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A network analysis of carbon emission flows among marine industries in China

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In recent years, environmental problems have become an important bottleneck restricting the sustainable development of Marine economy in China. CO_2 is one of the main greenhouse gases which contributes to marine environmental problems. As CO_2 emissions can transfer among industries, identifying the industries that release CO_2 most and clarifying the carbon emission flows among marine industries are helpful for decision-makers to curb CO_2 emissions of marine industries. This paper applies the network method to measure carbon emission flows. First, carbon transfer coefficient is calculated. Second, carbon transfer network of marine industry is constructed based on carbon transfer coefficient. Then the structure of marine industry carbon transfer network is analyzed. Finally, the method proposed in this paper is applied to the case of China and some suggestions for carbon reduction is put forward.

KEYWORDS

marine industry, carbon dioxide emissions, complex network, input-output table, weaver-thomas combination index, China

1 Introduction

As a major developing country with a coastline of 180 million kilometers, China attaches great importance to marine development and protection. Marine economy is a significant driving force for China's economic development. In 2021, the GDP of China's marine economy exceeded 9 trillion yuan, accounting for about 8% of the national GDP. However, environmental problems have become an important bottleneck restricting the sustainable development of marine economy. CO_2 is one of the main greenhouse gases which contributes to marine environmental problems. As CO_2 emissions can transfer among industries, identifying the industries that release CO_2 most and clarifying the carbon emission flows among marine industries are helpful for decision-makers to curb CO_2 emissions of marine industries.

Regarding the carbon emissions of the marine industry, from the current literature, some scholars have carried out research on some sub-sectors of the marine industry. Regarding the study of carbon emissions from marine fisheries, Parker et al. pointed out that the emissions of marine fisheries are often roughly summarized or directly excluded in the assessment of greenhouse gas emissions, so they quantified the fuel consumption and greenhouse gas emissions of global fishing vessels from 1990 to 2011, and found an increasing trend in total emissions (Parker et al., 2018); Greer et al. calculated the fuel consumption intensity, carbon dioxide emission intensity and emissions of global marine fisheries from 1950 to 2016, compared and discussed the differences with some of the previous research results of Parker et al., analyzed the changing trend, and pointed out that attention should be paid to the potentially important role of marine fisheries in carbon reduction strategies (Greer et al., 2019); Wang and Wang obtained carbon emission data by processing the oil consumption of marine

capture fisheries in China, then used the decoupling index to evaluate the relationship between carbon emissions and economic output, used the decomposition analysis method to determine the factors that affect carbon emissions, and finally analyzed the decoupling effect of 11 coastal provinces and cities in China using the decoupling effect model (Wang and Wang, 2022). Regarding the carbon emissions of the maritime transport industry, Bagoulla and Guillotreau calculated the environmental multiplier of the French maritime industry by using the input-output analysis method, obtained the emissions of various greenhouse gases, and analyzed the impact of the greenhouse gas emissions from the French maritime transport industry on air pollution (Bagoulla and Guillotreau, 2020); Wang et al. found that only a few of the large number of transportation activities are economically feasible. Among these economically feasible activities, there are a considerable number of trans-Arctic shipping routes that help reduce fuel consumption and greenhouse gas emissions, with a certain amount of greenhouse gas Emission reduction potential (Wang et al., 2021); Jing et al. noticed that the number of ships passing through the Northern Sea Route has increased, and pointed out that quantifying carbon dioxide emissions is very important to protect the environment. They used the system dynamics model and combined a variety of factors to predict carbon dioxide emissions, and proposed some effective measures to reduce carbon emissions according to the forecast results (Jing et al., 2021). Wu et al. reviewed the past studies related to the shipping industry and the Emissions Trading System (ETS), pointed out that the carbon dioxide emissions of the shipping industry should not be ignored by carbon emission reduction measures, and revealed the driving factors, challenges and possible consequences for adding the shipping industry to the ETS (Wu et al., 2022). The above achievements have laid the foundation for the research on the carbon emissions of the marine industry. However, a lot of researches are limited to the carbon emissions of a specific sub-sector, and there are many other sub-sectors that are not covered.

In addition, many studies quantify the carbon emissions of subdivided marine industries, but rarely consider the structure of carbon emissions. Graphs and networks are effective methods for structural problems, whereby scholars can build a variety of networks according to their research content, and use various network indicators and theories to analyze structural problems. For example, Tang et al. constructed China's interregional embodied energy flow network from the perspective of the regions and sectors, analyzed its structural characteristics, and identified the key sectors and regions for implementing energy conservation policies (Tang et al., 2019); while Wang et al. established a global network of embodied rare earths flows, identified China as an important network node through the network centrality theory, and analyzed the outflow path of embodied rare earths in China and the dependence of other economies on embodied rare earth (Wang et al., 2019).

The graph and network approach is very common in the study of carbon emission and carbon transfer, and a large number of studies have adopted the method of combining the input-output and network theory. The input-output approach is a common method for quantifying carbon emissions and is useful for analyzing trends in carbon transfer. For example, Wang et al. used the environmental input-output method to calculate and analyze the intermediate carbon emission transfer in 140 countries or regions around the world, and analyzed the carbon transfer in China (Wang et al., 2020); Shi et al.

combined the input-output method and the structural path analysis to describe the flow process of direct carbon emissions into embodied carbon emissions in China's supply chain (Shi et al., 2019); Wen and Wang used multi-regional input-output to quantify the carbon transfer between different provinces and different sectors in China (Wen and Wang, 2019); Zhou et al. used multi-regional input-output method to estimate the embodied carbon emissions between regions in China (Zhou et al., 2018). It is also difficult to study the problem of structure by only using the input-output method, so many scholars have combined it with network theory to dig deeper into the characteristics of network structure, and have conducted research at different levels. In the sector-level research, Jiang et al. combined input-output model and complex network theory to construct a production chain network based on carbon flow, and established indicators from both direct and indirect perspectives to quantify the importance of sectors in carbon emissions (Jiang et al., 2021); Du et al. constructed an indirect carbon emission flow network, tested its small-world nature, and identified key sectors based on complex network theoretical indicators (Du et al., 2018). At the regional level, Duan et al. combined multi-regional input-output (MRIO) with ecological network analysis (ENA) to identify key regions and sectors, and revealed inter-regional and inter-sectoral control or dependent relationship (Duan et al., 2018). Some scholars have advanced their research to the provincial level. For example, Lv et al. combined MRIO and social network analysis (SNA) to establish a provincial-level embodied carbon transfer network in China, analyzed network density, centrality, clustering characteristics and other indicators, revealed the spatial distribution and different roles of provinces, and used the quadratic assignment procedure (QAP) to study the influencing factors of carbon transfer networks (Lv et al., 2019); Similarly, Chen and Meng established China's carbon transfer network from the perspectives of provinces and sectors based on MRIO and complex network theory, analyzed topology, clustering characteristics and final demand decomposition, identified key sectors and critical paths, and used QAP to analyze the influencing factors of carbon transfer network (Chen and Meng, 2020).

Under the background that there are few studies on the carbon emission structure of the marine industry, this paper studies the carbon emission structure of China's marine industry to supplement the existing research. First, this paper splits 12 marine industries from the 2020 input-output table of 149 sectors, integrates them into 57 sectors based on the industry classification of the Energy Statistical Yearbook, and calculates the carbon transfer coefficient between industries based on energy consumption data, screened out the important inter-industry correlations according to the Weaver-Thomas combination index (hereinafter referred to as the W-T index) and built a marine industry carbon emission network. Secondly, we analyze the structural characteristics of the marine industry carbon emission network through the carbon transfer coefficient, degree, centrality and other network indicators, and identify the key sectors and their roles and functions in the network. Finally, we put forward some industrial-level carbon emission reduction recommendations. Our research incorporates the marine industry into the study of industrial carbon emissions, which helps to further understand the carbon emission structure, identify the key industries for carbon emission reduction. Our research also helps to formulate the effective ways to carbon emission reduction measures according to the carbon emission structure and take more targeted emission reduction measures.

2 Methods and data

2.1 Calculation of carbon transfer coefficient

Since the production process of a marine industry product has a direct or indirect consumption relationship to the intermediate products of other marine industries, and the product produces carbon emissions during the production process, carbon emissions are also transferred between industries with the transfer of products. For example, marine fishery j 's fishing fleet has a demand for fossil fuels, and marine oil and gas industry i emits CO_2 when producing fossil fuels. Then marine oil and gas industry i transfers CO_2 emission responsibility to marine fishery j , and carbon transfer occurs between the two industries. Similarly, the marine shipbuilding industry k has a demand for the electric energy produced by the marine electric power industry j , and the fossil fuels of the marine oil and gas industry i are used in the production process of the electric energy, so the CO_2 emission responsibility in the production process of the marine oil and gas industry i is transferred to the marine shipbuilding industry k . As a result, due to the technical and economic links between various industries, a complex inter-industry CO_2 transfer network has been formed.

According to the above analysis, the transfer of CO_2 emissions between marine industries is related to two factors: one is the demand for the intermediate input products of marine industry i in the production process of marine industry j ; the other is the CO_2 emissions per unit product produced by the marine industry i . Based on this, we define the carbon transfer coefficient to describe the CO_2 emission transfer relationship between industries: If a certain amount of intermediate input products of marine industry i are required in the production process of unit marine industry j product, the amount of CO_2 emitted by these intermediate input products in the production process is called the carbon emission transfer coefficient, which is expressed by c_{ij} . In order to express the two influencing factors, the carbon emission transfer coefficient is expressed as the following formula:

$$c_{ij} = a_{ij} \times e_i \quad (1)$$

Where a_{ij} is the direct consumption coefficient of marine industry j to marine industry i , which represents the amount of i industry product consumed when j industry produces a unit of product, this item can be calculated by the marine industry input-output table obtained by splitting; The definition of e_i here is the CO_2 emissions per unit product produced by the marine industry i , which can be calculated through the input-output table and the Energy Statistical Yearbook. The calculation method of e_i is as follows.

The first step is to calculate CO_2 the emissions from the fossil energy consumption of the marine industry. The basic idea is to multiply the consumption of various fossil energy by the CO_2 emission coefficient of the corresponding energy, and then sum it up (it is explained here that electricity is not included in the calculation because electricity consumption does not produce direct CO_2 emissions). The total CO_2 emissions of fossil energy consumed by marine industry i are represented by G_i , which can be calculated by the formula:

$$G_i = \sum_{s=1}^8 u_s(i) \times f_s \quad (2)$$

Where $u_s(i)$ represents the total consumption of the s th fossil energy in the production of marine industry i , which can be calculated from the China Energy Statistical Yearbook. Among them, fossil energy is divided into 8 categories (coal, crude oil, natural gas, coke, gasoline, kerosene, diesel, fuel oil). The f_s represents the CO_2 emission coefficient of the s th fossil energy.

The second step is to calculate the ratio of G_i obtained in the previous step to the total output X_i of marine industry i , which represents the CO_2 emissions per unit product produced by marine industry i . The formula is as follows:

$$e_i = G_i/X_i \quad (3)$$

The carbon transfer coefficients between all marine industries are expressed in the form of a matrix, as follows:

$$C = EA \quad (4)$$

Where C represents the carbon transfer coefficient matrix of the marine industry composed of elements c_{ij} , A represents the direct consumption coefficient matrix composed of a_{ij} , and E represents the diagonal matrix generated by e_i as the element.

2.2 Construction of carbon transfer network in marine industry

The carbon transfer coefficients between marine industries calculated above depict the emissions and correlation strengths of carbon transfers between different marine industries. Considering the marine industry as a network node and the carbon transfer coefficient between marine industries as a link, a carbon transfer network between all marine industries can be constructed. The carbon transfer coefficients between different marine industries are different, so the correlation of carbon transfer between different marine industries is also different. In general, the link with a larger carbon transfer coefficient plays a more critical role in the network, while some links with a smaller coefficient play a negligible role in the network, which leads to not all nodes and links in a complete carbon transfer network deserve significant attention. Therefore, in order to deeply explore the characteristics of the network structure, it is necessary to screen out the links with important roles and construct the carbon transfer network of the marine industry that our research needs.

In the determination of the critical value, most models adopt the method of subjective experience value, such as the empirical value of 0.2, 0.5, or use the average value. These methods can identify important associations to a certain extent, but they are highly subjective. Therefore, this paper uses the Weaver-Thomas index (hereinafter referred to as the W-T index) to determine the critical value in an endogenous manner. The W-T index is an effective tool to determine the significance index, which was first proposed by Weaver and improved by Thomas later. It identifies key elements in a numerical sequence by comparing an observed distribution with a hypothetical distribution to establish a closest approximate distribution. The W-T index is very effective when determining significance arrays from non-uniform arrays. Combined with the characteristics of inter-industry carbon emissions, this paper uses the W-T index to determine the critical value.

The inter-industry carbon transfer coefficient calculated according to the split input-output table of the marine industry is an $n \times n$

square matrix. We use $C(i, 1), C(i, 2), \dots, C(i, j), \dots, C(i, n) (i = 1, 2, \dots, n)$ to represent the n th indicator value under the i th sample, and $C(1, j), C(2, j), \dots, C(i, j), \dots, C(n, j) (j = 1, 2, \dots, n)$ to represent the n th sample under the j th indicator. Under this definition, the W-T index is calculated from the column direction of the carbon transfer coefficient between marine industries. The calculation steps are as follows:

Step 1, arrange the samples under each indicator of carbon transfer coefficient matrix C , that is, $C(1, j), C(2, j), \dots, C(i, j), \dots, C(n, j) (j = 1, 2, \dots, n)$, from large to small to obtain a new adjustment matrix $F(i, j) (i, j = 1, 2, \dots, n)$. Then, set a matrix $IndexC(i, j)$ representing the correspondence between the positions of the matrix $C(i, j)$ and the matrix $F(i, j)$, and each element value in the matrix $IndexC(i, j)$ is the arrangement number of the matrix $F(i, j)$ element in the j th column of the original matrix $C(i, j)$.

Step 2, calculate the W-T index matrix $w(i, j)$ corresponding to the matrix $F(i, j)$:

$$w(i, j) = \sum_{i=1}^n \left[s(k, i) - 100 \times \frac{F(k, j)}{\sum_{l=1}^n F(l, j)} \right]^2 \quad (5)$$

$$s(k, i) = \begin{cases} 100/i (k \leq i) \\ 0 (k > i) \end{cases} \quad (6)$$

The minimum value of each column in the W-T index matrix $w(i, j)$ is formed into a row vector α , which can be expressed as $\alpha = \min \{ w(1, j), w(2, j), \dots, w(n, j) \}$. Then a position vector β is constructed according to the corresponding position of each element in the vector α in the matrix $w(i, j)$.

Step 3, construct a 0–1 matrix B according to the position vector β . The construction principle is to compare i in any element $B(i, j)$ of the B matrix with the element value of the j th column of the position vector β . If $i < \beta(1, j)$, then $B(i, j) = 1 (i = 1, 2, \dots, n)$, otherwise it is 0, so the 0–1 matrix B is obtained.

Step 4, readjust the element position of 0–1 matrix B . The position of the 0–1 matrix is restored according to the position relationship matrix $IndexC(i, j)$ obtained in the first step, that is, the original industrial relationship is restored. We study the carbon transfer between industries, and do not study the transfer within the industry. Thus, the diagonal elements with a value of 1 in the 0–1 matrix are changed to 0. Finally, the 0–1 matrix C^* of the marine industry carbon transfer network is obtained.

The required carbon transfer network between marine industries can be drawn according to the 0–1 matrix C^* . For any element $C^*(i, j)$ in C^* , if it is 1, it means that there is a strong correlation between marine industry j and marine industry i , and the two industries can be connected, otherwise they are not connected.

2.3 Analysis of marine industry carbon transfer network structure

2.3.1 Definition and calculation method of node degree

Node degree is an important indicator to describe the direction and intensity of carbon transfer of a single industry in the marine

industry carbon transfer network. In the carbon transfer network, the number of edges that node i directly points to other nodes is called the out-degree of node i , denoted as Ok_i ; the number of edges that other nodes directly point to node i is called the in-degree of node i , denoted as Ik_i , the degree of node i is the sum of node out-degree Ok_i and node in-degree Ik_i . Assuming that the adjacency matrix of the marine industry carbon transfer network with n nodes is $A = (a_{ij})_{n \times n}$, then there are:

$$Ok_i = \sum_{j=1}^n a_{ij} \quad (7)$$

$$Ik_i = \sum_{i=1}^n a_{ji} \quad (8)$$

$$k_i = Ok_i + Ik_i \quad (9)$$

2.3.2 Definition and calculation method of betweenness centrality and eigenvector centrality

Betweenness centrality is an indicator that characterizes the importance of a node by the number of shortest paths through node i . Assuming that the number of shortest paths for material and information transmission between any two nodes s and t is g_{st} , and the number of shortest paths passing through the third node i is n_{sp}^i , the betweenness centrality BC_i of the nodes can be expressed as:

$$BC_i = \sum_{s \neq i \neq t} \frac{n_{sp}^i}{g_{st}} \quad (10)$$

Eigenvector centrality is an index that reflects the importance of a node through the importance of its neighbors. x_i is the importance value of node i , c is the proportional constant, and a_{ij} is the element of the adjacency matrix of the network, there are:

$$EC_i = x_i = c \sum_{j=1}^N a_{ij} x_j \quad (11)$$

Written in the form of a matrix, it can be expressed as $x = cAx$, and transformed into $Ax = c^{-1}x$, then x can be understood as an eigenvector with the eigenvalue c^{-1} of matrix A , so it is called eigenvector centrality.

2.3.3 Basic correlation structure of marine industry carbon transfer network

The basic association structure of the marine industry carbon transfer network emphasizes connecting all nodes in the network with the largest transfer weight and the fewest edges. This is mainly to find out the core path of carbon transfer in the marine industry carbon transfer network, and to identify the sub-networks with the strongest influence and the simplest structure.

The network N is directly constructed based on the unscreened carbon transfer coefficient matrix C , assuming $N = (V, E, W)$, V is the node set, E is the edge set, W is the weight of the edge, for any edge $(u, v) \in E$, there is a weight $w(u, v) \in W$, and $N_t = (V_t, E_t)$ is a generated subgraph N_t of the network N , if it satisfies: ① it is an acyclic connected network, ② $V_t = V$, ③ $E_t \subseteq E$, ④ $w(F) = \sum_{(u,v) \in T} w(u, v)$ is the largest, then N_t is called the basic correlation tree of marine industry

carbon transfer network, and the network structure reflected by N_t is called the basic correlation structure of marine industry carbon transfer network. The basic correlation structure algorithm of the marine industry carbon transfer network is designed as follows:

Step 1: Set an empty set E_t with a weight of 0.

Step 2: Select the edge $e_1 \in E$ with the largest edge weight and put it into E_t , so that $w(e_1)$ satisfies the maximum weight condition.

Step 3: If E_t already contains edge e_1, e_2, \dots, e_i , select $e_{i+1} \in E \setminus \{e_1, e_2, \dots, e_i\}$ so that $N[\{e_1, e_2, \dots, e_{i+1}\}]$ does not contain a circle, and $w(e_{i+1})$ is as large as possible;

Step 4: If $i < |V| - 1$, go back to step 3; if $i = |V| - 1$, stop.

2.4 Data sources

With the rapid development of the marine economy and its increasing role in China's national economy, the carbon emissions of the marine industry also deserve attention. This paper splits the marine industry from the national industry classification, aiming to study the carbon emission structure of the marine industry. The data sources are as follows: ① The 2020 China Marine Economy Statistical Bulletin issued by the Ministry of Natural Resources; ② 149 sectors 2020 National Input-Output Table issued by National Bureau of Statistics; ③ National Economic Industry Classification (GB/T 4754-2017) updated in 2017; ④ China Energy Statistical Yearbook 2021 issued by the National Bureau of Statistics.

3 Results

3.1 Source and processing of marine industry data

This paper subdivides the marine industry into 12 sectors, namely marine fishery, offshore oil and gas industry, marine mining industry, marine salt industry, marine chemical industry, marine biomedicine industry, marine power industry, seawater utilization industry, marine shipbuilding industry, offshore engineering construction industry, marine transportation industry and coastal tourism industry, and then constructs 12-sector marine industry input-output table for subsequent calculation and analysis.

The complex relationship between the marine industry and the national economy makes it difficult to separate (for example, the offshore oil and gas industry is included in oil and gas extraction products, extraction ancillary activities and other mining products, refined petroleum and nuclear fuel processing products; marine salt industry is included in non-metallic mineral processing products and other food products), and the 12 marine industries are not directly listed in the official input-output table of 149 sectors in 2020. Therefore, in order to obtain the 12-sector marine industry input-output table, we need to split the existing national input-output table.

The input-output table splitting process must follow the basic constraints of rows and columns, that is, "intermediate use + final use = total output, intermediate input + initial input = total input, total output = total input". According to this condition, horizontal splitting and vertical splitting are carried out for the relevant sectors respectively. Finally, the input-output table of 161 sectors including the marine industry in 2020 is obtained. The specific splitting process is as follows:

① Determine the split weight of the marine industry in the input-output table

The 2020 China Marine Economy Statistical Bulletin and the National Economic Industry Classification (GB/T 4754-2017) provide the conceptual definitions of the marine industry and the national economic industry. Based on this, we find out the corresponding industries that need to be split in the input-output table for each marine industry, and the added value data of these industries can be found from the marine economic statistical bulletin and the input-output table. After the added value is determined, the ratio of the added value of the marine industry to the added value of the corresponding industry (the sum of the added value if there are multiple corresponding industries) is the split weight of the marine industry. The sectoral correspondence and split weights are shown in **Table 1** below.

② Splitting and sorting of input-output tables

According to the split weights in the above table, the corresponding sectors in the national 149 sectors' input-output table are split. Complying with the above row and column constraints, the data of 12 marine industries are extracted by weight, and the data of the corresponding marine industry need to be subtracted from the divided industries. In addition, all marine industries need to be split horizontally and vertically. In the splitting process, the first quadrant needs to be split in both the row and column directions, the second quadrant only needs to be split horizontally, and the third quadrant only needs to be split vertically.

3.2 Processing of energy data and integration of input-output tables

The data used in this paper are the energy data by industry for 2020, which come from the China Energy Statistical Yearbook 2021 issued by the National Bureau of Statistics. The industry is subdivided into 47 categories in the energy statistical yearbook, which is different from the classification of the input-output table. There are 161 sectors in the input-output table obtained by splitting and sorting in the previous step, which is far more than the number of categories in the energy statistical yearbook. Therefore, both the split of the marine industry in the energy data and the consolidation of the input-output table are based on the industry classification of the energy statistical yearbook.

The splitting process of the marine industry in the energy data is similar to the splitting process in the input-output table. Firstly, the industries are determined that need to be split corresponding to the marine industry in the energy data, and the added value is used to calculate the split weights of the marine industry. Secondly, the corresponding industries in the energy data are split according to the weights, and the constraint that the sum of each column remains unchanged is obeyed. The difference from the split of the input-output table is that it only needs to be split horizontally. Finally, the energy data for 57 sectors including the marine industry is obtained. The inter-industry correspondence and the split weights used in the paper are shown in **Table 2**.

The merging of the input-output table still needs to follow the above constraint. According to the industry classification standard in the energy data, the input-output table with the 161 sectors obtained above is merged again, and it needs to be split horizontally and

TABLE 1 Input-output table marine industry split weight.

Marine industry	Added value (100 million yuan)	Split industry	Added value of the split industry (100 million yuan)	Split weight
Marine fishery	4712	Fishery products	7674.7923	0.39
Offshore oil and gas industry	1494	Agriculture, forestry, animal husbandry and fishery service products Processed aquatic products Oil and gas extraction products Mining ancillary activities and other mining products Refined petroleum and nuclear fuel processing products	3365.3399 997.5042 6772.379 853.8923 7866.1983	0.1
Marine mining industry	190	Coal mining and washing products Ferrous metal mining products Non-ferrous metal mining products Non-metallic ore mining and dressing products Mining ancillary activities and other mining products	11999.8498 3534.4637 2593.6041 4302.8915 853.8923	0.01
Marine salt industry	33	Non-metallic ore mining and dressing products	4302.8915	0.005
Marine chemical industry	532	Other food Basic chemical raw materials Fertilizer Pesticide Paints, inks, pigments and similar products Synthetic material Specialty chemical products and explosives, pyrotechnics, fireworks products	2847.809 5208.3628 1583.8308 539.0128 1477.6491 4424.4097 4006.2478	0.03
Marine biomedicine industry	451	Daily chemical products Pharmaceutical products Other food	1092.3268 8583.603 2847.809	0.04
Marine power industry	237	Electricity and heat production and supply industry	21834.9006	0.01
Seawater utilization industry	19	Water production and supply industry	2049.9066	0.01
Marine shipbuilding industry	1147	Ships and related installations Metal Products, machinery and equipment repair services	787.8183 473.1751	0.91

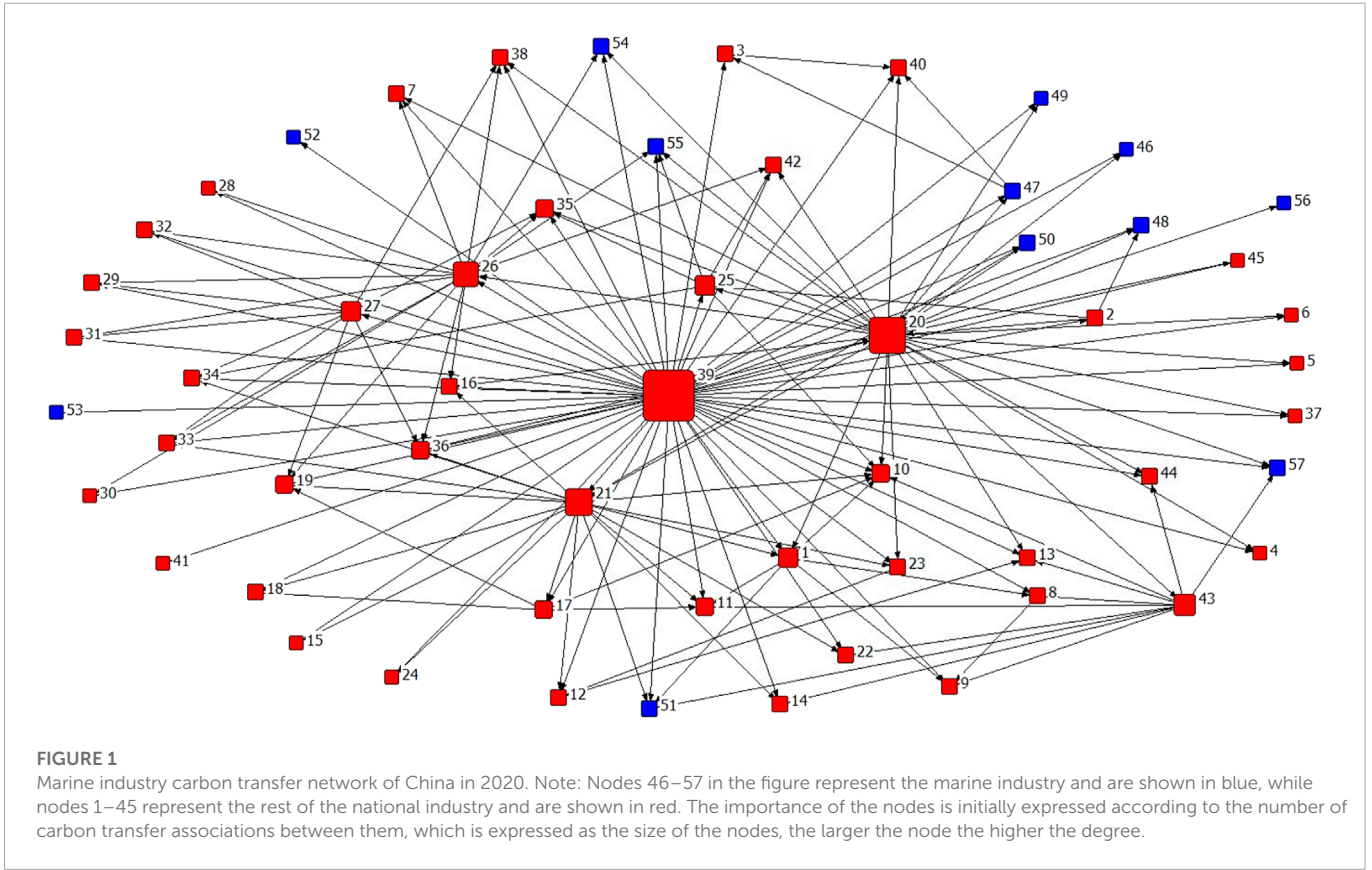
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TABLE 1 (Continued) Input-output table marine industry split weight.

Marine industry	Added value (100 million yuan)	Split industry	Added value of the split industry (100 million yuan)	Split weight
Offshore engineering construction industry	1190	Railway, road, tunnel and bridge engineering construction	16529.732	0.04
		Other civil engineering buildings	7798.8506	
Marine transportation industry	5711	Construction and installation	7417.346	0.37
		Water passenger transport	55.7742	
		Waterborne cargo transportation and transportation ancillary activities	2800.1965	
		Pipeline transportation	604.0131	
		Multimodal and shipping agency	4804.3962	
		Handling and warehousing	3169.0709	
Coastal tourism industry	13924	Postal	3990.9068	0.3
		Stay	3306.7938	
		Food	11245.2646	
		Business services	28561.5645	
		Physical education	928.9314	
		Entertainment	2606.3993	

TABLE 2 Split weight of marine industry based on energy data.

Marine industry	Industrial added value (100 million Yuan)	Split industry	Added value of the split industry (100 million yuan)	Split weight
Marine Fishery	4712	Agriculture, forestry, animal, husbandry and fishery	82174.7242	0.0573
Offshore oil and gas industry	1494	Extraction of petroleum and natural gas	6772.379	0.0845
Marine mining industry	190	Support activities for mining and mining of other ores	853.8923	
		Processing of petroleum, coal and other fuel	10059.0764	
		Mining and washing of coal	11999.8498	0.0082
		Mining and processing of ferrous metal ores	3534.4637	
		Mining and processing of non-ferrous metal ores	2593.6041	
		Mining and processing of non-metal ores	4302.8915	
		Support activities for mining and mining of other ores	853.8923	
Marine salt industry	33	Mining and processing of non-metal ores	4302.8915	0.0034
		Manufacture of foods	5288.7688	
Marine chemical industry	532	Manufacture of raw chemical materials and chemical products	18331.8397	0.029
Marine biomedicine industry	451	Manufacture of medicine	8583.603	0.0325
		Manufacture of foods	5288.7688	
Marine power industry	237	Production and supply of electric power and heat power	21834.9006	0.0109
Seawater utilization industry	19	Production and supply of water	2049.9066	0.0093
Marine shipbuilding industry	1147	Manufacture of railway ship, aerospace and other transport equipment	3224.1583	0.3102
		Repair service of metal products, machinery and equipment	473.1751	
Offshore engineering construction industry	1190	Construction	72009.294	0.0165
Marine transportation industry	5711	Transport, storage and post	45677.4937	0.125
Coastal tourism industry	13924	Wholesale and retail trades, hotels and catering services	109912.3998	0.0274
		Others and resident life	398111.6489	



vertically respectively. Finally, a 57×57 sectoral input-output table containing the marine industry is obtained. The energy consumption data and input-output table of 57 sectors are presented in the [Supplementary Material](#).

3.3 Analysis of carbon transfer coefficient of marine industry

Based on the definitions of the marine industries, the 2020 China Input-Output Table and the Energy Statistics Yearbook, the industries were split and combined to maintain a consistent industry classification, and the final industries and their codes are shown in [Table A1](#) in [Appendix A](#). A total of 3,249 carbon transfer coefficients, including 410 zero values, were obtained according to the methodology in 2.1. Excluding the coefficients for industry self-association, a total of 160 inter-industry strong associations were screened using the Weaver Index. The marine industry carbon transfer network of China in 2020 was constructed using the industries as nodes and the strong association as the connected edge criterion, as shown in [Figure 1](#).

The carbon transfer coefficients of the industries calculated above are ranked and the key industries are initially identified according to the rankings. First of all, the carbon transfer coefficients of all industries are ranked and the top 30 of them are taken out. The results are shown in [Table 3](#). Among the outflow industries, production and supply of electric power and heat power (39) appears 14 times, processing of petroleum, coal and other fuels (20) appears 9 times,

manufacture of raw chemical materials and chemical products (21) and smelting and pressing of ferrous metals (26) appear 3 times respectively, and non-ferrous metal smelting and rolling processing industry appears 1 time. In terms of frequency, No. 39 and No. 20 industries appear 76.7% of the time, so they are the key carbon outflow industries. In terms of the size of carbon transfer coefficient, the sum of carbon transfer coefficient of No. 39 industry as an outflow industry accounts for 61.07% of the total of the top 30, which is still the most critical outflow industry. Among the inflow industries, 9 of the top 30 industries are marine industries, accounting for 30% of the total, indicating the importance of marine industries in the inflow industries, among which marine chemical industry (50) needs special attention.

Secondly, the carbon transfer coefficients of all marine industries are ranked, and again the top 30 of them are taken out, and the results are shown in [Table 4](#). Among the outflow industries, offshore oil and gas industry (47) appears 10 times, marine transportation industry (56) appears 9 times, marine power industry (52) appears 8 times, marine chemical industry (50) and marine fishery (46) appear 2 times and 1 time respectively. No. 47, No. 56 and No. 52 industries have similar frequency and appear 90% of the time cumulatively, which are the most critical outflow marine industries. On the whole, the carbon transfer coefficient of marine industries is small, but the sum of carbon transfer coefficient of offshore oil and gas industry (47) accounts for 81.37% of the sum of the top 30, which obviously deserves key attention. Among the inflow industries, 12 marine industries appear with similar frequency and are relatively evenly distributed, with no marine industries requiring special attention.

TABLE 3 Top 30 carbon emission coefficients among all industries in 2020.

Rank	Carbon transfer coefficient	Sector i	Sector j	Rank	Carbon transfer coefficient	Sector i	Sector j
1	2.3946	39	52	16	0.4233	39	48
2	2.3946	39	39	17	0.4124	20	20
3	1.0261	39	41	18	0.4101	39	26
4	1.0261	39	53	19	0.4016	39	25
5	0.7410	20	56	20	0.4014	39	21
6	0.6986	39	5	21	0.4014	39	50
7	0.6311	39	6	22	0.3680	20	6
8	0.6280	26	28	23	0.3669	39	28
9	0.5815	20	50	24	0.3645	21	21
10	0.5815	20	21	25	0.3645	21	50
11	0.5693	20	43	26	0.3512	20	23
12	0.5579	26	26	27	0.3418	27	27
13	0.5071	20	26	28	0.3397	20	47
14	0.5019	39	4	29	0.3313	21	24
15	0.4409	39	27	30	0.3023	26	55

TABLE 4 Top 30 carbon emission coefficients among marine industries in 2020.

Rank	Carbon transfer coefficient	Sector i	Sector j	Rank	Carbon transfer coefficient	Sector i	Sector j
1	0.2846	47	47	16	0.0047	52	48
2	0.0599	47	56	17	0.0044	52	50
3	0.0524	47	50	18	0.0032	56	50
4	0.0264	52	52	19	0.0031	52	49
5	0.0173	56	56	20	0.0026	56	47
6	0.0157	47	48	21	0.0024	56	51
7	0.0145	47	52	22	0.0024	56	54
8	0.0136	47	49	23	0.0022	52	47
9	0.0124	47	55	24	0.0020	50	53
10	0.0113	52	53	25	0.0018	52	55
11	0.0109	50	50	26	0.0017	56	49
12	0.0071	47	57	27	0.0017	52	54
13	0.0052	47	46	28	0.0016	56	57
14	0.0052	47	54	29	0.0015	56	46
15	0.0050	46	46	30	0.0015	56	55

3.4 Analysis of the basic network characteristics

3.4.1 Out-degree and in-degree analysis

The out-degree and the in-degree are important indicators to portray the connected edge relationships between nodes, which directly reflects the carbon inflow and outflow relationship between

industries. The greater the number of industries to which an industry exports carbon, the greater the out-degree, and similarly, the greater the number of other industries that export carbon to an industry, the greater the in-degree. The out-degree and the in-degree reflect the scope of carbon outflow and carbon inflow of an industry, respectively, and directly indicate the number of industries with which carbon transfer occurs. If carbon reduction measures are taken

TABLE 5 Out-degree and in-degree of each industry.

Industry	OutDegree	InDegree	Industry	OutDegree	InDegree	Industry	OutDegree	InDegree
1	5	3	20	30	3	39	54	0
2	3	1	21	18	2	40	0	4
3	1	2	22	0	3	41	0	1
4	0	2	23	1	3	42	0	4
5	0	2	24	0	2	43	10	1
6	0	2	25	5	3	44	0	3
7	0	3	26	15	2	45	0	2
8	1	3	27	9	1	46	0	2
9	0	4	28	0	2	47	3	2
10	0	7	29	0	3	48	0	3
11	0	5	30	0	2	49	0	2
12	1	3	31	0	3	50	0	3
13	0	4	32	0	3	51	0	4
14	0	3	33	0	4	52	0	1
15	0	2	34	0	4	53	0	1
16	0	4	35	0	5	54	0	3
17	4	2	36	0	5	55	0	4
18	0	3	37	0	2	56	0	1
19	0	5	38	0	4	57	0	3

for industries with large out-degree and in-degree, the impact will spread to neighboring industries, resulting in a wider range of carbon reduction effects.

The marine industry carbon transfer network of China in 2020 involves 57 industries, with no isolated nodes in the network, and the in-degree and out-degree of each industry is shown in [Table 5](#). In terms of out-degree, the out-degree can reflect the carbon outflow relationship of industries in the carbon transfer network, and the key carbon outflow industries can be directly judged according to the out-degree. There are 42 industries with zero degree of output, representing that these industries do not emit carbon to the outside. Production and supply of electric power and heat power (39), processing of petroleum, coal and other fuels (20), manufacture of raw chemical materials and chemical products (21), smelting and pressing of ferrous metals (26) and transport, storage and post (43) have higher emissions of 54, 30, 18, 15 and 10 respectively, while the remaining 10 industries have emissions between 0 and 10. Production and supply of electric power and heat power (39) has the highest out-degree and is the most critical industry in terms of outflow for almost all industries. In terms of in-degree, the industry with the highest in-degree is the manufacture of liquor, beverages and refined tea (10) with an in-degree of 7, representing the industry as the most critical carbon inflow industry, followed by industries with an in-degree of 5, such as manufacture of articles for culture, education, arts and crafts, sport and entertainment activities (19) and other manufacture (36). For the marine industry, only the offshore oil and gas industry (47) has a non-zero out-degree and all marine industries have a non-zero in-degree.

3.4.2 Analysis of betweenness centrality

In a network, material or information flows between nodes along connected edges and betweenness centrality reflects whether a node is at a key position in the flow process. In the process of industrial carbon transfer, the betweenness centrality reflects whether a particular industry plays a key role as a "bridge" or "intermediary" in the transfer process. If a large proportion of the shortest carbon transfer paths between two industries pass through a certain industry, that industry has a high betweenness centrality and is in a key position of inter-industry carbon transfer. The industry with higher betweenness centrality has more carbon transfer paths passing through it, which usually leads to its own larger carbon transfer, so it is important to take measures for this industry to reduce its own carbon transfer and control the carbon transfer between industries on both sides of it. The calculation results of the betweenness centrality of each industry are shown in [Table 6](#).

According to the calculated data, production and supply of electric power and heat power (39) and the processing of petroleum, coal and other fuels (20) both have a betweenness centrality greater than 10, which is significantly higher than that of the other industries, and have the strongest control over carbon flows, making them the most critical "intermediary" industries in the network. The betweenness centrality of the manufacture of raw chemical materials and chemical products (21), the smelting and pressing of ferrous metals (26), and the transport, storage and post industry (43) are between 1 and 10, and are the next most critical industries, while the betweenness centrality of the rest of the industries is less than 1 and less important. For the marine industries, the marine biomedical industry (51), coastal

TABLE 6 Betweenness centrality and eigenvector centrality of industries.

Industry	Betweenness	Eigenvector	Industry	Betweenness	Eigenvector	Industry	Betweenness	Eigenvector
1	0.392	0.150	20	16.963	0.362	39	65.695	0.514
2	0.022	0.102	21	4.919	0.252	40	0.048	0.095
3	0.000	0.065	22	0.114	0.081	41	0.000	0.047
4	0.000	0.080	23	0.032	0.112	42	0.013	0.114
5	0.000	0.080	24	0.000	0.070	43	1.409	0.121
6	0.000	0.080	25	0.391	0.145	44	0.082	0.091
7	0.000	0.100	26	3.205	0.219	45	0.000	0.080
8	0.092	0.079	27	0.790	0.132	46	0.000	0.080
9	0.092	0.079	28	0.000	0.067	47	0.048	0.095
10	0.329	0.152	29	0.007	0.079	48	0.000	0.090
11	0.180	0.105	30	0.000	0.067	49	0.000	0.080
12	0.043	0.090	31	0.007	0.079	50	0.000	0.103
13	0.179	0.100	32	0.007	0.079	51	0.125	0.095
14	0.114	0.081	33	0.031	0.102	52	0.000	0.047
15	0.000	0.070	34	0.051	0.096	53	0.000	0.047
16	0.011	0.124	35	0.079	0.126	54	0.000	0.100
17	0.135	0.111	36	0.068	0.136	55	0.013	0.114
18	0.000	0.080	37	0.000	0.080	56	0.000	0.033
19	0.096	0.113	38	0.044	0.112	57	0.082	0.091

tourism industry (57) and the Offshore oil and gas industry (47) have a relatively low betweenness centrality, but are significantly more important than the other marine industries and are important "intermediary" marine industries.

3.4.3 Analysis of eigenvector centrality

Eigenvector centrality describes the closeness between a node and the core nodes of its network. The basic idea is that the importance of a node depends not only on the number of its neighboring nodes, but also on the importance of the neighboring nodes. The higher the eigenvector centrality value of a node, the higher its closeness to the core node and the more important the node is. In the marine industry carbon transfer network, the more carbon transfer linkages an industry has with other industries, and the more central the associated industries are in the network, the stronger the eigenvector centrality of the industry. Industries with high eigenvector centrality are often key industries that need to reduce carbon emissions, and when they take measures to reduce their own carbon emissions, the effect can be further radiated to their neighboring key industries, which is important for the overall carbon reduction of key industries in the network. The results of the calculation of the eigenvector centrality for each industry are shown in [Table 6](#).

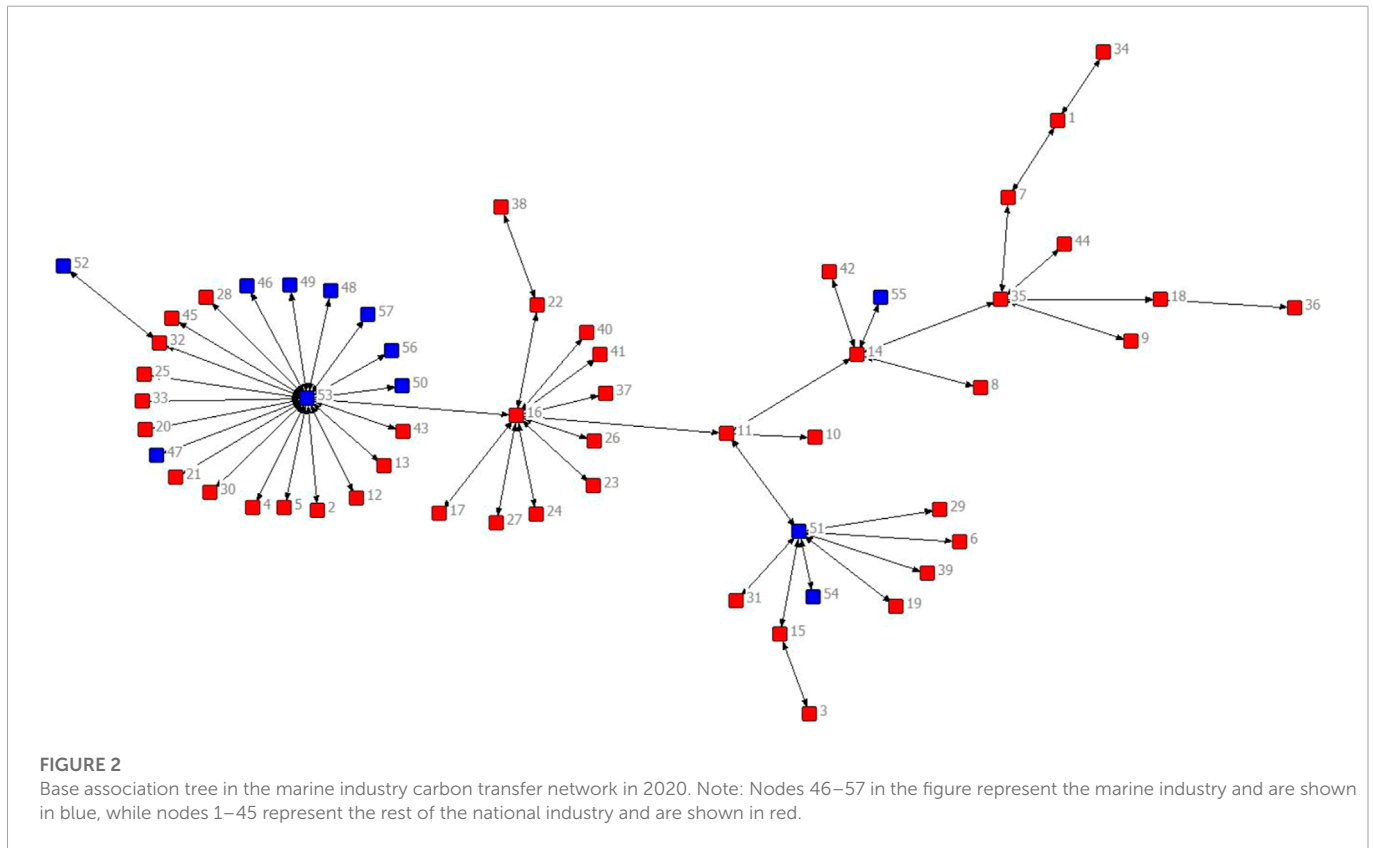
The results show that the production and supply of electric power and heat power (39) has the largest eigenvector centrality, significantly higher than other industries, and is the most important industry in the network. Processing of petroleum, coal and other fuels (20), manufacture of raw chemical materials and chemical products (21), and smelting and pressing of ferrous metals (26) have larger eigenvector centrality and are significantly different from

the remaining industries and are more important. The remaining industries have an even distribution of eigenvector centrality with no obvious gap. For the marine industries, except for the marine power industry (52), seawater utilization industry (53) and marine transportation industry (56), the remaining marine industries have no significant difference in eigenvector centrality and are significantly higher than the three industries, which are considered to be less important in the network and priority is given to the carbon reduction plans of other marine industries.

3.5 Analysis of the base association tree

The basic association tree is the core path in the network, which can clearly show the association structure among industries. In the marine industry carbon transfer network, the basic association tree reflects the largest carbon transfer weight among industries with the least association, its diameter is the longest carbon transfer path, and the center helps to determine the key industries in the carbon transfer network. It is important to reduce carbon emissions by taking measures on the diameter and key industries in the basic association tree.

The carbon transfer coefficient between industries is used as the weighting factor to identify the important edges with the greatest weight at each node, and the least number of edges is used to link all the nodes, which is the base linkage tree of the marine industry carbon transfer network in 2020, see [Figure 2](#). The maximum value of the furthest distance between any two nodes in the base association tree is called the diameter and the



minimum value is called the radius, as shown in the diagram, $52 \leftrightarrow 32 \leftrightarrow 53 \leftrightarrow 16 \leftrightarrow 11 \leftrightarrow 14 \leftrightarrow 35 \leftrightarrow 7 \leftrightarrow 1 \leftrightarrow 34$ is the diameter. The diameter contains 10 industries in total, which is the longest path in the basic association tree and an important basis for analyzing the network structure. Around the most important diameter, there are three more important flow paths, which are respectively connected to the No.16 industry, the No.11 industry and the No.35 industry. A node is considered to be a centroid of the graph if its furthest distance from any node in the graph is the radius. By definition, the centroid is located in the middle of the diameter, is closer to other industries, is more closely connected and is the key industry in the tree. In the base association tree, the centroid industry is No.11 industry and No.14 industry, and the industries furthest from the centroid are No.52 industry and No.34 industry, which are less connected to other industries. It is easy to see that the seawater utilization industry (No.53) has the largest number of links with other industries, and the industry is also located on the diameter, so it has an important position in the basic linkage tree. In addition, as many as seven marine industries are directly linked to the seawater utilization industry, indicating that it plays an important role in the carbon transfer process among marine industries.

4 Discussion

Based on the results of the above study, we have made some suggestions for the reduction of emissions in the carbon network of China's marine industries.

The carbon transfer coefficient directly reflects the intensity of the carbon transfer between two industries. From the results, the

electricity and heat production and supply industry (39) is the industry with the largest emission coefficient in the network and the industry with the most serious CO₂ emission problem. From the perspective of carbon transfer coefficient, carbon reduction measures for No.39 industry can be taken from two aspects. The first is the direct consumption coefficient. Usually, the demand for electricity and heat is very large and cannot be replaced, so it is difficult to achieve the goal of carbon emission reduction by reducing the direct consumption of its products. The second is the use of fossil energy. The CO₂ emission coefficient of each fossil energy is different. For example, diesel (3.25) is much higher than that of coal (2.69). Therefore, carbon reduction can be achieved by reducing the use of high-emission energy sources and changing the mix of fossil energy use, such as increasing the use of natural gas with the lowest emission factor to replace other high-emission fossil energy sources. In addition, due attention should be given to the use of renewable energy sources, such as promoting the development and use of hydro, wind, solar and geothermal energy. The conversion efficiency of electricity and heat can also be improved by promoting technological innovation, taking into account the loss of energy in the conversion process. In terms of carbon transfer coefficients among marine industries, similarly, for the offshore oil and gas industry (47), the marine transportation industry (56) and the marine power industry (52), which are important in the marine industry, emission reductions should be achieved through the combined use of fossil energy, and the marine power industry should also take measures to enhance the development and utilization of tidal energy, wave energy and wind energy to achieve emission reductions.

In terms of out-degree, No. 39 industry is clearly the industry with the largest out-degree, with carbon emissions covering almost all industries in the network. If measures are taken to reduce direct

carbon emissions from No. 39 industry, the carbon reduction effect will radiate throughout the network. The processing of petroleum, coal and other fuels (20) has an output of 30, covering most of the network nodes, and is the next most critical outflow industry. In terms of in-degree, industries with a high in-degree imply a large inflow of carbon, and this industry is likely to be an important industry in the indirect transfer of carbon. In order to avoid a large amount of indirect transfer, it is equally necessary to pay attention to the carbon emission reduction of such industries. It is worth noting that among the 12 marine industries, except for the offshore oil and gas industry (47), which has an out-degree of 3, the others are all 0, which proves that the overall carbon emission of the marine industry is relatively weak, and it is an inflow-based industry. Emphasis should be given to reducing carbon inflows from upstream industries, while at the same time considering the low-carbon production of the industry's own products. In terms of betweenness centrality and eigenvector centrality, the electricity and heat production and supply industry (39) is not only the core industry in the entire network, but also the most critical "bridge". As the core of the network, industries closely related to it also have a strong influence. As an "intermediary", it has a strong control over the carbon transfer in the network. Due to the strong controlling role of No. 39 industry, reducing its carbon emissions will have the same effect of reducing carbon flows or emissions of neighboring industries. The realization of carbon emission reduction should not only take measures at important nodes, but also consider important carbon transfer paths. As the longest path in the basic association tree, the diameter of the industry has an important impact on the entire network and should be paid attention to. In addition to the industries at the end of the diameter, the other industries on the diameter play an important supporting role for the entire network structure, and are also critical to the effective implementation of carbon emission reduction. Trying to reduce the path length will effectively reduce the carbon emission of the entire network. In addition, three important paths that are diametrically connected should also be paid attention to.

Compared with the previous literature, this paper makes the following contributions: firstly, the marine industry is split from the National Industry Classification and all industries are analyzed together; secondly, the carbon transfer coefficient of the industry is calculated and a marine industry carbon transfer network is constructed based on this; thirdly, the carbon transfer problem is analyzed from the perspective of the network structure, which enriches the research in this field. In addition, there are some limitations and shortcomings in this paper: firstly, this paper calculates the split weights of the marine industry based on the value added of the industry, the scientific nature of the value added index is open to question, and future research can adopt a more reasonable method of determining the weights; secondly, the indicators used to analyze the network structure can be further optimized, and future research can adopt or design more effective indicators to reflect the characteristics of the network structure.

5 Conclusion

Industry linkages are formed between industries based on their techno-economic linkages and various other linkages, and are expressed in the form of input and output activities of various products, so the carbon emitted during the production of products

is also transferred through industry linkages, and the responsibility for carbon emissions is also transferred between industries. This paper splits 12 marine industries from the industrial classification, calculates inter-industry carbon transfer coefficients based on national input-output tables and energy consumption data in 2020, constructs a marine industry carbon transfer network and analyses the structural characteristics of the network. This paper extends the study of the relationship between the carbon emission structures of marine industries, which can help in the formulation of carbon reduction policies. Based on the analysis of carbon transfer coefficients, out-degree and in-degree, betweenness centrality and eigenvector centrality, we conclude that the electricity and heat production and supply industry is always the most critical node in the network, with the greatest carbon emission intensity, and also assumes the most important role as a "bridge" in the network. For No. 39 industry, carbon emissions can be reduced directly by changing the mix of fossil energy use, and according to its strong control over neighboring industries, the carbon emission reduction effect will also radiate to neighboring industries. According to the analysis of the basic association tree, carbon emission reduction should pay attention to the critical path and its industry in the network, and try to reduce the length of the transmission path to reduce the carbon emission of the whole system. For the marine industry, the values of the indicators are generally low compared to other industries, and the industry has a limited role in the carbon transfer network. Currently, the marine industry is dominated by carbon inflows and mainly considers changing its demand for products from upstream industries to achieve emission reduction. The offshore oil and gas industry (47), as the most critical industry in the marine industry, presents obvious carbon outflow compared with other marine industries, and should also focus on the carbon emission of its own output.

Based on the above conclusions, this paper puts forward the following policy recommendations (1): The electricity and heat production and supply industry is the most critical industry in the carbon transfer network, and also an important driving force of China's rapid economic development, which needs to appropriately improve carbon emission standards and encourage the rationalization of its fossil energy use structure. It should also vigorously improve the construction of clean energy infrastructure through policy support, promote the innovative development of clean energy technologies, and increase the use of clean energy as a percentage, and these recommendations are also applicable to carbon emission reduction in the offshore oil and gas industry. (2) In general, the marine industry has a low position in the carbon transfer network and is a carbon inflow industry, so it needs to focus on its demand side and its own carbon emission reduction. It is recommended to promote the concept of low-carbon production in the marine industry, optimize the production structure, improve the low-carbon requirements for intermediate products in the upstream industry, and increase the proportion of low-carbon intermediate products in their production. At the same time, measures should be taken to promote the green technology innovation of the marine industry, appropriately improve the carbon emission standards and reduce the carbon emission of their own products. (3) Inter-industry carbon transfer covers the carbon outflow of upstream industries and the carbon inflow of downstream industries. Therefore, it is necessary to coordinate upstream and downstream industries and encourage inter-industry cooperation to reduce emissions. Upstream industries should take the initiative to reduce product carbon emissions through

measures such as green technology progress and improvement of industrial structure. Downstream industries should establish a strict intermediate product access mechanism to supervise and promote the low-carbon production of upstream enterprises and jointly undertake carbon emission reduction responsibility.

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

LWA: Idea formulation and conceptualization, design of research methods, assignment of tasks, supervision, writing. XZ: Data collection and processing, data visualization, writing and editing. LWE: Data collection and processing, writing and editing. WX: Assignment and supervision, funding acquisition, design research methodology, writing and review.

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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/fenrg.2023.1107796/full#supplementary-material>

Appendix A

In this study, after splitting and merging the final 57 industries, the industry names and codes are shown in [Table A1](#).

TABLE A1 Final industry division and code table.

Code	Industry	Code	Industry
1	Agriculture, forestry, animal husbandry and fishery	30	Manufacture of special purpose machinery
2	Mining and washing of coal	31	Manufacture of automobiles
3	Extraction of petroleum and natural gas	32	Manufacture of railway, ship, aerospace and other transport equipment
4	Mining and processing of ferrous metal ores	33	Manufacture of electrical machinery and apparatus
5	Mining and processing of non-ferrous metal ores	34	Manufacture of computers, communication and other electronic equipment
6	Mining and processing of non-metal ores	35	Manufacture of measuring instruments and machinery
7	Support activities for mining and mining of other ores	36	Other manufacture
8	Processing of food from agricultural products	37	Utilization of waste resources
9	Manufacture of foods	38	Repair service of metal products, machinery and equipment
10	Manufacture of liquor, beverages and refined tea	39	Production and supply of electric power and heat power
11	Manufacture of tobacco	40	Production and supply of gas
12	Manufacture of textile	41	Production and supply of water
13	Manufacture of textile, wearing apparel and accessories	42	Construction
14	Manufacture of leather, fur, feather and related products and footwear	43	Transport, storage and post
15	Processing of timber, manufacture of wood, bamboo, rattan, palm and straw products	44	Wholesale and retail trades, hotels and catering services
16	Manufacture of furniture	45	Others and resident life
17	Manufacture of paper and paper products	46	Marine fishery
18	Printing and reproduction of recording media	47	Offshore oil and gas industry
19	Manufacture of articles for culture, education, arts and crafts, sport and entertainment activities	48	Marine mining industry
20	Processing of petroleum, coal and other fuels	49	Marine salt industry
21	Manufacture of raw chemical materials and chemical products	50	Marine chemical industry
22	Manufacture of medicine	51	Marine biomedicine industry
23	Manufacture of chemical fibers	52	Marine power industry
24	Manufacture of rubber and plastics products	53	Seawater utilization industry
25	Manufacture of non-metallic mineral products	54	Marine shipbuilding industry
26	Smelting and pressing of ferrous metals	55	Offshore engineering construction industry
27	Smelting and pressing of non-ferrous metals	56	Marine transportation industry
28	Manufacture of metal products	57	Coastal tourism industry
29	Manufacture of general purpose machinery		