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[Yuyue Xu](#)* and Xin Li

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Article

Research on the Impact of New Quality Productivity on the Resilience of Energy Industry Supply Chain—Empirical Evidence from China's Provincial Panel Data

Yuyue Xu ^{1,2,*} and Xin Li ³

¹ School of Economics and Management, Southwest Petroleum University, Chengdu 610500, China

² China-Iraq Belt and Road Joint Laboratory on Oil & Gas Energy, Chengdu 610500, China

³ School of Management Sciences, Chengdu University of Technology, Chengdu 610500, China

* Correspondence: 851284315@qq.com; Tel.: 18280309634

Abstract

Energy security and stability are related to the national economy and people's livelihood. Effectively improving the resilience of energy industry supply chain is the key factor to realize energy security and stability. Based on the panel data of 30 provinces in China from 2013 to 2023, this paper empirically explores the influence mechanism of new quality productivity on the resilience of energy industry supply chain. It is found that the new quality productivity can significantly enhance the resilience of the energy industry supply chain. The regulatory effect test results show that the energy supply level can positively promote the new quality productivity and improve the resilience of the energy industry supply chain. The results of threshold effect test show that the level of urbanization development and the degree of industrial synergy agglomeration make the new quality productivity have a nonlinear influence on the resilience of energy industry supply chain. The results of heterogeneity test show that the productivity of new quality plays a more obvious role in promoting the resilience of energy industry supply chain in the central region and the regions with higher information level.

Keywords: new quality productivity; energy industry; supply chain resilience

1. Introduction

The report of the 20th National Congress of the Communist Party of China clearly stated that "efforts should be made to improve the resilience and safety level of industrial and supply chains"[1]. With the reorganization of the global industrial supply chain, major changes unseen in a century are accelerating, and China's energy industry supply chain is also facing an uncertain external environment. The resilience of the energy industry supply chain is an important foundation for the smooth circulation of the national economy. Ensuring the safety and reliability of the energy industry supply chain is a key element in achieving high-quality development in China's socialist modernization drive. With the current reorganization of the industrial technology industry, "stuck neck" areas have posed serious threats to China's energy industry supply chain, such as "broken chains" and "stuck chains". This has triggered rising energy prices and energy supply crises. Large-scale enterprises have relocated from China, and the layout of the global industrial supply chain has gradually become localized and decentralized. On January 31, 2024, General Secretary Xi Jinping pointed out when presiding over the 11th collective study session of the Political Bureau of the CPC Central Committee that we should take new quality productivity as the foothold, actively develop and layout the industrial chain, and continuously improve the safety and stability of the industrial supply chain resilience [2]. New productivity is a highly dynamic advanced quality, a new kinetic

energy for economic development, and the driving force for replenishing and strengthening chains [3,4]. Under the new quality, developing new technologies, allocating new production factors, deepening the transformation of strategic industries, and deploying future emerging industries are important paths to building a modern industrial system and building a new development pattern [5–7]. Therefore, clarifying the interaction between new quality productivity and the energy industry supply chain, and exploring how to influence the resilience of the energy industry supply chain through new quality productivity, are currently urgent tasks for China to respond to challenges and achieve high-quality development of the industry.

In the previous literature, research on new productivity and industrial supply chain resilience was mainly carried out from two dimensions: First, focusing on enterprise development, exploring the path to improve the resilience of industrial supply chains empowered by new quality productivity in the process of enterprise development. Zhang and Zhang(2025) [8] Based on panel data of listed companies, the study concluded that new quality productivity can enhance the resilience of industrial supply chains. Akhtar et al.(2022) [9]、Zhao et al.(2023) [10] Based on enterprise panel data, we explore that supply chain digitalization has a significant positive impact on supply chain resilience and performance. Second, The enabling role of productivity in industrial supply chains. Lin et al.(2025)[11] Based on provincial panel data to explore the relationship between new quality productivity and high-quality agricultural development, focusing on "how new quality productivity affects high-quality agricultural development" and heterogeneity and non-linear characteristics, empirical analysis shows that new quality productivity significantly promotes high-quality agricultural development. Chen and Ye(2025) [12] Taking the sports industry as the research object, based on panel data from 11 provinces and cities, it was found that new productivity has a significant positive impact on the quality and efficiency reform of the sports industry and that there is regional heterogeneity.

Through sorting out the above studies, it can be found that most scholars have conducted extensive research on new productivity and industrial supply chain resilience. However, when studying the role of new quality productivity in industrial supply chain resilience, enterprise panel data are mostly used, and relevant research is still lacking. There is even less research on the impact of new quality productivity on the resilience of the energy industry supply chain. Based on existing research, this article focuses on the two key points of new quality productivity and energy industry supply chain resilience, and deeply explores the impact of new productivity on the resilience of the energy industry supply chain. Possible research contributions are: on the one hand, based on China's provincial panel data and from the perspective of the energy industry, we construct an index system for new quality productivity and energy industry supply chain resilience, and empirically analyze the impact mechanism of new quality productivity on the energy industry supply chain. This provides a basis for the construction of indicators for new quality productivity and energy industry supply chain resilience, and expands new ideas for the role of new productivity in energy industry supply chain resilience. On the other hand, based on theoretical analysis and from the perspective of industrial collaborative agglomeration and urbanization development level, the nonlinear impact characteristics of new quality productivity on the resilience process of the energy industry supply chain are explored, providing new inspiration for the implementation of government decision-making.

2. Theoretical Analysis and Research Hypotheses

2.1. *The Impact of New Quality Productivity on the Resilience of the Energy Industry Supply Chain*

New quality productivity is the driving force behind improving the resilience of the energy industry supply chain. According to existing research results, industrial supply chain resilience is generally represented by resilience and recovery capabilities. Resistance is represented by human capital, innovation output and input, etc. Resilience is represented by industrial benefits, degree of digitalization, etc. [13]. First of all, new quality productivity can help improve workers' quality and

thus enhance their resilience [14]. In order to meet the needs of the new situation, on the one hand, new quality productivity can cultivate professional talents that meet the needs of the energy industry, and on the other hand, it can attract high-quality, high-tech and new-quality labor force. Secondly, on the supply side, new quality productivity will help improve the advancement of labor materials in the energy industry, thereby improving recovery capabilities [15]. New quality productivity constantly updates and iterates new technologies and products in aspects such as digital and physical integration and artificial intelligence to enhance the advancement of labor materials in the energy industry, break through technical barriers, and ensure the stability of the supply chain. New quality productivity can gather workers, labor materials and labor objects, innovate and optimize the allocation structure among production factors, improve resilience and recovery capabilities, thereby promoting the improvement of the supply resilience of the energy industry. Based on this, the following hypothesis is put forward:

Hypothesis 1: *New quality productivity can promote the resilience of the energy industry supply chain*

2.2. The Moderating Effect of New Quality Productivity on the Resilience of Energy Industry Supply Chain

The regulating effect of energy supply level. The fundamental to improving the resilience of the energy industry supply chain is to solve the energy supply problem. By improving the energy supply level, we can realize the chain replenishment, strengthening the chain, cooperating and controlling the chain, thereby effectively promoting the positive impact of new quality productivity on the resilience of the energy industry supply chain [16]. With the geopolitical impact of the Russia-Ukraine conflict and the Palestine-Israel conflict, the global energy supply and demand pattern has experienced significant fluctuations. The improvement of energy supply level can realize innovative allocation and optimal combination of production factors, and then help new productivity to improve the resilience of the energy industry supply chain through new qualities such as technological revolution, management innovation, and product upgrades. Based on this, the following hypotheses are put forward:

Hypothesis 2: *The level of energy supply can enhance the role of new quality productivity in promoting the resilience of the energy industry supply chain*

2.3. The Threshold Effect of New Quality Productivity on the Resilience of Energy Industry Supply Chain

The threshold effect of the degree of industrial synergy and agglomeration. Different degrees of industrial synergy and agglomeration make new quality productivity forces important to energy industry supply chain resilience makes a difference [17]. When the degree of industrial synergy and agglomeration is high, it can promote the flow of production factors to break regional restrictions, improve the quality of labor matching, realize information and technology sharing, and broaden the overall layout of the industrial chain. When the degree of industrial synergy and agglomeration is low, the development potential of new quality productivity forces is not fully realized, and its role in promoting the resilience of the energy industry supply chain is weakened.

Threshold effect of urbanization development level. The development degree of urbanization in different regions has different degrees of influence on the supply chain of new quality production [18]. The level of urbanization development can maximize the role of new quality productivity. When the level of urban development is high, it can accelerate the development of new quality productivity by stimulating the creation of new labor materials, new workers, and new labor objects in the energy industry, thus promoting the improvement of the resilience of the regional energy industry supply chain. When the level of urban development is low, it will have varying degrees of impact on the resilience of the energy industry supply chain in the area where new quality productivity plays a role. Based on the above analysis, this article puts forward the following hypotheses:

Hypothesis 3: The degree of industrial collaborative agglomeration and the level of urbanization development cause new quality productivity to have a non-linear impact on the resilience of the energy industry supply chain.

3. Research Design

3.1. Research Methods

In order to scientifically and reasonably measure the level of new quality productivity and energy industry supply chain resilience, this article uses the entropy weight method to conduct an objective weighting analysis of all levels of indicators of new quality productivity and energy industry supply chain resilience to ensure the objectivity of the indicators. The main steps of entropy method weighting are as follows:

Index standardization. In order to ensure the dimensionality and direction consistency of the indicator data, the data must be standardized before the entropy weight method. At the same time, considering that the data in this article uses provincial panel data, the time variable t is introduced for measurement during indicator analysis. The specific standardization announcement is as follows:

$$\text{positive indicator : } \gamma_{ijt} = \frac{x_{ijt} - \min(x_j)}{\max(x_j) - \min(x_j)} \quad (1)$$

$$\text{negative indicator : } \gamma_{ijt} = \frac{\max(x_j) - x_{ijt}}{\max(x_j) - \min(x_j)} \quad (2)$$

Among them, i in the formula represents the province and city ($i = 1, 2, 3, \dots, n$); j represents each indicator ($j = 1, 2, 3, \dots, m$); t represents the year ($t = 1, 2, 3, \dots, 5$); γ_{ijt} represents the dimensionless data of j indicators in i province and city in t year after standardization; $\max(x_j)$ represents the maximum value of the j indicator in each province and city; $\min(x_j)$ represents the minimum value of the j indicator in the province and city. If the indicator positively affects the entropy value, formula (1) is used to calculate it; if the indicator negatively affects the entropy value, formula (2) is used to calculate it.

Indicator translation. Considering that 0 values will appear when calculating the data range dimension method, in order to ensure the operability of the data when calculating the entropy weight method, the standardized index values are slightly shifted here. The specific shift calculation is as follows: Formula (3)

$$\eta_{ijt} = \gamma_{ijt} + \alpha \quad (3)$$

In the formula, α is the translation amount. To ensure that the translation has little impact on the actual indicator data, the value of α here is 0.0001

Calculate the proportion of each indicator η_{ijt} in the new productivity and energy industry supply chain resilience indicators respectively. The specific calculation is as formula (4).

$$\mathcal{P}_{ijt} = \frac{\eta_{ijt}}{\sum_{t=1}^T \sum_{i=1}^n \eta_{ijt}} \quad (4)$$

Calculate the information entropy value e_j of the indicator j . The specific calculation is as shown in formula (5).

$$e_j = -k \sum_{t=1}^T \sum_{i=1}^n \mathcal{P}_{ijt} \ln \mathcal{P}_{ijt} \quad (5)$$

In the formula, $k = 1 / \ln nT$.

Calculate the weight w_j of each indicator j . The specific calculation is as shown in formula (6).

$$w_j = \frac{1 - e_j}{\sum_1^m (1 - e_j)} \quad (6)$$

In the formula, $1 - e_j$ is the difference coefficient of each indicator j . The greater the difference coefficient, the more important the indicator is in the comprehensive evaluation index system.

Calculate the linear weighted comprehensive index w_{ij} . The specific calculation is as follows: formula (7).

$$w_{ij} = \sum_{j=1}^m w_j \times \gamma_{ijt} \quad (7)$$

3.2. Model Settings

3.2.1. Baseline Regression Model

In order to test hypothesis 1, the following benchmark regression model is constructed :

$$Re_{it} = \alpha_0 + \beta_1 NQPF_{it} + \beta_2 Pop_{it} + \beta_3 Open_{it} + \beta_4 IS_{it} + \mu_i + \delta_t + \varepsilon_{it} \quad (8)$$

Among them, i and t represent the region and year respectively; Re_{it} is the core explained variable, indicating the energy industry supply chain resilience of the i region in the t period; $NQPF_{it}$ is the core explanatory variable, indicating the new quality productivity of the i region in the t period. The control variable Pop_{it} represents population density, $Open_{it}$ represents the degree of openness to the outside world, IS_{it} represents production industry structure; μ_i , δ_t , ε_{it} represent region fixed effects, time fixed effects and random disturbance terms respectively; α_0 is the intercept term, β_1 , β_2 , β_3 , β_4 are all coefficients to be estimated.

3.2.2. Moderating Effect Model

In order to deeply explore the impact of new productivity on the resilience of the energy industry supply chain and verify Hypothesis 2, considering that the energy supply level may have endogenous problems with the energy industry supply chain resilience, the energy supply level and the energy industry supply chain resilience are first decentralized, and the interaction term is included in the regression model. Construct the following moderating effect model :

$$Re_{it} = \alpha_1 + \beta_5 NQPF_{it} + \beta_6 ESL_{it} + \beta_7 C_NQPF_{it} \times C_ESL_{it} + \beta_8 X_{it} + \mu_i + \delta_t + \varepsilon_{it} \quad (9)$$

Among them, ESL_{it} is the regulating variable, representing the energy supply level respectively; $C_NQPF_{it} \times C_ESL_{it}$ represents the interaction term between decentralized new productivity and the resilience of the energy industry supply chain. The meanings of the remaining variables are the same as in equation (8).

3.2.3. Threshold Effect Model

In order to test Hypothesis 3 and explore the nonlinear characteristics of new productivity on the resilience of the energy industry supply chain, the following threshold effect model was constructed :

$$Re_{it} = \theta_0 + \beta_9 NQPF_{it} \times I(DIA_{it} \leq \xi_1) + \beta_{10} NQPF_{it} \times I(DIA_{it} > \xi_1) + \beta_{11} X_{it} + \mu_i + \delta_t + \varepsilon_{it} \quad (10)$$

$$Re_{it} = \theta_0 + \beta_{12} NQPF_{it} \times I(UI_{it} \leq \xi_2) + \beta_{13} NQPF_{it} \times I(UI_{it} > \xi_2) + \beta_{14} X_{it} + \mu_i + \delta_t + \varepsilon_{it} \quad (11)$$

Among them, θ_0 and θ_1 are intercept terms; DIA_{it} represents the degree of industrial collaborative agglomeration, UI_{it} represents the level of urbanization development; $I(\cdot)$ is an indicative function, which takes the value 1 when the conditions are met, and 0 otherwise; ξ is the corresponding threshold value; the meanings of the remaining variables are the same as equation (8).

3.3. Variable Definition and Description

3.3.1. Explained Variable

The explained variable in this article is the energy industry supply chain resilience (Re_{it}). The resilience of the energy industry supply chain refers to the ability of the energy industry to avoid chain disconnection and quickly return to its original state after being impacted. Considering the particularity of the energy industry, this article refers to relevant research [19–21] to construct the

following energy industry supply chain resilience index system (see Table 1), and uses the entropy weight method to complete the weight calculation of the index system.

Table 1. Energy industry supply chain resilience index system.

First-level indicator	Secondary indicator	Third-level indicator	Indicator Properties	Weight
Resistance ability	Energy processing and conversion losses	The amount of energy input during energy processing and conversion – the amount of energy output	Negative	0.0005
	Terminal energy consumption	The sum of all types of energy used for consumption (rather than for processing and conversion into other energy sources)	Positive	0.0297
	Energy production capacity	Primary energy production	Negative	0.0049
		Installed power generation capacity by region	Positive	0.0301
	Energy supply capacity	Energy production/terminal energy consumption	Positive	0.1009
Recovery ability	Energy and Environmental Governance	Comprehensive utilization of industrial solid waste	Positive	0.0386
		Industrial waste gas treatment investment costs	Positive	0.0639
	Energy industry foundation	Number of energy production and supply companies	Positive	0.0383
		Highway mileage	Positive	0.0274
		Electricity consumption of the whole society	Positive	0.0372
Reorganization ability	Market environment level	Residential water, electricity and fuel price index	Positive	0.0102
	Cyclical sustainable development	CO2 emission intensity	Negative	0.0070
	Industrial cooperation capabilities	Increment of tertiary industry/increment of GDP	Positive	0.0167
	Industrial synergy capabilities	Operating income of electronic information manufacturing owners	Positive	0.1312
Innovation ability	Innovation output	Number of valid patents	Positive	0.1017
		Purchase expenses abroad	Positive	0.1483
	Innovation investment	Purchase of domestic technology expenses	Positive	0.1581
		Technical transformation expenditure	Positive	0.0553

3.3.2. Core Explanatory Variables

New quality productivity ($NQPF_{it}$) is the core explanatory variable of this article. This article refers to existing research [22–24], combines the characteristics of the energy industry, from the three dimensions of new energy workers, new energy labor objects, and new energy labor materials, a new quality productivity index system was constructed (see Table 2), and the entropy weight method was used to calculate the weight of the index system.

Table 2. New quality productivity index system.

First-level indicator	Secondary indicator	Third-level indicator	Indicator Properties	Weight	First-level indicator	
New energy workers	Energy new quality worker capabilities	Energy industry labor productivity	Primary energy production per capita	Positive	0.1198	
		Energy industry workers' salary levels	Average salary of people employed in energy industries such as electricity, heat, gas and water production and supply industries	Positive	0.0251	
	Scale of energy new quality workers	Scale of human capital of industrial enterprises	Number of R&D personnel in industrial enterprises above designated size	Positive	0.0687	
		Energy industry labor force size	Proportion of employment in energy industries such as electricity, heat, gas and water production and supply industries	Positive	0.0162	
	Concept of new energy workers	Career preference	Scientific research and technical services industry employment share	Positive	0.0312	
		Entrepreneurial awareness	Number of energy-related companies incubated by technology business incubators	Positive	0.0596	
	Energy new quality labor object	Energy supply level	City gas supply level	Urban natural gas supply	Positive	0.0398
			Urban Liquefied Petroleum Gas Supply Level	Urban Liquefied Petroleum Gas Supply	Positive	0.0474
		Energy green development level	Green and clean energy power generation situation	Wind, nuclear, hydro, solar power generation	Positive	0.0354
			Industrial pollution control capabilities	Completed investment in industrial pollution control	Positive	0.0424
Development level of new quality enterprises		Enterprise informatization development level	Number of computers used by the unit enterprise at the end of the period	Positive	0.0281	
		Economic development of industrial enterprises above designated size	Operating income profit rate of industrial enterprises	Positive	0.0006	
		State-owned energy industry enterprise fixed investment capacity	Growth rate of fixed assets investment in state-owned economy, energy industry	Positive	0.0239	
Energy new quality labor data	Energy infrastructure	City gas supply capacity	City artificial gas, natural gas, liquefied petroleum gas pipeline length	Positive	0.0365	
		Power supply capacity	Circuit length of 35KV and above transmission lines in the region	Positive	0.0445	
	Digital infrastructure	Digital development level of industrial enterprises	Number of Internet broadband access users in industrial enterprises	Positive	0.0331	
		Industrial enterprise information technology service level	Industrial Enterprise Information Technology Services Revenue	Positive	0.1666	

Technological innovation construction	Industrial enterprise technology innovation achievement level	Number of valid invention patents of industrial enterprises	Positive	0.1025
	Industrial enterprise technology innovation construction level	New product development expenses for industrial enterprises	Positive	0.0786

3.3.3. Regulated Variable

This paper uses the energy supply level (ESL_{it}) as the adjusting variable. According to Gu et al. (2023) [25], the energy supply level is characterized by the length of transportation routes in different regions.

3.3.4. Threshold Variable

This article uses the industrial synergy agglomeration index (DIA_{it}) and urbanization development level (UI_{it}) as threshold variables. Referring to existing research [26–28], an improved E-G index is used to calculate the industrial synergy agglomeration index. The calculation announcement is as follows:

$$DIA_{it} = 1 - \frac{|HP_{it} - HM_{it}|}{HP_{it} + HM_{it}} + (HP_{it} + HM_{it}) \quad (12)$$

$$HP_{it} = \frac{r_{ip}/r_p}{r_i/r} \quad (13)$$

$$HM_{it} = \frac{r_{im}/r_m}{r_i/r} \quad (14)$$

Among them: DIA_{it} is the industrial synergy agglomeration index, HP_{it} and HM_{it} represent the production service industry and manufacturing respectively. the degree of industrial agglomeration; r_{ip} and r_{im} are the number of employees in the production service industry and manufacturing industry in the i region respectively; r_p and r are the total number of employees in the production service industry and manufacturing industry in the country; r is the total number of employment in the country; i and t represent the region and year respectively.

The transportation capacity of a city reflects the process of urbanization to a certain extent. Refer to the existing literature [29,30], This article characterizes the level of urbanization development by the number of public transportation vehicles per 10,000 people.

3.3.5. Control Variables

In order to more accurately reflect the relationship between new quality productivity and the resilience of the energy industry supply chain, this paper further introduces control variables: population density (Pop_{it}), represented by the ratio of the regional population at the end of the year to the regional area; degree of openness to the outside world ($Open_{it}$), represented by the ratio of import and export trade volume to regional GDP; industry Structure (IS_{it}) is characterized by the ratio of the gross production value of the secondary and tertiary industries to the gross regional product.

3.4. Data Source

The panel data of 30 provinces in China from 2013 to 2023 (considering the integrity and availability of data, the data of Xizang, Hong Kong, Macao and Taiwan have been excluded) are selected as the research samples. The data mainly comes from the "China Statistical Yearbook", "China Science and Technology Statistical Yearbook", "China Education Statistical Yearbook", "China High tech Industry Statistical Yearbook", "China Basic Unit Yearbook", "China Environment

Yearbook", "China Power Yearbook", "China Natural Resources Statistical Yearbook", "China Trade and Economic Statistical Yearbook", "China Economic and Trade Yearbook", "China Business Yearbook", "China Population and Employment Statistical Yearbook", "China Energy Statistical Yearbook", "China Industrial Statistical Yearbook", and CSMAR database. And use linear interpolation to fill in missing data.

4. Empirical Analysis

4.1. Benchmark Regression Results Analysis

The benchmark regression results are shown in Table 3, which mainly reports the impact of new quality productivity on the resilience of the energy industry supply chain. Column (1) is the regression result of adding only the explanatory variable new productivity and fixing the time and province effects. The regression coefficient is 0.463, passing the significance test at the 1% level. Columns (2) and (3) are the regression results after introducing control variables such as openness, population density, and industrial structure. Column (3) does not fix the time and province. The regression results show that new quality productivity has a significant role in improving the resilience of the energy industry supply chain. Hypothesis 1 is thus verified.

Table 3. Baseline regression results.

	(1)	(2)	(3)
	<i>Re</i>	<i>Re</i>	<i>Re</i>
<i>NQPF</i>	0.463*** (7.87)	0.330*** (5.91)	0.809*** (29.89)
<i>Open</i>		-0.193*** (-7.33)	0.052*** (4.38)
<i>Pop</i>		0.118* (1.95)	0.004** (2.06)
<i>IS</i>		-0.035** (-2.50)	0.050*** (6.47)
Constant term	0.069*** (8.23)	-0.483 (-1.46)	-0.060*** (-4.41)
Time fixed effects	Y	Y	N
Province fixed effects	Y	Y	N
N	330	330	330
R^2	0.926	0.940	0.820

Note: *, **, and *** indicate significance at the 10%, 5%, and 1% levels, respectively, and () represents the t-statistic. the same below.

4.2. Endogeneity Test

4.2.1. Instrumental Variable Method

Considering that there is a certain endogenous causal relationship between new productivity and the energy industry supply chain, there may be problems with omitted variables in the constructed model. This article further introduces instrumental variables and estimates them using two-stage least squares (2SLS) method. The instrumental variable is selected as the one-period lag of new quality production ($l.NQPF$). As can be seen from column (2) in Table 4, the selected instrumental variables pass the non-identifiable and weak instrumental variable tests, proving that the instrumental variables are effective. The regression results in column (1) show that the instrumental variable coefficient is significant at the 1% level, indicating that the instrumental variable is positively related to new qualitative productivity. The regression results in column (2) show that new productivity positively promotes the resilience of the energy industry chain and supply chain, verifying that the baseline regression results are robust.

4.2.2. Differential and System GMM Estimation Method

Considering that the resilience of the energy industry supply chain may be related to the sequence, this paper further introduces the lag term $l.Re$ of regional energy industry supply chain resilience in the model, in order to separate unobservable factors and reduce the bias of model estimation. Estimate and test the model using the GMM system differential method. According to column (3) of Table 4, AR (1) is significant at the 10% level, while AR (2) is not significant. If there is first-order autocorrelation in the residuals and no second-order autocorrelation, then there is no significant sequence correlation in the error term of the original model. If the P-value of Hansen's overidentification constraint test is greater than 0.1, the null hypothesis is accepted that the instrumental variable overidentification constraint is valid. After endogeneity treatment, the regression results in column (3) indicate that new quality productivity has a significant promoting effect on the resilience of the energy industry supply chain, further verifying hypothesis 1.

Table 4. Endogeneity test results.

	(1)	(2)	(3)
	One-stage regression	Two-stage regression difference	System GMM
<i>NQPF</i>		0.529*** (5.81)	0.179*** (9.51)
<i>l.NQPF</i>	0.647*** (15.55)		
<i>l.Re</i>			0.764*** (40.96)
Constant term	-0.474* (-1.80)		-0.024 (-1.48)
DWH inspection P-value		0.0001	
Cragg Donald Wald F-value		241.951	
Unrecognized inspection		145.476***	
AR (1) test P-value			0.095
AR (2) test P-value			0.249
Control variable	Y	Y	Y
Time fixed effects	Y	Y	Y
Province fixed effects	Y	Y	Y
N	300	300	300
R^2	0.973	0.326	

Note: The critical value of Cragg Donald Wald F is 10% and 16.38. The values in parentheses for columns (1) and (2) are t values, and the values in parentheses for column (3) are Z values.

4.3. Robustness Check

In order to verify the reliability of the empirical results, this paper conducts a robustness test on the regression results by replacing models, eliminating abnormal samples, replacing core explanatory variables, and replacing explained variables. (1) Replace the model. Due to the range of resilience values in the energy industry supply chain being between 0 and 1, which meets the condition of limited dependent variables, the Tobit model was chosen for re estimation. After adding control variables and fixed provinces and years, the regression results in column (1) of Table 5 show that new quality productivity significantly positively affects the resilience of the energy industry supply chain, which is consistent with the baseline regression results. (2) Remove abnormal samples. Due to the impact of the COVID-19 in 2020, the consumption demand has changed. Therefore, the regression estimation is conducted again after removing the data of the epidemic year. The regression results in Column 5 (2) of Table 5 show that they are still significant. (3) Replace the dependent variable and the core explanatory variable separately. Using principal component analysis, the weights of new quality productivity and energy industry supply chain resilience were re estimated, and the entropy

weight method was used to obtain the replaced new quality productivity weight $C.NQPF$ and energy industry supply chain resilience weight $c.Re$. After re estimating the regression model, it can be seen from columns (3) and (4) of Table 5 that the regression results are significant. Based on this, the conclusion that new quality productivity enhances the resilience of the energy industry supply chain is relatively robust.

Table 5. Mechanism Inspection Results.

	(1)	(2)	(3)	(4)	(5)
	Tobit model	Excluding samples from the year of the epidemic	Replace the core explanatory variable	Replace the dependent variable	Re
$NQPF$	0.542*** (9.98)	0.333*** (6.27)		3.750*** (7.88)	0.319*** (5.82)
$C.NQPF$			0.0360*** (10.09)		
$NQPF \times ESL$					0.162*** (3.78)
Control variable	Y	Y	Y	Y	Y
Constant term	-0.0632 (-1.60)	-0.421 (-1.33)	0.110 (0.35)	-9.146*** (-3.24)	-0.356 (-0.97)
Time fixed effects	Y	Y	Y	Y	Y
Province fixed effects	Y	Y	Y	Y	Y
N	330	300	330	330	330
R^2	-	0.948	0.950	0.974	0.943

Note: Column (1) parentheses represent Z values, while the remaining parentheses represent t values.

4.4. Mechanism Test

4.4.1. Moderating Effect

According to equation (9), the regression results using energy supply level as the moderating effect are shown in column (5) of Table 5. It can be seen from the regression results that the regression coefficient of the interaction term between energy supply level and new quality productivity is 0.162, which is significant at the 1% level. It shows that the level of energy supply can effectively enhance the positive enabling effect of new productivity on the resilience of the regional energy industry supply chain. Therefore, hypothesis 2 is confirmed.

4.4.2. Threshold Effect

According to equations (10) and (11), this paper tests the threshold effect of the level of urbanization development and the degree of industrial collaborative agglomeration. After testing, as shown in Figure 1, both the urbanization level and the degree of industrial collaborative agglomeration passed the single threshold test. According to the threshold effect test results in Table 6, it can be seen that the level of urbanization and the degree of industrial collaborative agglomeration are not significant in the double threshold test. This shows that there is a nonlinear relationship between new quality productivity and the resilience of the energy industry supply chain, with the level of urbanization development and the degree of industrial synergy agglomeration as threshold variables. According to column (1) in Table 7, it can be seen that the first threshold value of urbanization level is 11.93. Among the threshold values of urbanization level at each stage, new productivity is significant at the 1% level. However, the coefficient of new quality productivity is different at different stages. As the level of urban development continues to improve, new production factors will rapidly upgrade, making new quality productivity more dependent on new production factors and weakening the positive promotion effect on urban development level. According to column (2) in Table 7, it can be seen that the first threshold value of industrial collaborative

agglomeration is 3.098. When the degree of industrial collaborative agglomeration is less than the threshold value of 3.098, the regression coefficient of new quality productivity is 0.286, which is significant at the 1% level; when the degree of industrial collaborative agglomeration is greater than the threshold value of 3.098, the regression coefficient of new quality productivity is 0.529, which is significant at the 1% level. This shows that as industrial synergy agglomeration continues to grow, the role of new quality productivity in improving the resilience of the energy industry supply chain is also increasing. Taken together, under different levels of urbanization development and industrial synergy agglomeration, new quality productivity shows non-linear characteristics to the energy industry supply chain. And it shows significant promotion effect in different threshold ranges. Therefore, hypothesis 3 is confirmed.

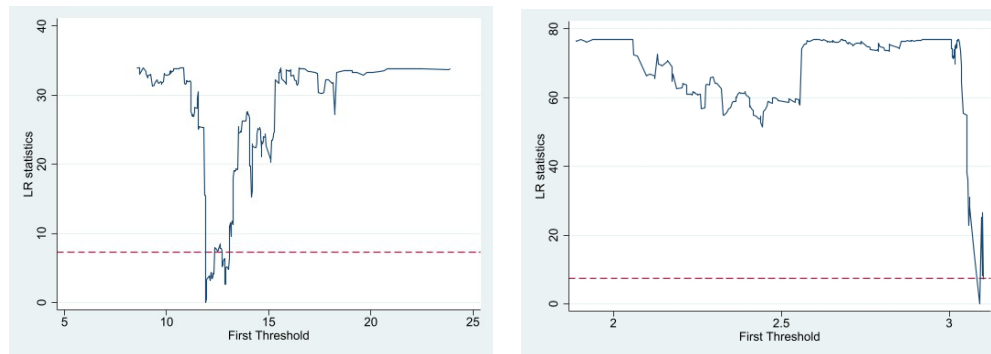


Figure 1. Threshold variable likelihood ratio function graph.

Table 6. Threshold effect test.

Threshold variable	Threshold effect	F value	P value	BS frequency	Threshold value			Threshold inspection threshold type	Threshold value	95% confidence interval
					10%	5%	1%			
Urbanization development level	Single threshold	34.46	0.020	500	21.326	26.115	39.614	Single threshold	11.9300	[11.9100,11.9500]
	Double threshold	16.01	0.200	500	21.082	25.795	40.767		-	-
Industrial synergy agglomeration	Single threshold	78.21	0.000	500	26.929	34.291	45.154	Single threshold	3.0918	[3.0739,3.0991]
	Double threshold	20.81	0.140	500	23.113	51.424	112.337		-	-

Table 7. Threshold effect test results.

	(1)	(2)
	<i>Re</i>	<i>Re</i>
<i>NQPF</i> ($UI \leq 11.9300$)	0.553*** (10.19)	
<i>NQPF</i> ($11.9300 < UI$)	0.415*** (8.74)	
<i>NQPF</i> ($DIA \leq 3.0918$)		0.286*** (6.12)
<i>NQPF</i> ($3.0918 < DIA$)		0.529*** (11.23)
Constant term	-0.481 (-1.39)	-0.531 (-1.63)
Control variable	Y	Y
Time fixed effects	Y	Y
Province fixed effects	Y	Y
N	330	330
R^2	0.456	0.516

4.5. Heterogeneity Analysis

In order to verify the differences in the resilience of new quality productivity to the energy industry supply chain among different informatization levels and different regions. This article uses the ratio of the total volume of postal and telecommunications services in a region to the year-end population to characterize the level of informatization; According to the geographical location division of the National Bureau of Statistics, the sample is divided into two parts: the eastern and western parts and the central part for separate regression analysis. (1) Heterogeneity in the degree of informatization. Taking the median of the informatization degree ratio as a reference, it is divided into areas with a high degree of informatization and areas with a low degree of informatization. It can be seen from columns (1) and (2) in Table 8 that the new quality productivity coefficient of areas with a high degree of informatization is 0.511, and that of areas with a low degree of informatization is 0.203, and both are significant at the 1% level. The test results for the difference in coefficients between groups are significant at the 10% level, indicating that a high degree of informatization can help promote new quality productivity and improve the resilience of the energy industry supply chain. (2) Regional heterogeneity. It can be seen from columns (3) and (4) of Table 8 that the new quality productivity coefficient in the eastern and western regions is 0.306, and the new quality productivity coefficient in the central region is 0.711, both are significant at the 1% level, and the difference test results of the coefficients between groups are significant at the 5% level. The economy of the eastern region is developing rapidly, and the impact of new productivity on it is smaller than that of other regions. The western region is a resource-based agglomeration area, and the energy industry supply chain has relatively good resilience. Compared with other regions, the energy industry supply chain resilience is less affected by new quality productivity. Therefore, the central region is more susceptible to the impact of new quality productivity than the eastern and western regions, thereby improving the resilience of the energy industry supply chain.

Table 8. Heterogeneity test results.

	(1)	(2)	(3)	(4)
	High degree of informatization	Low level of informatization	East West Department	Central
<i>NQPF</i>	0.511*** (5.53)	0.203*** (3.66)	0.306*** (4.46)	0.711*** (3.11)
Constant term	-1.132 (-1.33)	-0.408** (-2.09)	-1.149* (-1.96)	0.327 (0.59)
Control variable	Y	Y	Y	Y
Time fixed effects	Y	Y	Y	Y
Province fixed effects	Y	Y	Y	Y
P-Value	0.060*		0.040**	
N	165	163	231	99
R^2	0.951	0.968	0.948	0.937

Note: P-Value value is the significance of the coefficient difference between the test groups, obtained through Fisher's test sampling 1000 times.

5. Conclusion and Recommendations

5.1. Conclusion

This article uses panel data from 30 provinces in China from 2013 to 2023 as a research sample, using a two-way fixed effect model, an adjustment variable model and a threshold effect model to deeply explore the impact of new quality productivity on the resilience of the energy industry supply chain. The study found that: first, new quality productivity has a positive role in improving the resilience of the energy industry supply chain. Second, in the impact of new quality productivity on

the resilience of the energy industry supply chain, the level of urbanization and the degree of industrial synergy agglomeration both have a single threshold effect and exhibit non-linear impact characteristics. Third, energy supply levels can enhance the positive impact of new quality productivity on the resilience of the energy industry supply chain. Fourth, the role of new quality productivity in promoting the resilience of the energy industry supply chain is more significant in areas with a high degree of informatization and central regions.

5.2. Suggestion

Based on the above conclusions, the following suggestions are put forward: (1) The development of new productivity in the energy industry should be based on technological innovation and increase investment in research and development funds, vigorously improve the level of information digitization, through the continuous updating of key technologies, the continuous iterative upgrading of new technologies and new products, to provide a steady stream of momentum for strengthening and replenishing chains, thereby empowering the stable development of the regional energy industry supply chain. (2) Accelerate the regional energy market reform, focus on industrial synergy and agglomeration, optimize the allocation of energy industry resources, and enhance the flexibility of the regional energy industry supply chain. Further improve regional energy supply levels, coordinate the layout of energy industry gradients, strengthen infrastructure construction, and optimize industrial upgrading. Based on new production factors, build a digital and intelligent development path for the energy industry and enhance the resilience of the energy industry supply chain. (3) Further develop informatization construction, actively promote the reform of medical and educational systems and mechanisms, and improve human capital levels. Vigorously develop advantageous industries, support the development of high-tech industrial enterprises such as artificial intelligence, big data, cloud computing and new energy, improve labor productivity, continue to empower new productive forces, and continuously improve the resilience of industrial supply chains.

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