Research Article

An Empirical Analysis on the Scientific Operation Mode of Entrepreneurial Incubators in Higher Vocational Colleges

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1. Introduction

As an important platform for gathering regional creative resources, cultivating the development of small and micro-enterprises, and promoting regional economic growth, business incubators have increasingly received high attention from government departments and all sectors of society. Communities around the world are always looking for ways to encourage and support the development of new businesses to boost economic growth and create jobs. An entrepreneurial incubator is one of the important mechanisms of the entrepreneurial ecosystem. Since its inception decades ago, startup incubation has grown into an essential part of the entrepreneurial ecosystem in both advanced and emerging economies. The concept of a startup incubator has evolved considerably over the years. In both developed and emerging economies, the number of incubators is increasing rapidly, and the degree of networking of incubators is increasing. The innovation ability of higher vocational colleges has been continuously proven in recent years, and it is of great value to give full play to the ability of entrepreneurial incubators in higher vocational colleges to improve the overall innovation ability of China.

2. Related Work

The government regards incubators as a tool to solve the lack of visibility and novelty of new companies and provides financial support for incubators and their client companies [1]. However, many incubators rely on government funding, which makes them vulnerable to economic volatility and political changes. Incubators support startups to develop entrepreneurial business models, but often, their own sustainable business model is missing. Relying on government...
funds is not a sustainable model, and incubators need to find more sustainable funding to guarantee their operations [2].

Given that business models enable the construction and ongoing operation of innovative techno-economic networks [3, 4], and help explain, run, and develop an organization’s operations [5], research on incubator business models is necessary. A business model is a collection of activities carried out to meet the perceived needs of the market, and it is also a specification for different entities to carry out relevant activities. The business model addresses the question of which revenue models will be employed to complement these activities [6, 7]. Simply put, a business model articulates how to identify and satisfy customers by creating value and how an organization and its partners can monetize that value. Therefore, business models are highly relevant to startup incubators.

The research on the business model of incubators began in the 1980s. Scholars first analyzed the impact of key elements of incubators on operational results. The incubator manager and the incubator’s screening criteria for the incubating enterprises have an important impact on the operation of the startup enterprises, and the services provided by the incubator and the integrated resource network have a promoting effect on the startup enterprises [8, 9]. Traditional incubators mainly provide office space and property services [10], but this cannot guarantee the source of income for the sustainable development of the incubator itself, and the incubator should expand its source of income channels [11].

In recent years, scholars have further discussed the overall framework of the incubator business model and constructed the core elements and descriptive framework of the company incubator business model based on technology and capital markets [12, 13]. The operation mechanism of the business incubator is shown in Figure 1. The state and governments at all levels have issued special subsidy policies, and higher vocational colleges have set up special incubators. Different types of incubators have different operation modes, which will directly affect the performance of entrepreneurial enterprises, and then affect the feedback of entrepreneurial enterprises to incubators, so that the operation efficiency of entrepreneurial incubators is affected. The research on how business incubators promote business model change and innovation from the aspects of organizational change, value chain improvement, network location improvement, and institutional entrepreneurship is also increasingly enriched [14, 15]. The business strategy, operation system, and profit model of the incubator should be designed in accordance with the value network theory and the basic logic of value creation [16, 17].

Most of the research on entrepreneurial incubators currently selects entrepreneurial incubators dominated by the government or large companies [18, 19] and analyzes the impact of different business operation models on the performance of incubating enterprises from the perspective of investment [20, 21]. Few scholars have conducted in-depth research on this topic from the professional level of incubators. Relevant studies have focused on the impact of the business incubator operation mode on the performance of startups, including output capacity, innovation capacity, talent attraction capacity, etc., ignoring the impact on the business incubator itself. At the same time, as higher vocational colleges are an important innovation driving force in the current economic growth [22, 23], few scholars have studied how to improve the entrepreneurial incubation capacity of higher vocational colleges.

Compared with the existing literature, this paper may have the following marginal contributions. Firstly, the business incubator operation mode and the business incubator operation efficiency are brought into an analysis framework, providing new evidence to improve the incubator operation efficiency by optimizing the business incubator operation mode. Secondly, different from the existing macroresearch, this paper takes the entrepreneurship incubator of higher vocational colleges as the research sample to analyze the operation efficiency of entrepreneurship incubators of higher vocational colleges with different operation modes from the microlevel.

The rest of this paper is arranged as follows. The third part describes the selection criteria of indicators, data sources, and theoretical models. The fourth part uses DEA to calculate the operation efficiency of entrepreneurship incubators in higher vocational colleges and analyzes the empirical results. The fifth part draws research conclusions and corresponding policy recommendations.

3. Indicator Selection and Data Description

3.1. Principles for Selecting Indicators. In order to construct a scientific evaluation system dedicated to the operation efficiency of national-level technology incubators and obtain feasible solutions, it is necessary to obtain quantitative reports reflecting the structure, operation status, and operation efficiency of the incubators through various methods and conduct further analysis. That is to say, the constructed evaluation system should reflect the overall situation including the resource, economic, social, and environmental aspects of the system. At the same time, according to the overall situation of the incubator, it is necessary to find out the factors that affect the operation status and efficiency and select quantitative factors that can reflect the overall characteristics and interrelationships. In order to ensure the reasonable objectivity of the final result, this paper takes the following standards as the principles to structure the rating system.

3.1.1. The Principle of Purpose. In the process of selecting the indicators of technology business incubators, the index system must be set closely around the main purpose of studying the influencing factors of operating efficiency, and the selected indicators should be more typical indicators in the operating efficiency of technology business incubators. At the same time, the indicator system should reflect the current level of operating efficiency of technology business incubators from as many angles as possible.

3.1.2. Systematic Principles. In order to more accurately evaluate the operating efficiency of the technology business incubator, when selecting indicators, it is necessary to systematically show the use of input resources and the output of the technology business incubator during the operation process. In the process of selecting indicators, more important indicators should be grasped, and the input and output in the operation process should be highly representative.
3.1.3. The Principle of Authenticity. The indicators selected for technology business incubators should use objective data as much as possible and use directly obtained raw data.

Using raw data to support data analysis can reduce your subjective judgment on data and ensure that you use real and reliable data to calculate the actual results. In addition, the original data of the selected indicators should come from the relevant data in the statistical annual report, which is authoritative and reliable.

3.1.4. The Principle of Timeliness. The selected index data should change with the change of time, and the change of the index should be detected in time to adjust the index data to ensure that the evaluation results of the operation efficiency of the technology business incubator are timely. At the same time, in the process of selecting indicators, adjustments should be made with the change of people’s concepts to prevent outdated and invalid evaluation results caused by outdated concepts.

3.1.5. The Principle of Operability. The concept and definition of each indicator in the indicator system should be very clear and not ambiguous, so that the indicators can be understood more easily during the evaluation process, and the indicators will not be confused and the evaluation results will not be invalid. Although the indicators in the system should reflect the input and output situation during the operation of the technology business incubator as much as possible, they should not formulate too subtle indicators, and should formulate important and small quantities of quality to prevent the evaluation work from being too complicated during the evaluation process.

3.1.6. Scientific Principles. The selection of indicators should be based on science, and nonrepetitive and nonoverlapping indicators should be formulated according to the characteristics of technology business incubators, so that the selected indicators are independent. Adhering to the indicators selected by the scientific principle, the analysis results obtained after the analysis will be more reliable, objective, and convincing.

3.2. Selection of Specific Indicators. The choice of input-output index is an important factor that affects the calculation results of the DEA method [24, 25]. The input-output indicators of this paper are as follows.

3.2.1. Input Indicators. The normal operation of the business incubator requires the investment of human, financial, and material resources. First of all, the important investment factors to improve the incubation ability of incubators are high-quality entrepreneurial service talents with professional management ability, which is measured by the number of in-service personnel in incubators. Secondly, incubators provide financing support for incubator enterprises through their own incubation funds, government financial subsidies, and social venture capital. In view of the availability of data, this paper uses the total incubation amount of the incubator as an index of the incubator’s financial investment. Finally, the site status of the incubator is the hardware infrastructure for incubating technology companies. This paper takes the site area of the incubator as the material input of the technology business incubator.

3.2.2. Output Indicators. In this paper, the incubation capability is taken as the factor layer for evaluating the output index of the operation efficiency of technology business incubators, and the cumulative number of graduate companies and the annual graduation rate are taken as the index layer. The performance of entrepreneurial incubators in higher vocational colleges from different provinces across the country is studied through the data on the number of approved intellectual property rights of incubating enterprises.

When collecting data, the data of some provinces were not found. In order to ensure the accuracy of the analysis results, the paper comprehensively selected the “China Torch Statistical Yearbook” data from 22 college and university incubators in Beijing, Shanghai, Jiangsu, Zhejiang, and other regions for research and analysis. The relevant indicators of the operational efficiency of business incubators in higher vocational colleges are composed of seven subindicators selected from two aspects of input and output. The selected indicators are shown in Figure 2.

3.3. DEA Theoretical Basis. The data envelopment analysis method is used to analyze and evaluate the relevant operation efficiency of business incubators in colleges and universities from different provinces through objective methods, so as to further enrich and develop the quantitative research theory and practice of business incubators. Using the Malmquist index to analyze the dynamic efficiency from the perspectives of input and output, we can obtain the technology and efficiency trends of entrepreneurial incubators in different
provinces. The DEA and Malmquist mathematical models will be described below. There are \( j \) decision-making units \( (j = 1, 2, 3, \cdots J) \) in the DEA model, which are represented by DMU. Each decision-making unit \( J \) corresponds to an input vector and an output vector combination \( (X_J, Y_J) \). Based on the following axiomatic assumptions, we can get a DEA model where in the \( J_0 \) decision-making units DMU \( J_0 \) meet the constant return to scale when the input and output are effective.

\[
\begin{align*}
\min \{ \theta \}, \\
\text{s.t. } & \sum_{j=1}^{n} x_j \lambda_j + s^- \leq \theta x_{00}, \\
& \sum_{j=1}^{n} x_j \lambda_j \geq y_{00}, \\
& \lambda_j \geq 0.
\end{align*}
\]

The relaxation variables \( s^- \) and \( s^+ \) are introduced into the above formula, which is transformed into the following formula.

\[
\begin{align*}
\min \{ \theta \}, \\
\text{s.t. } & \sum_{j=1}^{n} x_j \lambda_j + s^- \leq \theta x_{00}, \\
& \sum_{j=1}^{n} x_j \lambda_j - s^+ \geq y_{00}, \\
& \sum_{j=1}^{n} \lambda_j \geq 1, \lambda_j \geq 0, \\
& s^- \geq 0, s^+ \geq 0.
\end{align*}
\]

\( x_{00} \) represents the input quantity of the \( J_0 \)th DMU, \( y_{00} \) represents the output quantity of the \( J_0 \)th DMU, \( \theta \) represents the reduction ratio of input, \( A \) represents the coefficient of the linear unit decision combination, and the symbol with * represents the optimal solution. If there is an optimal solution, it satisfies \( \theta^* = 1 \) and \( S^- = S^* = 0 \); it can be considered that the efficiency of the \( J_0 \) unit is effective. If there is a nonzero value, it meets \( \theta^* = 1 \) and \( S^- = S^* \); it can be considered that the efficiency of the \( J_0 \) unit is invalid. At this time, the closer the nonzero value is to 1, the closer the unit is to being effective.

The Malmquist production index is proposed based on the DEA model method using the ratio of the distance functions to calculate the input-output efficiency. As the index continues to improve and continue, the following classical formula will explain the principle of the Malmquist productivity index.

\[
M_{it+1}(x'_t, y'_t, x'^{t+1}_t, y'^{t+1}_t) = \left[ \frac{D'_t(x'_t, y'_t)}{D'^{t+1}_t(x'^{t+1}_t, y'^{t+1}_t)} \cdot \frac{D'_t(x'^{t+1}_t, y'^{t+1}_t)}{D'^{t+1}_t(x'_t, y'_t)} \right]^{1/2}.
\]

Among them, \( x'_t, x'^{t+1}_t \), respectively, represent the input vector of the \( i \)-th incubator in the period \( t \) and \( t + 1 \). \( y'_t, y'^{t+1}_t \), respectively, represent the output vector of the \( i \)-th incubator in the period \( t \) and \( t + 1 \). \( D'_t(x'_t, y'_t), D'_t(x'^{t+1}_t, y'^{t+1}_t) \), respectively, represent the technology \( T^t \) in the period as the distance function for the referred production points at period \( t \) and period \( t + 1 \).

\[
M_{it+1}(x'_t, y'_t, x'^{t+1}_t, y'^{t+1}_t) = \left[ \frac{D'^{t+1}_t(x'^{t+1}_t, y'^{t+1}_t)}{D'_t(x'_t, y'_t)} \cdot \frac{D'_t(x'_t, y'_t)}{D'^{t+1}_t(x'^{t+1}_t, y'^{t+1}_t)} \right]^{1/2}.
\]
Equation (4) is a modification of Equation (3), which is used to express the separation of technological change and technological efficiency change. The first half is the change in production efficiency from \( t \) to \( t + 1 \); the second half is the rate of change in technology from \( t \) to \( t + 1 \).

\[
M^{t+1}_{x,c} = \frac{D_v^{t+1}(x^{t+1}_i, y^{t+1}_i)}{D_c^{t}(x^{t}_i, y^{t}_i)} \times \left[ \frac{D_v^{t+1}(x^{t+1}_i, y^{t+1}_i)}{D_c^{t+1}(x^{t+1}_i, y^{t+1}_i)} \right] \times \left[ \frac{D_c^{t}(x^{t}_i, y^{t}_i)}{D_c^{t+1}(x^{t+1}_i, y^{t+1}_i)} \right]. \tag{5}
\]

Equation (5) relaxes the assumption of fixed return to the scale of equations (3) and (4), describes the situation of the variable return to scale, and further decomposes technical efficiency changes into pure technical efficiency changes and scale efficiency changes. Footnote \( v \) indicates the variable return to scale. Footnote \( c \) is the case of constant compensation, the first term represents the change of the pure technical efficiency under the variable scale, the second term is the change of the scale efficiency, and the third term represents the rate of technological change. Through the decomposition of the formula, we can get the source of the total efficiency change. Further, we can see in the following that the differences in the operation efficiency of incubators with different operation modes are mainly reflected in what aspects.

### 4. Empirical Result Analysis

#### 4.1. Comprehensive Analysis

To investigate the overall development situation of Chinese business incubators, we use the DEA data envelopment analysis to analyze the operational efficiency of state-level business incubators in the eastern region, the central and western region, and the northeastern region. This part of the data comes from “China Torch Statistical Yearbook 2017-2019.” DEAP2.1 is selected as a tool to measure the operational efficiency of entrepreneurship incubators in higher vocational colleges.

At present, there are a lot of DEA software developed in China, mainly including DEA Solver, Frontier Analyst, OnFront, Warwick DEA, DEAP, EMS, and MaxDEA. By considering the operating platform, system configuration, model, and economic value of the software, this paper select DEAP2.1. The software has the following features.

1. There is no limit to the number of decision-making units
2. There are more than 30,000 option combination models
3. Fully support the directional distance function model
4. The result processing is clear and easy to interpret

Using DEAP2.1 software, the input-output data from 2016 to 2018 in the above 22 provinces were measured, and the efficiency values were obtained. The 22 provinces studied were divided into the eastern region, the central and western region, and the northeastern region. The eastern region includes Beijing, Tianjin, Hebei, Liaoning, Shanghai, Jiangsu, Zhejiang, Fujian, Shandong, and Guangdong, and the central and western region includes Anhui, Jiangxi, Henan, Hubei, Hunan, Chongqing, Sichuan, Guizhou, and Shaanxi. The northeastern region includes Heilongjiang, Jilin, and Liaoning. Considering that there may be abnormal values in the data used to calculate the operation efficiency of business incubators in higher vocational colleges, which makes the calculation results biased, this paper makes a tailing treatment for the continuous variables involved in the DEA model at the level of 1% before and after. That is, if the variable value is less than 1% quantile and greater than 99% quantile, we make its value equal to 1% quantile and 99% quantile, respectively.

It can be seen from Figure 3 that from 2016 to 2018, the comprehensive efficiency value of the national-level entrepreneurial incubators in the eastern region, the central and western region, and the northeastern region are all 1. The operation efficiency of incubators in these three regions is effective. The first reason may be that entrepreneurial incubators provide human support to startups, including professional exchanges or management training, so that startups can reduce the cost of mistakes. The second possibility is that entrepreneurial incubators use their own resource reserves to provide entrepreneurial companies with opportunities from upstream and downstream of the value chain, so that startups can get more development opportunities and then feedback the incubator, making the incubator efficient.

Because of the modeling defect of the fundamental DEA model itself, the calculated efficiency of the DEA effective decision unit can only be 1, so it is impossible to compare the same DEA effective decision unit at depth. Therefore, Andersen and Petersen constructed a superefficient DEA model [26]. This scalar measure deals directly with the input excesses and the output shortfalls of the decision-making unit (DMU) concerned. It is units are invariant and monotone decreasing with respect to input excess and output shortfall [27]. A reference set composed of a superefficient DEA model does not include an evaluator itself that is different from the base DEA model. Therefore, when the efficiency value is 1, the efficiency value obtained by the above method may exceed 1, so the superefficient DEA model is very effective when comparing between DEA-effective decision-making units. In order to further analyze the operating efficiency of incubators in the eastern region, the central and western region, and the northeastern region of China, this paper uses DEAP2.1 software to calculate the maximum efficiency of DEA in these three regions. The superefficient value means that these incubators can still remain effective if they are scaled up in a certain proportion.

It can be seen from Figure 4 that from 2016 to 2018, the average superefficient values of the eastern region, the central region, and the northeast region are all greater than 1. The average superefficiency of the eastern region is 130.68%, indicating that the incubators in the eastern region can still maintain the relative effectiveness of DEA by increasing the investment by 30.68%. The central and
western region can still keep DEA relatively effective by increasing investment by 23.92%. The northeast region can still maintain the relative effectiveness of DEA by increasing the investment by 171.9%. Therefore, compared with the eastern region and the central and western region, incubators in northeastern region still have a lot of room to rise, and it is necessary to increase the investment of entrepreneurial incubator resources in the northeastern region.

Through the “China Torch Statistical Yearbook 2017-2019,” it can be found that compared with other regions, the investment in entrepreneurial incubators in the northeastern region is less. In addition, entrepreneurial incubators in the northeastern region have a lot of room for growth and need to increase investment. In order to find out the crux of the development problem of the northeastern region’s business incubator, this paper conducts a microanalysis of the operation efficiency of the northeastern region’s business incubator. We selected microsamples from the northeastern region to further analyze its operating efficiency. The data of this part of the sample comes from the “China Torch Statistical Yearbook 2018.” In 2017, 43 incubators in the northeastern region were selected as national-level incubators.

Excluding the incubators with incomplete data, this paper finally obtained 32 valid samples.

4.2. Northeastern Region Sample Analysis. It can be seen from Figure 5 that, first of all, the average comprehensive
Table 1: Analysis results of the operational efficiency of incubators in higher vocational colleges in northeastern region.

| Code | Professional business incubators | Comprehensive business incubators |
|------|---------------------------------|---------------------------------|
|      | Comprehensive efficiency | Technology efficiency | Scale efficiency | Scale remuneration | Code | Comprehensive efficiency | Technology efficiency | Scale efficiency | Scale remuneration |
| C1   | 1 | 1 | 1 | — | — | C17 | 0.308 | 0.497 | 0.62 | drs |
| C2   | 1 | 1 | 1 | — | — | C18 | 0.539 | 0.558 | 0.967 | drs |
| C3   | 1 | 1 | 1 | — | — | C19 | 1 | 1 | 1 | — |
| C4   | 0.647 | 0.727 | 0.615 | drs | C20 | 0.785 | 1 | 0.785 | irs |
| C5   | 1 | 1 | 1 | — | — | C21 | 1 | 1 | 1 | — |
| C6   | 1 | 1 | 1 | — | — | C22 | 0.769 | 0.791 | 0.972 | drs |
| C7   | 0.653 | 1 | 0.653 | irs | C23 | 1 | 1 | 1 | — |
| C8   | 0.694 | 0.802 | 0.865 | irs | C24 | 0.737 | 0.853 | 0.864 | drs |
| C9   | 1 | 1 | 1 | — | — | C25 | 1 | 1 | 1 | — |
| C10  | 0.514 | 0.631 | 0.815 | drs | C26 | 0.873 | 0.881 | 0.99 | irs |
| C11  | 1 | 1 | 1 | — | — | C27 | 0.513 | 0.635 | 0.807 | drs |
| C12  | 0.262 | 0.359 | 0.73 | drs | C28 | 0.341 | 0.559 | 0.611 | drs |
| C13  | 0.621 | 0.776 | 0.8 | irs | C29 | 1 | 1 | 1 | — |
| C14  | 0.633 | 0.659 | 0.96 | irs | C30 | 0.885 | 0.966 | 0.916 | drs |
| C15  | 1 | 1 | 1 | — | — | C31 | 0.664 | 0.668 | 0.994 | irs |
| C16  | 1 | 1 | 1 | — | — | C32 | 1 | 1 | 1 | — |
| Average | 0.853 | 0.872 | 0.902 | — | Average | 0.798 | 0.838 | 0.908 |

Note: drs: diminishing returns to scale; irs: increasing returns to scale.

efficiency of the 32 entrepreneurial incubators is 0.789, indicating that 21.1% of the investment in the overall entrepreneurial incubators of higher vocational colleges in the northeastern region is wasted or does not have any benefit or contribution to the output. At the same time, among the 17 incubators with ineffective operation efficiency, 10 have diminishing returns to scale, and these incubators should be controlled to continue to expand or shrink to improve overall efficiency. Therefore, we cannot blindly increase the investment in business incubators of higher vocational colleges in the northeastern region. For incubators with decreasing scale efficiency, it is necessary to control their scale and reduce investment. Secondly, only 15 of the 32 incubators in higher vocational colleges in the northeastern region are relatively efficient in operation. From a microscopic point of view, it shows that the operation efficiency of incubators in higher vocational colleges in the northeastern region is not very good. Finally, the value of pure technical efficiency of noneffective incubators is between 0.3 and 0.99, indicating that the pure technical efficiency of incubators in higher vocational colleges in the northeastern region varies greatly, indicating that the development of operational capabilities among these ineffective incubators is very uneven. Therefore, improving the operational capability of the incubator is an effective way to improve the operational efficiency of ineffective incubators in higher vocational colleges in the northeastern region.

In order to further explore the internal problems of the development of entrepreneurial incubators in the northeastern region, referring to existing research, this paper divides the incubators into two types, professional incubators and comprehensive incubators, and conducts a classification analysis to explore the internal development differences of the incubator operating efficiency of different dominant modes in the northeastern region to find out the crux of the problem that ultimately affects the development of entrepreneurial incubators.

Comprehensive incubators refer to startup incubators that do not specifically limit the industrial direction of the hatched enterprises, and startups that meet the category of small- and medium-sized enterprises can apply for admission. Professional incubators refer to startups that are incubated in specific industries and need to meet the direction of industrial development. Based on this classification principle, this paper classifies the 32 business incubators in the northeastern region and observes the operating efficiency of each enterprise. The summary is shown in Table 1.

The comprehensive efficiency and technical efficiency of professional incubators in higher vocational colleges in the northeastern region are obviously better than those of comprehensive incubators.

The overall efficiency and technical efficiency of incubators are unevenly distributed. It can be seen from Figure 6 that 56.25% of the professional incubators have a comprehensive efficiency value of 1, and only 37.5% of the comprehensive incubators have a comprehensive efficiency value of 1. In addition, the comprehensive efficiency value and technical efficiency value of 64.23% of the professional incubators with ineffective operation efficiency are concentrated above 0.6, close to the effective value of 1, indicating that as long as the operating capacity of the incubator is slightly improved and resource allocation is optimized, most of the professional incubators can operate efficiently. The average comprehensive efficiency of the comprehensive incubator is
0.798; it means that 21.2% of the input elements are wasted or have no benefit or contribution to the output. This shows that there is considerable room to improve the operational efficiency of incubators.

Because the comprehensive incubated enterprises involve many industries, it is impossible for the incubator to have a deep understanding of each industry, and it is difficult to provide strong strategic guidance for each entrepreneurial enterprise. And because entrepreneurial enterprises in different industries have a great difference in the degree of demand for resources, the demand of enterprises for resources will lead to a variety of improper resource-seeking actions. This will affect the allocation efficiency of entrepreneurial incubator resources, making the overall operation efficiency low. When choosing enterprises, professional business incubators will focus on their best areas, which is easy to achieve win-win results with startups, and the relative efficiency will be higher. The resource input of comprehensive incubators should be controlled, and investment in professional incubators should be increased.

In addition, as shown in Figure 7, the increasing return to scale of professional incubators with low operating efficiency accounts for 60%, while the decreasing return to scale of income of integrated incubators with low efficiency accounts for 70%. This shows that most professional incubators can further improve the operation efficiency of incubators by expanding the coverage of incubators; while most comprehensive incubators should avoid blindly expanding the scale of incubators, the focus should be on improving the efficiency of business capacity and investment.

Considering the use of DEAP2.1 to calculate the operation efficiency of higher vocational colleges, there may be deviations that affect the experimental results; this paper tests the robustness. MaxDEA is used to measure the operation efficiency of entrepreneurial incubators in different dimensions of regions and enterprises. The relative level of the calculation results is consistent with the previous text, which shows that the research results of this paper are reliable.

5. Conclusion

To sum up, through an overall analysis of the operation efficiency of business incubators in higher vocational colleges in my country from a regional perspective, it is found that business incubators in higher vocational colleges play a significant role in cultivating high-quality enterprises as a whole. First, on the whole, the business incubators in higher vocational colleges are still in the stage of increasing the investment scale to improve the operation efficiency. Therefore, local governments can improve the overall entrepreneurial efficiency by standardizing the support policies for business incubators. Second, the northeastern region currently has the largest room for improvement, and the state should give priority to increasing support for entrepreneurship incubators in higher vocational colleges in the northeastern region. Third, the efficiency of comprehensive entrepreneurial incubators in higher vocational colleges in the northeastern region is low. Guiding the transformation to professional business incubators is an important means to improve the operation efficiency of local and overall business incubators. Priority should be given to guiding higher vocational colleges to develop industrial incubators related to their key disciplines.
Data Availability

The datasets used during the current study are available from the corresponding author on reasonable request.

Conflicts of Interest

The authors declare that they have no conflict of interest.

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