows that the fiscal decentralization in the eastern and central regions can bolster the industrial green transformation, and the western fiscal decentralization is not promotive to the industrial green

transformation. This is because the economic development in the western region is backward and the fiscal decentralization is increased, which will increase the political and economic incentives of local governments. Therefore, it is easy to fall into an extensive economic development mode, which is not helpful to the industrial green transformation. The eastern and central regions have a higher economic development, and governments have more power to develop the industrial green transformation under the decentralization in local.

In addition, From the perspective of the interaction coefficient of fiscal decentralization and environmental fiscal expenditure, the coefficient of the eastern region is significantly positive, the coefficient of the central region is significantly positive, and the coefficient of the western region is not significant. Compared with the eastern region, the interaction coefficient between fiscal decentralization and environmental fiscal expenditure in the central region is smaller. Due the high-tech industry in the central region remains in its infancy, and the environmental fiscal expenditure needs to be digested and then further transformed into green productivity. From the perspective of the interaction coefficient between fiscal decentralization and environmental tax revenue, the interaction coefficient in the eastern and central regions is positive, and the interaction coefficient in the central region is more significant, while that in the western region is negative. Moreover, the coefficient of LnFD×LnET is greater than that of LnFD×LnEF. It shows that it is more efficient for the central region to select environmental tax policies to enhance industrial green transformation under the fiscal decentralization.

4.5.3 Marketization heterogeneity

The marketization degree have a positive impact on the green economy (Ren et al., 2022). Therefore, Referring to Fan et al. (2003), This paper classifies regions with market levels above or below the current median market level as high market level regions and low market level regions.Table 12 shows the results of the heterogeneity tests based on differences in marketability levels. Columns (1)–(3) show the results of dynamic spatial regressions for regions with high levels of marketization. The results show that fiscal decentralization in high marketization regions has a significant positive impact on the green transformation of industry, and the coefficient of the interaction term between fiscal decentralization and environmental fiscal expenditure is significantly positive. In contrast, the coefficient on the interaction term between fiscal

TABLE 11 Estimated results of regional heterogeneity analysis.

Variables	Eastern region			Central region			Western r	Western region		
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	
LnUPgrade _{t-1}	0.550***	0.530***	0.559***	0.600***	0.568***	0.418***	0.578***	0.586***	0.573***	
	(7.270)	(7.140)	(7.430)	(6.060)	(5.450)	(4.400)	(6.370)	(6.360)	(6.340)	
LnFD	0.116	0.063	0.068**	0.337**	0.369**	0.545***	-0.549*	-0.515*	-0.621*	
	(1.260)	(0.790)	(0.710)	(2.030)	(2.370)	(3.510)	(-1.820)	(-1.670)	(-1.960)	
LnFD×LnEF		0.012***			0.006**			0.011		
		(2.260)			(0.500)			(0.500)		
LnFD×LnET			0.004			0.032**			-0.005	
			(1.550)			(3.350)			(-0.200)	
LM-Lag test	170.631***	118.174***	155.096***	32.526***	19.254***	22.218***	119.041***	83.866***	82.489***	
Robust LM-Lag test	73.895***	34.260***	74.474***	3.780*	1.114	2.309	0.595	5.001**	0.469	
LM-Error test	97.805***	85.109***	81.180***	33.005***	25.079***	24.764***	133.962***	96.109***	112.496***	
Robust LM-Error test	1.069	1.195	0.558	4.258**	6.939***	4.855**	15.516***	17.244***	30.475***	
LR Test (SAR)	10.450**	7.010	10.830*	29.340***	31.560***	33.010***	55.170***	55.710***	37.320***	
Test (SEM)	9.670**	6.680	10.060*	14.830***	19.570***	23.370***	52.770***	53.330***	30.920***	
Control Variables	YES	YES	YES	YES	YES	YES	YES	YES	YES	
LogL	210.081	173.069	211.071	-31.739	-17.058	-220.431	-850.197	-1724.218	-931.681	
ρ	0.347**	0.328**	0.345**	0.292**	0.828***	0.965***	0.564*	0.556*	0.561*	
R2	0.047	0.095	0.086	0.228	0.233	0.310	0.532	0.488	0.509	
Ν	300	300	300	300	300	300	300	300	300	

Note: *, ** and *** donate significance at 10%, 5% and 1% levels, respectively. T values are given in parentheses.

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Variables	High marketization			Low marketization		
	(1)	(2)	(3)	(4)	(5)	(6)
LnUPgrade _{t-1}	0.830***	0.944***	0.761***	0.353***	0.406***	0.439***
	(13.950)	(16.000)	(12.460)	(3.600)	(4.230)	(4.580)
LnFD	0.409***	0.695***	0.097	-0.223**	-0.190*	-0.204*
	(6.540)	(8.990)	(1.340)	(-2.510)	(-1.950)	(-1.680)
LnFD×LnEF		0.057***			0.011	
		(7.530)			(1.120)	
LnFD×LnET			0.004			0.009
			(0.880)			(0.900)
LM-Lag test	303.928***	242.127***	221.871***	100.892***	97.106***	30.660***
Robust LM-Lag test	18.009***	14.022***	53.177***	0.004	1.330	2.878*
LM-Error test	386.867***	337.875***	202.616***	201.421***	172.507***	59.575***
Robust LM-Error test	100.948***	109.770***	33.922***	100.533***	76.731***	31.793***
LR Test (SAR)	18.670***	20.080***	13.480**	27.340***	33.180***	25.940***
Test (SEM)	11.450**	12.68**	14.09**	27.600***	32.550***	22.930***
Control Variables	YES	YES	YES	YES	YES	YES
LogL	-167.463	-355.541	267.422	-66.563	173.794	175.440
Р	0.602***	0.446***	0.679***	0.988***	0.119	0.046
R2	0.054	0.138	0.814	0.204	0.035	0.028
Ν	200	200	200	100	100	100

TABLE 12 Estimated results of heterogeneity analysis of marketization.

Note: *, ** and *** donate significance at 10%, 5% and 1% levels, respectively. T values are given in parentheses.

decentralization and environmental taxes is insignificant. Columns (4)-(5) show the results of dynamic spatial regressions for low-market areas. The results show that fiscal decentralization in low-market areas has a negative impact on the green transformation of industry, and the interaction term between fiscal decentralization and environmental fiscal policy is not significant. The reason for this is that more marketoriented regions are able to open up further to the outside world, and the influx of advanced technology and other technologies reduces the cost of transformation and upgrading for enterprises and regions with a low level of marketisation lack effective resource allocation, which is thus detrimental to industrial transformation.

5 Discussion

For China, industrial green transformation is necessary and important to achieve the win-win goals of sustainable economic growth and environmental protection. It is well known that fiscal decentralization has made an outstanding contribution to China's rapid economic development. However, it has also been blamed for causing serious environmental problems. In recent years, the central government has imposed increasingly stringent requirements on environmental management, which inevitably affects the local governments. In this context, it is essential to re-examine the impact of fiscal decentralization on the green economy. Based on this, this paper gives an insight into the impact of fiscal decentralization on the green transformation of industry.

First, we measure the industrial green transition through a projection tracing model and provide a detailed analysis of the regional differences and spatial distribution of the industrial green transformation based on the decomposition of the Dagum Gini coefficient and the spatial Moran index. This makes our findings more in-depth than previous studies on industrial green transition measurement (Chen et al., 2016; Deng and Yang, 2019). From 2007-2017, the industrial green transformation in China is on the rise year by year. China's industrial green transformation presents an unbalanced development state with high in the east and low in the west, and the internal gap among the three regions has been reduced. Industrial green transformation has significant "time inertia" and a positive spatial spillover on neighbor provinces.

Second, we think that fiscal decentralization is beneficial to the green transformation of industry. This result is inconsistent with many scholars (Song et al., 2018; Dong et al., 2022; Liu et al., 2022; You and Ouyang, 2020). This is because most previous studies have been derived from the premise that GDP growth is the only criterion for the performance of local officials in terms of promotion. However, with the introduction and improvement of China's green GDP assessment mechanism since 2007, it has been clear that the previous conclusions do not quite match reality. In terms of research methodology,

compared to You and Ouyang (Research group of institute of industrial economics CASS, 2011) who used a static spatial model, this study used both static and dynamic spatial models with the lagged effects of time, making the estimates more accurate and comprehensive. And this paper analyses the moderating role of environmental fiscal policy and finds that fiscal decentralization effectively promotes the role of environmental fiscal policy in the green transformation of industries, and that environmental fiscal expenditure has a greater role than environmental taxation. Unfortunately, this mechanism has rarely been discussed. It also differs from the study by You and Ouyang (2020), who argue that fiscal decentralization has a significant negative effect on the green innovation efficiency of provincial firms under different environmental regulatory instruments.

Third, we have tested the above results for heterogeneity. This part of the findings is a useful addition to existing research. The study by Wang and Li (2021) also distinguishes between revenue decentralization and expenditure decentralization, but their study ignores the information on internal and external budget revenues and expenditures in the selection of indicators, and they argue that revenue decentralization has a greater effect on curbing environmental pollution. Our research finds that expenditure decentralization can more fully exploit the positive regulatory mechanisms of environmental fiscal policy than revenue decentralization, effectively accelerating the green transformation of industry. And by testing for regional heterogeneity, we find that fiscal decentralization drives industrial green transformation in the east and central regions, while the west hinders it. The eastern region tends to work more through environmental fiscal spending, while the central region tends to work through environmental taxation and the western region has little role for environmental fiscal policy. The impact of regional heterogeneity is also discussed to provide policy considerations for regional fiscal strategy decisions. Based on a heterogeneity test of marketization, we find that fiscal decentralization in areas with high levels of marketization is conducive to a green transformation of industry, while the opposite is true for low marketization areas. This is similar to the findings of Ren et al. (2022).

6 Conclusion and policy implication

This paper adopts the balanced panel data of 30 provinces in China from 2007 to 2017. It calculates the industrial green transformation and fiscal decentralization by using projection pursuit model and entropy method respectively. It employs static and dynamic spatial Durbin model to conduct empirical research. The conclusions drawn as follows.

First, the industrial green transformation in China is on the rise year by year. China's industrial green transformation presents an unbalanced development state with high in the east and low in the west, and the internal gap among the three regions has been reduced. Industrial green transformation has significant "time inertia" and a positive spatial spillover on neighbor provinces. Besides, fiscal decentralization has a significant advancing effect on the local industrial green transformation, while it has a restraining effect on the neighbors' industrial green transformation. The short-term effect is more significant than the long-term effect. That means we should deepen the reform of fiscal decentralization and implement differentiated fiscal decentralization strategies combined with local conditions.

Second, fiscal decentralization effectively promotes the industrial green transformation by the environmental fiscal policy, and the role of environmental fiscal expenditure is greater than that of environmental tax revenue. That indicates we should continue to increase the scale of environmental protection expenditure, enhance the utilization efficiency of waste generated in production and life, and add the comprehensive utilization value of waste. Meanwhile, we should improve China's existing environmental tax institute and realize the integration of sustainable development and tax reform.

Third, compared with revenue decentralization, expenditure decentralization can more fully allocate resources, give full play to the positive moderating mechanism of environmental fiscal policy, and effectively speed industrial green transformation. The fiscal decentralization push for the industrial green transformation in the eastern and central regions, while that in the western region hinders the industrial green transformation. Under the fiscal decentralization, the environmental fiscal expenditure in the eastern region significantly boosts the industrial green transformation than the environmental tax revenue, and the environmental tax revenue in the central region better upholds the industrial green transformation, while the role of the environmental fiscal policy in the western region is not significant. Based on this, For the eastern regions, we should gradually expand fiscal decentralization and exploit economic, technological, talent and information advantages, which accelerate the speed of industrial green transformation. For the central region where heavypollution industry is concentrated, we should strengthen the government's financial intervention, set the bottom line towards the ecological environment, and form effective constraints and incentives for industrial emission reduction. For the western region with relatively backward economy, we should appropriately delegate power, encourage local governments to develop the economy, and make corresponding environmental standards to supervise industrial pollution emissions. Fiscal decentralization in high-market areas facilitates the green transformation of industry, while in low-market areas is the opposite. Therefore, local governments should further deepen the reform of market mechanisms, stimulate market vitality and provide a good competitive environment for the green transformation of industry.

In addition, the research of this paper still has some work leaving behind. In the above analysis, it confirms the heterogeneity in the impact of revenue decentralization and expenditure decentralization on industrial green transformation. There may be vertical fiscal imbalance, so we will consider the impact of vertical fiscal imbalance in future research.

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

YQ designed the computational framework and wrote the manuscript. XZ designed the model and analyzed the data. MX conducted empirical research. All authors discussed the results and contributed to the final manuscript. All authors have read and agreed to the published version of the manuscript.

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construction universities in Zhejiang province (Zhejiang Gongshang University- Statistics).

Conflict of interest

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

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fenvs.2022. 1006274/full#supplementary-material

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