

# Study on Logistics Efficiency of Key Provinces and Cities along the "Belt and Road"

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**Abstract.** According to the "Belt and Road" initiative, the logistics efficiency determine the operating costs of logistics directly, and indirectly influence the competitiveness of China's export commodities internationally. In this paper,DEA method is used to study the logistics efficiency of eighteen key provinces in China. And the result shows that the overall average of logistics efficiency to some provinces is relatively low, except other five provinces and cities, which can provide policy guidance for the "Belt and Road" development strategy.

#### 1. Introduction

Chinese President Xi Jinping proposed "One Belt One Road" strategy in 2013, which can construct the muti-layered and compound network of sea and land with many continents-Asia, Africa and Europe, and promote cooperation, communication and development among countries and regions along the route<sup>[1,2]</sup>. And China involves eighteen provinces, cities and municipalities. The logistics efficiency directly influences logistic cost. In order to make full use of the relative advantages of each region and actively integrate into the "One Belt One Road" significant development strategy, higher requirements have been put forward for the provinces and cities along the line.

## 2. Evaluation methodology

Most computational methods of studying logistics efficiency are based on the ratio between input and output of materials at present<sup>[3,4]</sup>. And there are some methods such as ABC cost method, index tree method, AHP method, DEA method commonly used to evaluate the logistics efficiency<sup>[3-11]</sup>. But ABC only applies to the influential analysis of single variable, index tree method has poor integrality and hardly reflects on the changes of the logistics efficiency in general. And AHP method needs a few quantitative information, has strong subjectivity, which results tend to be less objective. However, DEA method rightly avoids subjectivity, simplifies the calculation and reduces errors. Accordingly, DEA method is used as a research method in this paper. The specific information are as follows.

## 2.1 CCR model

It is supposed to have n decision-making units(DMU) which are comparable and belong to one type. And each DMU has m input and s output. Then we can get the model as follows by using linear programming and duality transformation methods.

$$D(\varepsilon) = \min \left[ \theta - \varepsilon \left( e^{-}S^{-} + e^{+}S^{+} \right) \right]$$



$$S.t. \begin{cases} \sum_{j=1}^{n} X_{j} \lambda_{j} + S^{-} = \theta X_{0} \\ S.t. \begin{cases} \sum_{j=1}^{n} Y_{j} \lambda_{j} - S^{+} = Y_{0} \\ \lambda_{j} \geq 0, \ j = 1, 2...n \\ S^{+} = (s_{1}^{+}, s_{2}^{+} ... s_{s}^{+})^{T} \geq 0, S^{-} = (s_{1}^{-}, s_{2}^{-} ... s_{m}^{-})^{T} \geq 0 \end{cases}$$

$$(2-1)$$

Where  $\theta$ ,  $\lambda_j$ , j=1,2,...,n are all dual variables, m is the unit vector.  $\mathbf{e}^- = (1,1,...,1) \in E_m$ , s is the unit vector  $\mathbf{e}^+ = (1,1,...,1) \in E_s$ . s+ and s- are slack variables,  $X_j = (x_{1j},x_{2j}...x_{mj})^T$ ,  $Y_j = (y_{1j},y_{2j}...y_{sj})^T$ ,  $X_0 = (x_{ij_0},x_{2j_0}...x_{mj_0})^T$ ,  $Y_0 = (y_{1j_0},y_{2j_0}...y_{sj_0})^T$ .

## 2.2 BCC model

CCR model exists certain limitation, because it is based on the assumed condition which is constant returns to scale to evaluate the comparative efficiency [5-7]. Thus, a new BCC model is constructed as follows by changing the hypothetical condition about constant returns to scale. It is assumed that constant returns to scale changes to be variable, and constraint condition  $\sum_{i=1}^{n} \lambda_i = 1$  is adde, we can get

the input-oriented BCC model<sup>[5,6]</sup>:

$$V(\varepsilon) = \min \left[ \theta - \varepsilon (e^{-}S^{-} + e^{+}S^{+}) \right]$$

$$\sum_{j=1}^{n} X_{j} \lambda_{j} + S^{-} = \theta X_{0}$$

$$\sum_{j=1}^{n} Y_{j} \lambda_{j} - S^{+} = Y_{0}$$

$$S.t. \begin{cases} \sum_{j=1}^{n} \lambda_{j} = 1 \\ \lambda_{j} \ge 0, j = 1, 2...n \\ S^{+} = (s_{1}^{+}, s_{2}^{+} ... s_{s}^{+})^{T} \ge 0, S^{-} = (s_{1}^{-}, s_{2}^{-} ... s_{m}^{-})^{T} \ge 0 \end{cases}$$

$$(2-2)^{N}$$

Where  $\lambda^0$ ,  $S^{0-}$ ,  $S^{0+}$ ,  $\theta^0$  are the optimum value. If  $\theta^0 = 1$ , and  $S^{0-} = S^{0+} = 0$ , DMU is valid both on technique and scale; else if  $\theta^0 = 1$ ,  $S^{0-} = 0$ , or  $S^{0+} = 0$ , DMU is weakly efficient, but not valid on technique and scale at the same time; else if  $\theta^0 < 1$ , DMU is the non-efficient unit.

## 3. Empirical research

The paper has selected the indexes include X1(Unit million)=Logistics Employees' number, X2(Unit billion yuan)=Logistics Employees' salary, X3(Unit billion yuan)=Logistics Investment, X4(Unit ten thousand kilometers)=Grade Highway Mileage, Y1(Unit million tons)=Freight Turnover, Y2(Unit billion yuan)=Logistics Added Value, Y3(Unit billion yuan)=GDP, etc from various domestic and foreign literature<sup>[3-9]</sup>. And also chooses 18 key provinces and cities along the "Belt and Road" as DMU according to the indexes above. The relevant data mainly comes from China Statistic Yearbook and each province and cities Statistic Yearbook in 2016 published by National Bureau of Statistics, just as follows.

Table 3-1 Data of 18 Key Provinces along "One Belt One Road" in 2015

Province	Input Indexes				Output Indexes				
	X1	X2	X3	X4	Y1	Y2	Y3		
Sinkiang	16.7	133.9	1017.4	13.701	1772.94	536.06	9324.80		
Tsinghai	4.3	31.9	419.8	6.464	445.58	90.55	2417.05		
Shaanxi	28.0	177.6	1439.5	15.385	3263.52	713.02	18021.86		



Gansu	12.6	75.5	814.9	12.045	2225.81	274.65	6790.32
Ningxia	3.8	23.9	261.7	3.305	816.93	200.66	2911.77
Inner Mongolia	20.6	133.2	1251.2	16.377	4190.30	1087.32	17831.51
Heilongjiang	27.5	162.4	995.0	13.633	1545.35	707.03	15083.67
Jilin	16.6	97.0	961.3	9.009	1425.35	529.79	14063.13
Liaoning	36.1	236.6	1263.6	10.651	11711.92	1702.80	28669.02
Guangxi	20.0	124.0	1532.9	10.502	4061.82	803.10	16803.12
Yunnan	17.1	112.9	1815.2	19.707	1500.27	304.49	13619.17
Tibet	0.9	6.4	346.9	5.842	119.64	31.76	1026.39
Chongqing	27.1	162.6	1436.5	11.289	2709.53	761.31	15717.27
Shanghai	51.5	516.3	794.6	1.320	19495.88	1133.17	25123.45
Fujian	24.5	161.0	2230.7	8.749	5447.49	1547.30	25979.82
Zhejiang	32.0	239.8	2312.8	11.557	9869.72	1631.88	42886.49
Guangdong	82.8	669.4	3037.6	20.146	14882.21	2928.90	72812.55
Hainan	6.5	44.9	421.8	2.630	1181.73	187.80	3702.76

# 3.1 Calculation results

It is calculated by using MaxDEA7.0 software based on CCR model and BCC model, and the results just as following table 3-2.

Table 3-2 Evaluation Form of Logistics Efficiency of 18 Provinces and Cities along the "One Belt One Road" in 2015

No	Province	Comprehensive Efficiency	Pure Technical Efficiency	Scale Efficiency	Scale Merit	
1	Sinkiang	0.5931	0.6114	0.9701	Ascending	
2	Tsinghai	0.4237	0.6930	0.6114	Ascending	
3	Shaanxi	0.6414	0.6763	0.9483	Ascending	
4	Gansu	0.6339	0.6650	0.9533	Ascending	
5	Ningxia	0.9555	1	0.9555	Ascending	
6	Inner Mongolia	0.9781	0.9812	0.9968	Ascending	
7	Heilongjiang	0.7052	0.7502	0.9401	Ascending	
8	Jilin	0.8107	0.8674	0.9346	Ascending	
9	liaoning	1	1	1	Static	
10	Guangxi	0.8242	0.8330	0.9894	Ascending	
11	Yunnan	0.6745	0.6786	0.9940	Ascending	
12	Tibet	0.8967	1	0.8967	Ascending	
13	Chongqing	0.6267	0.6362	0.9849	Ascending	
14	Shanghai	1	1	1	Static	
15	Fujian	1	1	1	Static	
16	Zhejiang	1	1	1	Static	
17	Guangdong	1	1	1	Static	
18	Hainan	0.5825	1	0.5825	Ascending	



## 3.2 Results analysis

#### 3.2.1 Comprehensive efficiency analysis

According to table 3-2, there are five provinces' logistics comprehensive efficiency getting the optimum, they are Liaoning, Shanghai, Fujian, Zhejiang and Guangdong, indicating that the combination of input and output of logistics comprehensive efficiency has obtained the optimum, the logistic input has been made the greatest use and has received the best achievement which doesn't need any adjustment. In other 13 provinces like Heilongjiang, Inner Mongolia, Yunnan do not reach 1, meaning that they do not achieve the comprehensive efficiency.

## 3.2.2 Pure technical efficiency analysis

According to table 3-2,there are eight provinces' pure technical efficiency getting the optimum, they are Ningxia, Liaoning, Tibet, Shanghai, Fujian, Zhejiang, Guangdong and Hainan, indicating that their output in logistics compared with the input has reached the maximum, and the efficiency of logistic industry development is relatively high,but temporarily do not exist excess input or insufficient output. However, other three provinces' scale efficiency do not get the overall efficiency on the grounds that their production scale of logistic industry is small and is still in the stage of increasing returns to scale, they are Ningxia, Tibet and Hainan.

The pure technical efficiency of the rest ten provinces and cities has not reached the optimum, and even at the lower ranking provinces and cities, it has not reached 0.7, indicating that the logistic industry in these provinces is still using comparatively out-dated management styles, logistics technology, etc, and also has some deficiencies leading to the waste of logistics resources and lower logistics input efficiency. Because lower pure technical efficiency is the main reason for lower efficiency of logistics input and output, these provinces and cities need improvements in configuration and management of existing logistics resources, which have a great space to improve in efficiency.

## 3.2.3 Scale efficiency analysis

According to table 3-2, there are five provinces and cities- Liaoning, Shanghai, Fujian, Zhejiang, and Guangdong, their scale efficiency which equals 1 has not changed. And the scale efficiency of other provinces and cities is lower than 1, especially, the scale efficiency of Hainan (0.582) and Tsinghai (0.6114) is the lowest, far below the average scale efficiency. As a result, these two provinces are supposed to put more eyesight on adjusting the scale of logistics industry. And the scale merit of Liaoning, Shanghai, Fujian, Zhejiang and Guangdong has not changed, as to the remaining 13 provinces and cities, they are all at the stage of increasing returns to scale, which infers that it is advisable for these 13 provinces and cities to increase the input of logistics elements, to optimize the existing logistics modes, to strengthen the construction of logistics network system among regions, to promote the rapid and stable development of logistics industry and to enhance logistics scale efficiency.

## 3.3 Slack variables of input and output

In order to verify the validity of DEA calculation, we have calculated slack variables of logistics input and output of DEA invalid decision-making units. And results are obtained as the following table 3-3:

Table 3-3 Slack Variables of Logistics Input and Output Indexes

		Inputs Redundancy				Outputs Insufficient		
No	Province	$s_1^-$	$s_2^-$	$s_3^-$	$s_4^-$	<i>s</i> <sub>1</sub> <sup>+</sup>	<i>s</i> <sub>2</sub> <sup>+</sup>	<i>s</i> <sub>3</sub> <sup>+</sup>
1	Sinkiang	-6.79	-54.48	-413.94	-5.57	1006.61	0	0
2	Tsinghai	-2.48	-18.39	-241.95	-3.73	110.67	1.42	0
3	Shaanxi	-10.04	-63.69	-516.25	-5.52	1764.57	78.10	0
4	Gansu	-4.61	-27.64	-298.32	-4.41	0	63.07	0
5	Ningxia	-0.17	-1.06	-11.65	-0.15	77.81	0	459.97
6	Inner Mongolia	-0.45	-2.92	-27.39	-0.36	1287.43	0	447.64



7	Heilongjiang	-8.11	-47.87	-293.32	-4.02	3949.26	109.01	0
8	Jilin	-3.14	-18.37	-182.01	-1.71	1811.08	5.33	0
9	liaoning	0	0	0	0	0	0	0
10	Guangxi	-3.52	-21.80	-269.50	-1.85	0	0	0
11	Yunnan	-5.57	-36.75	-590.84	-6.41	1633.99	213.73	0
12	Tibet	-0.09	-0.66	-35.82	-0.60	116.57	7.30	0
13	Chongqing	-10.12	-60.71	-536.31	-4.21	1660.12	0	0
14	Shanghai	0	0	0	0	0	0	0
15	Fujian	0	0	0	0	0	0	0
16	Zhejiang	0	0	0	0	0	0	0
17	Guangdong	0	0	0	0	0	0	0
18	Hainan	-2.71	-18.74	-176.09	-1.10	0	0	142.67

PS:  $s_i^-$  (i=1,2,3,4) in this table refers to logistics employees,logistics employees' salary,logistics investment,and grade highway mileage; and  $s_i^+$  (j=1,2,3) refers to freight turnover,logistics value added and GDP.

Based on table 3-3, there are five provinces with valid DEA - Liaoning, Shanghai, Fujian, Zhejiang, and Guangdong, which inputs redundancy and outputs insufficient are zero and do not require adjustment. Other 13 provinces and cities with invalid DEA all have inputs redundancy and outputs insufficient. If they want to obtain the optimal efficiency, they have to adjust certain inputs and outputs, for instance, if Tsinghai wants to obtain the optimal logistics efficiency, under the existing circumstance of logistics resources input, it has to reduce logistics employees with 24.8 thousand people, logistics employees' salary with 1.838 billion yuan, logistics investment with 24.195 billion yuan and grade highway mileage with 37,300 kilometers, at the same time ,to increase freight turnover with 11.067 billion tons kilometers and logistics value added with 142 million yuan.

#### 4. Conclusion

Through the empirical research on logistics efficiency of key provinces and cities along the "Belt and Road", we can draw the conclusion. Firstly, the overall development and efficiency of logistics is relatively low. Secondly, optimal logistics output can be obtained by optimizing the combination of factors of production. Thirdly, it is essential for provinces and cities which lack of logistics efficiency to introduce the advanced logistics technology and to improve logistics management modes and infrastructures.

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