

Original Research

Measurement and Evaluation of Carbon Emission Efficiency in Logistics Industry – Taking Anhui Province as an Example

Heping Ding^{1,2}, Yu Wang¹, Wenjing Sheng¹, Yanting Li¹, Conghu Liu^{3,4*}

¹School of Business, Suzhou University, Suzhou 234000, China

²School of Transportation, Southeast University, Nanjing 210000, China

³School of Mechanical and Electronic Engineering, Suzhou University, Suzhou 234000, China

⁴Antai School of Economics and Management, Shanghai Jiaotong University, Shanghai 200000, China

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Abstract

Reducing carbon emissions from the logistics industry is beneficial for controlling environmental pollution and achieving sustainable development. To improve the carbon emission efficiency of the logistics industry (LCEY) for sustainable development, we propose social network analysis for LCEY evaluation and improvement. We established an evaluation index system for LCEY, measured it with the Super-SBM model considering unexpected output, created its spatial correlation matrix using a modified gravity model, and analyzed its spatial network characteristics with social network analysis. Taking 16 cities in Anhui Province as examples, this study demonstrates the implementation process of the method in detail. The results show that the average value of LCEY in 16 cities in Anhui Province increased from 0.3851 in 2012 to 0.7371 in 2021, showing a steady upward trend. The network efficiency decreased from 0.7619 in 2012 to 0.6857, indicating that the spatial spillover effect is significant, the spatial network relationship is gradually enhanced, and the network tightness must be improved. The characteristics of the spatial network are unbalanced; therefore, improving the structure of the benefit-spillover relationship is crucial. Hence, we propose countermeasures and suggestions to improve the LCEY to provide theoretical and methodological support for the research and management of the LCEY.

Keywords: logistics industry, carbon emission efficiency, Super-SBM model, social network analysis

Introduction

In recent decades, amid rapid economic growth, global environmental deterioration, and climate

warming, there has been a worldwide focus on controlling greenhouse gas emissions [1]. A crucial metric for assessing a region's overall strength is the logistics industry (LN), which is a fundamental, strategic, and leading sector that supports the growth of the national economy [2]. However, although LN encourages economic growth, its energy use, environmental pollution, and social impacts are also

*e-mail: liuconghu@ahszu.edu.cn,

Tel. 13305675380

increasing annually [3]. China, with the second-largest economy globally, has placed a high value on LN in recent times. Its GDP has tripled in the past decade, while the carbon emissions from the petrochemical energy consumed by LN have become increasingly serious [3].

To achieve global sustainable development (SP), it is critical to understand how to enhance the carbon emission efficiency of the logistics industry (LCEY) from the standpoints of environmental protection, social advancement, and economic development. Currently, the following research issues require immediate attention: (1) how to thoroughly measure and assess LCEY in light of SP; (2) how to analyze the complex relationship structure of LCEY and comprehensively investigate its spatial network characteristics; and (3) how to propose targeted suggestions for the government and employees. China's LN must consider economic development, environmental protection, and social impact and minimize the emission of environmental pollutants while promoting economic development, that is, achieving higher carbon emission efficiency.

Numerous scholars believe that, with rapid economic development, enhancing carbon emission efficiency (CEY) will advance the development of a low-carbon economy. In this study, we combined the literature into the following three aspects:

The first is the measurement and evaluation of the LCEY. The input and output perspectives are primarily used to construct an evaluation index system (EIS). Input indicators include human, financial, and material factors, while output indicators include economic output and carbon emissions [4]. The methods used to measure the CEY mainly include the life cycle method, energy analysis [5], stochastic frontier method, and data envelopment analysis [6]. Among them, for its advantages, data envelopment analysis (DEA) is widely used by scholars [7]. However, the traditional DEA models (CCR and BBC models) are radial models, have a certain bias in dealing with non-desired outputs, and are prone to inaccurate evaluation results in the actual measurement [8]; therefore, a more scientific model must be used to measure LCEY.

The second is the evolution mechanism of the LCEY. Based on measurements, scholars have mostly evaluated the LCEY in terms of time and space dimensions.

In the time dimension, DEA, Malmquist index, and kernel density estimation are combined to analyze the law and trend of the evolution of LCEY with time [9, 10], where the spatial autocorrelation model and synthetic control method are often used to analyze the spatial evolution properties of LCEY [11, 12]. However, most studies used "attribute data" to examine the spatial pattern properties of LCEY. Although some studies are based on geographical or economic proximity, analyzing the spatial network relationships of the LCEY among distant municipalities is impossible [13]. Therefore, this study focuses on the comprehensive and objective construction of the EIS of the LCEY and a more

scientific method for analyzing its evolution mechanism.

The third consists of countermeasures to improve the LCEY. Specific countermeasures and suggestions include developing economies, optimizing the layout of LN, strengthening the training of logistics talent, improving the level of logistics informatization, strengthening government investment, and optimizing the energy structure [14]. These recommendations and countermeasures are mostly used by linked businesses, sectors, governments, and other domains and have achieved certain results in practice. However, there are few countermeasures and suggestions from the perspective of spatial networks in existing research; consequently, it is necessary to formulate corresponding countermeasures to improve the spatial network characteristics of the LCEY.

Scholars have made considerable progress and valuable achievements in research on LCEY, but the following research gaps remain:

(1) Increasingly close inter-regional logistics activities gradually present complex spatial network characteristics, and it has become a new trend to reexamine logistics activities from the perspective of complex networks. Almost all the studies of LCEY are based on "attribute data" and described by a spatial econometric model. However, "attribute data" cannot accurately reflect the spatial network characteristics of LCEY. The social network analysis (SNA) approach is an effective tool for analyzing the overall state of "relational data." It has been in use for a long time but has rarely been applied to the study of LCEY.

(2) Most existing studies promote countermeasures and suggestions based on the influencing factors, and the LCEY presents different spatial network characteristics in different periods. Thus, the SP of the regional LD must be advanced, as must the global SP goals.

This study builds the LCEY's EIS, measures the LCEY using the Super-SBM model, creates a spatial network matrix using the updated gravity model, examines the LCEY's spatial network properties using the SNA technique. The contributions of this study are as follows:

(1) This study constructs LCEY's EIS based on the sustainability perspective, establishes the Super-SBM model of LCEY, and achieves scientific assessment and measurement of the efficiency level.

(2) The LCEY study employs the SNA method. A modified gravity model is used to build the spatial network matrix of the LCEY. The "attribute data" of LCEY is converted into "relational data," and both the overall and individual network characteristics are examined. The study also provides a new perspective on the study of LCEY.

(3) Based on analyzing the spatial network characteristics of LCEY, this study proposes more scientific and targeted countermeasures to improve LCEY. This will guide the government in the effective direction of the development of LN. It will also help the government direct pertinent practitioners.

Table 5. Spatial network centrality of LCEY in Anhui Province from 2012 to 2021.

Year	Item	Hefei	Huaibei	Bozhou	Suzhou	Bengbu	Fuyang	Huainan	Chuzhou	Luan	Maanshan	Wuhu	Xuancheng	Tongling	Chizhou	Anqing	Huangshan
2012	nDegree(%)	53.300	20.000	40.000	33.300	26.700	26.700	46.700	26.700	33.300	53.300	33.300	20.000	93.300	6.700	13.300	6.700
2013	nDegree(%)	53.300	26.700	40.000	33.300	33.300	40.000	46.700	26.700	33.300	46.700	40.000	20.000	93.300	6.700	13.300	6.700
2014	nDegree(%)	80.000	26.700	26.700	26.700	20.000	26.700	26.700	26.700	33.300	26.700	46.700	26.700	93.300	20.000	13.300	13.300
2015	nDegree(%)	93.300	26.700	26.700	20.000	33.300	26.700	26.700	20.000	26.700	80.000	73.300	20.000	40.000	33.300	33.300	20.000
2016	nDegree(%)	93.300	20.000	26.700	33.300	26.700	26.700	26.700	20.000	26.700	73.300	73.300	20.000	40.000	33.300	33.300	26.700
2017	nDegree(%)	93.300	20.000	26.700	33.300	33.300	26.700	26.700	20.000	26.700	86.700	73.300	20.000	46.700	26.700	26.700	26.700
2018	nDegree(%)	93.300	20.000	26.700	33.300	26.700	20.000	33.300	26.700	20.000	86.700	73.300	20.000	33.300	26.700	26.700	20.000
2019	nDegree(%)	86.700	20.000	26.700	33.300	33.300	20.000	33.300	26.700	20.000	80.000	80.000	20.000	20.000	20.000	20.000	20.000
2020	nDegree(%)	86.700	20.000	26.700	33.300	40.000	26.700	33.300	33.300	33.300	73.300	80.000	20.000	40.000	33.300	33.300	26.700
2021	nDegree(%)	86.700	26.700	26.700	26.700	26.700	26.700	33.300	33.300	33.300	73.300	86.700	20.000	46.700	33.300	33.300	26.700
2012	FreeClo(%)	60.000	55.600	62.500	60.000	57.700	57.700	65.200	57.700	60.000	68.200	60.000	53.600	93.800	50.000	53.600	50.000
2013	FreeClo(%)	60.000	57.700	62.500	60.000	60.000	62.500	65.200	57.700	60.000	65.200	62.500	53.600	93.800	50.000	53.600	50.000
2014	FreeClo(%)	83.300	57.700	57.700	57.700	55.600	57.700	57.700	57.700	60.000	57.700	65.200	57.700	93.800	55.600	53.600	53.600
2015	FreeClo(%)	93.800	57.700	57.700	55.600	60.000	57.700	57.700	55.600	57.700	83.300	78.900	55.600	62.500	60.000	60.000	55.600
2016	FreeClo(%)	93.800	55.600	57.700	60.000	57.700	57.700	57.700	55.600	57.700	78.900	78.900	55.600	62.500	60.000	60.000	57.700
2017	FreeClo(%)	93.800	55.600	57.700	60.000	60.000	57.700	57.700	55.600	57.700	88.200	78.900	55.600	65.200	57.700	57.700	57.700
2018	FreeClo(%)	93.800	55.600	57.700	60.000	57.700	55.600	60.000	57.700	55.600	88.200	78.900	55.600	60.000	57.700	57.700	55.600
2019	FreeClo(%)	88.200	55.600	57.700	60.000	60.000	55.600	60.000	57.700	55.600	83.300	83.300	55.600	55.600	55.600	55.600	55.600
2020	FreeClo(%)	88.200	55.600	57.700	60.000	62.500	57.700	60.000	60.000	60.000	78.900	83.300	55.600	62.500	60.000	60.000	57.700
2021	FreeClo(%)	88.200	57.700	57.700	57.700	57.700	57.700	60.000	60.000	60.000	78.900	88.200	55.600	65.200	60.000	60.000	57.700
2012	nBetween(%)	7.524	0.190	3.255	1.898	0.787	0.787	2.699	1.192	1.668	6.431	1.787	0.190	50.226	0.000	0.413	0.000
2013	nBetween(%)	7.365	0.635	2.732	1.445	1.326	2.732	2.461	1.113	1.541	4.312	2.748	0.159	48.162	0.000	0.413	0.000
2014	nBetween(%)	26.032	0.397	0.714	0.397	0.079	0.556	0.317	0.794	1.587	0.714	4.444	0.794	38.968	0.238	0.079	0.079
2015	nBetween(%)	32.619	1.190	0.967	0.476	2.078	0.967	0.173	0.173	0.173	15.635	15.635	0.173	0.649	0.173	0.173	0.173
2016	nBetween(%)	33.524	16.111	13.254	2.007	1.277	0.975	0.737	0.737	0.714	0.372	0.372	0.372	0.372	0.238	0.182	0.182
2017	nBetween(%)	28.571	0.238	0.967	1.760	0.714	0.491	0.491	0.173	0.173	19.206	15.397	0.173	1.602	0.173	0.173	0.173

Table 5. Continued.

2018	mBetween(%)	29.762	0.238	0.967	1.919	0.238	0.173	0.967	0.173	0.173	0.173	0.173	0.411	0.173	0.173	0.173	0.173	
2019	mBetween(%)	28.810	0.238	1.190	1.667	0.476	0.238	0.476	0.238	0.238	0.238	0.238	0.238	0.238	0.238	0.238	0.238	0.238
2020	mBetween(%)	26.905	0.238	1.205	1.681	0.967	0.253	0.491	0.491	0.808	0.253	0.253	1.270	0.491	0.491	0.491	0.491	0.491
2021	mBetween(%)	22.698	0.311	0.723	0.723	0.246	0.437	0.485	0.485	0.754	0.246	0.246	2.216	0.437	0.437	0.437	0.437	0.437

that of Tongling dropped sharply in 2016 and lost its central position.

(3) Proximity centrality

As shown in Fig. 2b), proximity centrality can be used to judge whether each city in Anhui Province is easily associated with other cities. In 2021, the average proximity centrality of 16 cities in Anhui Province was 63.89, among which the proximity centrality of Hefei, Ma’anshan, and Wuhu was higher than the average. This suggests that it is simple to connect these cities to other cities in the LCEY spatial network and cannot be easily controlled by other cities. These cities have a strong ability to gather resources, sophisticated transportation networks, and a high degree of economic growth. The low proximity centrality in other cities indicates that the regional gap in LCEY in Anhui Province is too large, which limits the LCEY in Anhui Province. Compared to the change in proximity centrality between 2012 and 2021, except for Tongling, which declined in 2015, the other cities changed little and increased steadily, indicating that the LCEY in Anhui Province was stable.

(4) Intermediate centrality

Fig. 2c) shows that the average intermediate centrality in 2021 is 4.29, and the intermediate centrality of Hefei, Ma’anshan, and Wuhu is higher than the mean. This indicates that these cities have strong control over resources related to logistics carbon emission efficiency and play the roles of intermediaries and bridges. The intermediate centrality of other cities is far lower than the average, and the spatial correlation between cities is not close, which is in a “dominated” role. Compared with the change in intermediate centrality between 2012 and 2021, except for Tongling, which declined in 2015, there was no obvious difference in other cities, and it steadily increased.

Generally, the Matthew effect of the LCEY in Anhui Province is obvious and is characterized by a serious range and uneven spatial distribution. From 2012 to 2021, Hefei, Ma’anshan, and Wuhu ranked first, indicating that these three cities were at the center of Anhui Province, controlling most resources and having a linkage effect with other cities. Affected by geographical conditions, economic levels, and other factors, other cities have a relatively weak attraction to the resource elements of the LCEY. They are frequently passive members of the spatial network who are unable to establish friendly interactions with other cities, especially in northern Anhui.

Block Module and Clustering Feature Analysis

Cluster analysis and plate division were performed using a block model analysis (Fig. 3 and Table 6).

According to the division results of Fig. 3, in 2012, Hefei, Ma’anshan, Wuhu, and Tongling belonged to the first plate; Huaibei and Huainan belonged to the second plate; Bozhou, Suzhou, Bengbu, and Fuyang belonged to the third plate; and Chuzhou, Lu’an, Xuancheng,

