# An Application of the Data Envelopment Analysis Method to Evaluate the Performance of Academic Departments within A Higher Education Institution

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# Abstract

This paper aims to evaluate the efficiency or the productivity of academic departments within a university using Data Envelopment Analysis. As an illustrative example, we investigate the performance of 57 departments of National Economics University (NEU) for three years, from 2013 to 2015. The data set consists of one input variable, which is the number of academic staff, and three output variables in which the number of research hours is considered as research output and the number of graduates and teaching load are defined as teaching outputs. Particularly, the output-oriented CCR, BCC, and SBM model under both the CRS and VRS assumptions are applied in order to determine accurate degrees of efficiency of individual departments and directions for performance improvement for less efficient ones. The output-oriented radial Malmquist DEA model is also employed to make a comparative analysis of the productivity change of the departments over the period. The results reveal some clear policy-making implications for departments to adjust their development plan in an appropriate way.

**Keywords**: Data envelopment analysis; relative efficiency; higher education institution; academic departments' efficiency.

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# 1. Introduction

It is widely recognized that tertiary education is a major driver of economic development in an increasingly knowledge-driven global economy. Therefore, in recent decades the tertiary education sector in all countries has been expanded rapidly in both width and depth.

In almost all countries, this sector has been experiencing reforms aiming at improving efficiency and quality, thus becoming more responsive to the needs of society and the economy. With regard to an individual higher education institution (HEI), academic departments are the core, so the reform of an HEI should start from its academic departments. The evaluation of the performance of academic departments is part of the process of resource allocation within a university, which is a politically difficult task for most public and private universities (Arcelus and Coleman, 1997). The measurements of the performance of academic activities consist of two major categories: efficiency and effectiveness. 'Efficiency' is a measure of the work-rate of a process by which system inputs are turned into system outputs. 'Effectiveness' on the other hand is considered to be a measure of the 'quality' or 'fitness-for-purpose' of the outcomes being achieved by the system (Woodhouse, 2001). The focus of this paper is to examine the efficiency or the productivity of academic departments within a university.

Efficiency has been studied using economic methods including *cost-benefit analysis*, *cost-effectiveness analysis* or *unit cost analysis*, and *cost-consequences analysis*. Using these methods, it is required to have accurate cost and/or benefit data. However, in the education sector, some costs and benefits are not easy to quantify, so the result gained cannot be ensured to be accurate. An alternative approach to examine internal and external efficiency of educational institutions is modeling the education production function. Educational institutions here are treated as 'producers' of educational outcomes. Regression analysis may be used to test these models of the education production function. However, the traditional parametric estimations, like ordinary least squares, fit a regression line through the data. That means they characterize the behavior of the average institution. Institutions lying below the regression line have costs less than the minimum cost. This violates the cost-minimization assumption (Salerno, 2003). Moreover, the education production process is complex and unobtrusive, so it is not easy to model education production functions.

In recent decades, there have been numerous studies assessing the productivity and efficiency of operational units of different fields including education, and thus contributing to the development of the Stochastic Frontier Estimation (SFE) and Data Envelopment Analysis (DEA). With these tools, researchers are capable of, and more flexible in, modeling the complex production processes and cost structures within higher education institutions. SFEs are parametric programming techniques or regression-based estimators, which are only different from traditional parametric regressions in the error term. SFE analyses has some advantages, but still raises some concerns among researchers. First, it is required to specify production function which is largely unknown in the case of higher education. Second, it is also required to make assumptions on the distribution of efficiencies. Third, it is not generally possible to jointly estimate the influence explanatory variables have on multiple expenditures (in the case of cost functions) or multiple outputs (in the case of production functions) (Salerno, 2003). Meanwhile, DEA is a set of non-parametric programming techniques or mathematical programming estimators which "assist in identifying which set of decision making units may be considered as best practice" in the use of resources among a group of like units (Abbott and Doucouliagos, 2003). DEA is the predominant tool for evaluating the performance of HEIs for a number of reasons. First, it proves to be a useful tool to identify degrees of efficiency of individual units and the direction for performance improvement for less efficient ones. Second, DEA, as a non-parametric technique, has a technical advantage over SFE in the sense that it can avoid "the need to make assumptions regarding the functional form of the best practice frontier ... as well as ... the need to make distributional assumptions regarding the residuals in the regression analysis" (Abbott and Doucouliagos, 2003). Third, "DEA is widely lauded for its ability to estimate efficiency where firms use multiple inputs to produce multiple outputs and the underlying production relationship is not well understood" (Salerno 2003). Fourth, this method allows us to model "the complex production processes and cost structures within higher education institutions" (Salerno, 2003). Fifth, like production function modeling, DEA is used to discover which types of resources and which allocations have the greatest effect on outcomes and how input-choice and allocative efficiency could be improved. This method "yields more

information about the education process than cost-effectiveness analysis" (Belfield, 2000). Besides, it is noted that DEA is a good tool to assess relative efficiency, i.e., the best practice units are recognized as efficient ones relative to the ones being evaluated. However, this may lead to some problems. First, the best performing units or the efficient units may not be operating efficiently in absolute terms. Moreover, it is not possible to undertake tests of statistical significance with DEA scores as is possible with regression analyses (Abbott and Doucouliagos, 2003). Second, there may be the issue of the quality of outputs. However, this issue is not limited to DEA. Third, DEA is sensitive to sample size and data errors as outliers in the data may alter the shape of the efficient frontier and distort efficiency scores of units using similar input/output proportions.

This paper uses DEA to evaluate the efficiency of the academic departments of NEU. We choose NEU as an example of the analysis because it is a leading university in the field of economics and business in Vietnam and it is also one of the first few universities of Vietnam given autonomy by the Ministry of Education and Training (MOET). This pilot program began with financial independence since 2006 and then continued with autonomy in making decisions on training, research, international cooperation and higher tuition fees in recent years. The analysis illustrates how the recent developments in efficiency analysis can be applied when evaluating the performance of an educational institution. Moreover, the results will hopefully be a subjective source of reference for resource allocation policy.

The paper is organized as follows. Section 2

reviews the existing relative literature. In section 3, methodology and data are presented. Section 4 discusses the results of the empirical analysis. Finally, section 5 gives conclusions on the gained results and some policy implications.

# 2. Literature review

As mentioned above, DEA is a very popular and useful tool to assess efficiency of operating units like HEIs, which use multiple inputs to produce multiple outputs and have an underlying production relationship that is not well understood. We can say that the studies on HEIs' efficiency evaluation applying DEA are very diversified.

The majority of the studies, including both within a country and cross-country ones, relating to performance evaluation are at the university level.

For example, Johnes (2006) introduces an application of DEA to a data set of more than 100 HEIs in England using data for the year 2000/01. The study considers the number and quality of undergraduates, the number of postgraduates, expenditure on administration, and the value of interest payments and depreciation as inputs, and the number and quality of undergraduate degrees, and the number of postgraduate degrees and research as outputs. The results indicate that the English higher education sector has high technical and scale efficiency on average. However, bootstrapping procedures suggest that differences between the most and least efficient English HEIs are significant.

Another example is the study by Kocher et al. (2006) which measures productivity in topedge economic research of 21 OECD countries using DEA. The output variable is the number of publications in 10 economics journals with the highest average impact factor over the time period 1980-1998 and inputs are R&D expenditures, number of universities with economics departments and (as an uncontrolled variable) total population.

Interestingly, there have been numerous studies evaluating the performance of departments of the same field for different universities such as Tomkins and Green (1988), Johnes and Johnes (1993), Madden et al. (1997), Kao and Hung (2008), Agasisti et al. (2012) and others.

Madden et al. (1997) analyze the effect of the higher education system policy changes on the efficiency of Australian economics departments using the data collected from 29 universities for the years 1987 and 1991. Number of staff is the unique input, while core journals, other journals, books, and edited books are considered as research outputs and undergraduates and postgraduates are teaching outputs.

Agasisti et al. (2012) use DEA to measure the efficiency of 69 academic departments (focused on scientific subjects) located in the Lombardy region of Italy. The study focuses on investigating the trade-off among different research outputs of academic departments which include quantity (publications), quality (citation indexes), research funds obtained through research grants, and applied research funds obtained through external orders. The results of the output-oriented DEA model show that efficiency rankings change significantly when considering different research-related outputs, and thus the study has room for discussion of different research strategies among the academic departments.

While the majority of the relative existing studies evaluate the performance of departments of the same field for different universities, a smaller number of studies (e.g. Sinuany-Stern et al., 1994; Arcelus and Coleman, 1997; Kao and Hung, 2008; George et al., 2012; and others) examine efficiency of departments of different fields within the same university.

Sinuany-Stern et al. (1994) were the first to measure the efficiency of departments within the same university (the Ben-Gurion University) using the DEA CCR model. For them, DEA has many advantages, but still encompasses some drawbacks in dealing with this kind of analysis. The main drawback of DEA is that it assumes that *DMU*s (decision-making unit–departments in this situation) are homogeneous. So, the study applies cluster analyses to divide the departments into several sets and uses discriminant analysis to test the match of the efficiency/inefficiency division of the CCR ratio. It also tests organizational changes for inefficient departments.

George et al. (2012) measure efficiency of the departments of a public owned university (the University of Thessaly) using DEA. Efficiency scores are estimated under the CRS and VRS assumption. This study assumes that departments inside a university may be considered as homogeneous because they have similar activities and are willing to achieve the same goals. However, the departments under examination are somewhat heterogeneous in some aspects. First, since the university under examination is a state owned university, the funding from the government is allocated to the different departments based on different resource criteria. Second, the departments are of different fields. Thus, bootstrap techniques are applied to overcome these problems and determine more accurate performance estimates. The inclusion of these procedures helps minimize the heterogeneity related problems regarding the comparability of departments from different fields and thus produce bias corrected results. The results show that there are strong inefficiencies among the departments, indicating the misallocation of resources or/and inefficient application of departments' policy developments. Like other studies using DEA, the study provides benchmarks for the long term sustainability of the departments.

In Vietnam, we could find only a few studies on evaluating the performance of the Vietnamese higher education system. Nguyen (2008) evaluated efficiency of the state budget for universities in Vietnam and Austria using the data of 2001 and 2006. The two-stage DEA method is used to measure relative efficiency compensating for non-homogeneity of *DMU*s. The study also compares the efficiency between two points in time using the output-oriented radial Malmquist DEA model. The results show that the overall efficiencies of the system were rather low and the discrepancy of efficiency scores between universities was big in both time periods evaluated.

Carolyn-Dung et al. (2016) conducts an analysis of the academic performance of HEIs in Vietnam with 50 universities and 50 colleges in 2011/12 using the two-stage semi-parametric DEA. The study reveals some important findings in that: (i) there are still potential avenues to improve the existing performance; (ii) there appears to be a difference in the efficiencies of public and private HEIs in the reported year; and (iii) the inefficiency of HEIs is not entirely a result of managerial performance, but is also influenced by other factors such as location, age and the contribution of tuition fees.

It should be noted that we have not found so far any research on evaluating the efficiency of different departments within a university in Vietnam.

## 3. Methodology and data

### 3.1. Methodology

#### DEA framework

DEA was developed on the basis of the seminar paper by Farrell (1957) and first introduced by Charnes et al. (1978) to measure relative efficiency between like organizations. They introduced a ratio definition of efficiency, also called the CCR ratio definition, which generalizes the classical engineering science ratio definition of single output and single input to multiple outputs and multiple inputs without requiring pre-assigned weights. The efficiency of a DMU (department in this case) is measured in relation to the other observed DMUs while assuming that all DMUs lie on or below the efficiency frontier. The data set used in this study contains 57 DMUs, assuming that all efficient DMUs' positions represent the efficiency frontier, below which lie inefficient DMUs.

This definition is expressed in the following fractional model:

$$\max f_{o} = \frac{\sum_{r=1}^{m} u_{r} y_{ro}}{\sum_{i=1}^{m} v_{i} x_{io}}$$
(3.1)  
subject to  
$$\frac{\sum_{i=1}^{m} v_{i} x_{ij}}{\sum_{i=1}^{m} v_{i} x_{ij}} \le 1, \quad j = 1, ..., n$$
$$u_{r}, v_{i} > 0, \qquad i = 1, ..., m, \quad r = 1, ..., s.$$

whereb  $y_{rj}$ ,  $x_{ij}$  (> 0) represent output and input data for  $DMU_j$  with the ranges for *i* and *r*. The objective is to obtain weights  $v_i$  and  $u_r$ that maximize the ratio of the evaluated  $DMU_o$ ( $\theta^*$ ). The model can be written in the linear programming form as follows:

 $\theta = \sum_{r=1}^{s} u_r y_{r0}$ 

max

subject to

$$\sum_{i=1}^{m} v_i x_{i0} = 1$$
  
-
$$\sum_{i=1}^{m} v_i x_{ij} + \sum_{r=1}^{s} u_r y_{rj} \le 0$$
  
 $v_i \ge 0, \ u_r \ge 0.$ 

(3.2)

The CCR-type models, under "weak efficiency"<sup>1</sup> (or Farrell efficiency), evaluate the radial (proportional) efficiency  $\theta^*$ , but do not take account of the input excesses and output shortfalls. Under the CCR-efficiency, which adds the Pareto-Koopmans efficiency conditions, a *DMU* is called CCR-efficient if it satisfies both conditions:  $\theta^*= 1$  and all slacks are zero (Cooper et al., 2007).

In reality, there are many approaches to the definition of efficiency in its relation to productivity. Efficiency can be considered as an attempt to minimize inputs while producing at least the given output levels, or in another way, efficiency involves maximizing outputs while using no more than the given inputs (Cooper et al., 2007). The first approach is an input-oriented approach, and the second is an output-oriented one. In this study, we employ the output-oriented CCR model to measure the relative efficiency of all observed *DMU*s. The model is written in the dual linear programming form as follows:  $\max_{\delta,\mu}$ 

subject to

$$x_{i0} - \sum_{j=1}^{n} x_{ij} \mu_j \ge 0$$
$$\delta y_{r0} - \sum_{j=1}^{n} y_{rj} \mu_j \le 0$$
$$\mu_j \ge 0.$$

(3.3)

δ

In this model,  $\delta$  is a scalar satisfying  $\delta \ge 1$ and a measure of technical efficiency of the observed *DMU* or a measure of the distance of its position to the efficiency frontier. If  $\delta > 1$ , the *DMU* is inside the frontier or inefficient. If  $\delta$ = 1, the *DMU* is on the frontier or efficient.  $\mu$ , a vector of constants, measures the weights to project inefficient *DMU*s on the frontier.

We also test the relative efficiency of *DMUs* taking into account variable return-to-scale characterizations by adopting the output-oriented BCC model developed by Banker et al. (1984). The constraint  $e\mu = 1$  will be added to the equation (3.3).

As mentioned above, the CCR and BCC models have a drawback, that is they do not take into account non-radial non-zero slacks, while the slack-based models (SBM) proposed by Tone (2001) do. According to Tone, a DMU is CCR-efficient if and only if it is SBM-efficient. Based on this relationship between CCR-efficiency and SBM-efficiency, the study will also use the output-oriented SBM model to have a deeper look into the status of DMUs when knowing their non-radial non-zero slacks. The output-oriented SBM model is formulated as follows:  $\rho_0^* = \min_{\lambda, s^*} \frac{1}{1 + \frac{1}{s} \sum_{r=1}^s s_r^* / y_{r0}}$ (3.4)subject to  $x_0 \ge X\lambda$  $y_0 = Y\lambda - s^+$  $\lambda \ge 0, s^+ \ge 0.$ 

In the formula (3.4),  $\rho^*$  is the efficiency score of the observed *DMU*.  $s_r^+$  is the output shortfall. Please note that  $\rho^*$  in (3.4) is never greater than  $\delta^*$  in (3.3) because (3.4) includes output slacks.

# Comparison of efficiency between different time periods using the Malmquist index

The Malmquist index was first introduced by Malmquist (1953) and further developed by many authors to measure total factor productivity (TFP) growth of a *DMU* between different points in time in the non-parametric framework. In other words, it allows one to make a comparative analysis of the productivity change of a *DMU* between different periods of time.

The Malmquist index consists of two components: catch-up and frontier-shift. The catchup component indicates "the degree to which a DMU improves or worsens its efficiency", and the frontier-shift represents "the change in the efficient frontiers between the two time periods" (Cooper et al., 2007, 328). As mentioned in the previous sections, the data set has n DMUs, each of which uses m inputs (denoted by vector  $x_i$ ) to produce s outputs (denoted by vector  $y_i$ ) over the periods 1 and 2, assuming that  $x_i > 0$  and  $y_i > 0$ . Here,  $DMU_o$  in period t is denoted by  $(x_o, y_o)^t$  and efficiency score of  $DMU_{o}$  at period t by  $\delta^{t}(x_{o}, y_{o})^{t}$ . The production possibility set  $(X, Y)^t$  (t = 1 and 2) is defined as follows:

$$(X,Y)^{t} = \left\{ \left(x,y\right) \mid x \ge \sum_{j=1}^{n} \lambda_{j} x_{j}^{t}, 0 \le y \le \sum_{j=1}^{n} \lambda_{j} y_{j}^{t}, L \le e\lambda \le U, \lambda \ge 0 \right\}$$

Where *e* is the row unit vector,  $\lambda$  is the intensity vector, and *L* and *U* are the lower and upper bounds for the sum of the intensities. Then, the catch-up effect (*C*) from period 1 to 2 is calculated as follows:

$$Catch - up = \frac{\delta^{2}(x_{o}, y_{o})^{2}}{\delta^{1}(x_{o}, y_{o})^{1}}.$$
 (3.5)

If C> 1, there is progress in relative efficiency from period 1 to 2. If C = 1, there is no change in efficiency. If C< 1, there is regress in efficiency.

Frontier-shift effect (F) is measured by:

Frontier - shift = 
$$\left[\frac{\delta^{1}((x_{o}, y_{o})^{1})}{\delta^{2}((x_{o}, y_{o})^{1})} \times \frac{\delta^{1}((x_{o}, y_{o})^{2})}{\delta^{2}((x_{o}, y_{o})^{2})}\right]^{1/2}$$
 (3.6)

If F > 1, there is progress in the frontier technology around  $DMU_o$  from period 1 to 2. If F = 1, there is no change. If F < 1, there is regress in the frontier technology.

The Malmquist index (*MI*) is synthesized from catch-up and frontier-shift as follows:

$$\begin{aligned} Malmquist index &= (Catch - up) \times (Frontier - shift) \\ &= \left[ \frac{\delta^1 \left( (x_o, y_o)^2 \right)}{\delta^1 \left( (x_o, y_o)^1 \right)} \times \frac{\delta^2 \left( (x_o, y_o)^2 \right)}{\delta^2 \left( (x_o, y_o)^1 \right)} \right]^{1/2} \quad (3.7) \end{aligned}$$

If MI> 1, there is growth in the total factor productivity of the  $DMU_o$  from period 1 to 2. If MI = 1, there is no change. If MI< 1, there is decay in the total factor productivity (Cooper et al., 2007).

As presented in the previous section, this study measures efficiency of *DMUs* on a performance basis or on an output-oriented basis. Thus, following Nguyen (2008), it will employ the output-oriented CCR-, output-oriented BCC-, output-oriented SBM-, and output-oriented radial Malmquist DEA models to evaluate the change in efficiency scores, the technological change as well as the total factor productivity change of the *DMUs* from period 1, represented by the year 2013, to period 3, represented by the year 2015.

To illustrate an approach to evaluate efficiency of academic departments within a university in Vietnam, we took NEU as an example. The data for the three years 2013, 2014 and 2015 were requested from the Office of Personnel Management, the Office of Research Management, the Office of Training Management, the Advanced Educational Programs, the Department of In-Service Training, and the Graduate School of NEU. This data set consists of input and output variables for all 57 departments at the NEU (except for the Department of Physical Education and the Faculty of Military Education).

## Inputs

Alongside capital and funding amounts and sources, academic staff is the principal input into the departmental production and also the most commonly used input variable in the existing literature. However, capital and funding are commonly shared between departments and are not allocated by apparent criteria. Therefore, this study used as input the number of academic staff, which is constituted only by lecturers. Furthermore, there are five ranks of lecturers at NEU, namely professors, associate professors, lecturers having Ph.D.'s, master's and bachelor's degrees, so we pre-assigned weights to each rank in order to construct a proper aggregate measure to capture both the quantity and the quality of academic staff. Weights were pre-assigned based on the assumption that higher-ranking lecturers are expected to be more efficient in teaching and produce more research works than lower-ranking ones so they need to be assigned with a greater weight. Thus, professors were assigned with 1, associate professors with 0.8, PhDs with 0.6,

							8							
	2013				2014			2015						
	Research	Grad	Teach		Research	Grad	Teach		Research	Grad	Teach			
Research	1.000	0.151	0.349	Research	1.000	0.160	0.253	Research	1.000	0.274	0.396			
Grad	0.151	1.000	0.083	Grad	0.160	1.000	0.128	Grad	0.274	1.000	0.234			
Teach	0.349	0.083	1.000	Teach	0.253	0.128	1.000	Teach	0.396	0.234	1.000			

**Table 1: Correlations among outputs** 

Masters with 0.4 and Bachelors with 0.2. Results from various robustness checks show that the choice of the weights does not alter the final results; therefore our pre-assigned arbitrary choice is not affecting the calculated efficiencies (George et al., 2012). These weights are chosen so the distance between the two ranks is 1/5=0.2.

# Outputs

Outputs were classified into teaching outputs and research outputs. Teaching outputs consist of the teaching load and the number of graduating students of each department yearly. Specifically, a department's teaching load was constituted by the total number of periods taught by the department in graduate and undergraduate programs, which comprise all types of programs ranging from the ordinary programs, second-degree programs, advanced educational programs to in-service training programs. Whereas the number of graduates was calculated by the total number of postgraduate and undergraduate degrees awarded each year weighted by training levels. That is, doctoral students were assigned with 1, master students with 0.666, and undergraduates with 0.333 (George et al., 2012). Besides, research outputs were measured by the total number of research hours of each department in a year following the current NEU's internal controlling regulations. That is, the amount of a department's research was calculated by the sum of the weighted numbers of projects of different levels, recognized journal articles, conference papers, text-books and reference books published, prized student research projects instructed, and other forms of research done by all academic staff of the department.

At the end, we have completed data for 57 academic departments at NEU for the period of three years, 2013-2015. The provision of incomplete data from nine departments in the year 2015, however, reduced the sample of 2015 to 48 observations.

As can be seen from Table 1, there is little accordance among the three outputs. Table 2 illustrates the descriptive statistics for the input and outputs employed in the study. It is evident from the summary statistics that while the number of graduates and the teaching load saw a slight decline due to some changes in regulations, the number of staff and the number of research hours of each department increased steadily over the three-year period. It is also noteworthy that the standard deviations are significantly high for all variables, especially for the number of graduating students, which implies considerable differences among the de-

	Variables	Staff number	Research hours (1)	Graduates number (2)	Teaching load (3)		
	Max	12.6	43977	498	33210		
2013 Obs = 57	Min	1.4	1275	0	381		
	Mean	5.87	10945.77	76.22	4708.16		
	SD	2.7	7158.21	126.53	4700.33		
_	Max	13.2	30766	473	37665		
14 = 57	Min	1.4	1900	0	450		
20 Jbs	Mean	5.97	11885.12	76.6	4441.49		
0	SD	2.75	6108.68	124.51	4997.88		
~~~	Max	13	45200	325	23895		
15 = 48	Min	0.8	2476	0	501		
20 Jbs	Mean	6.3	13300.4	60.67	4165.05		
0	SD	2.88	8413.72	80.43	3769.98		

Table 2: Descriptive statistics: input and outputs

partments.

We ran four different DEA models, namely CCR, BCC, SBM (CRS) and SBM (VRS) for the data set in each year, using an output-oriented approach. The output-oriented radial Malmquist DEA model is applied to the full data set to examine the improvement in efficiency of the departments from 2013 to 2015.

# 4. Results

Results from the four DEA models are presented in Tables 3, 4 and 5.

In the output-oriented CCR models, which are obtained under the hypothesis of constant returns to scale, *DMUs* 5, 7, 28, and 57 were reported to be efficient in 2013; *DMUs* 7, 28, 32, and 39 in 2014; and *DMUs* 7, 18, 20, 28, 32, 33, 37, and 39 in 2015. These *DMUs* also achieved full efficiency in all the other three models, and, according to CCR models, had no slack. The remaining departments had efficiency scores less than 1, thus were less efficient. The average efficiency score across all 57 departments saw slight improvement from 2013 to 2015, which were 0.608, 0.608, and 0.673, respectively.

The output-oriented BCC models, which are built on the assumption of variable returns to scale, on the other hand, indicate more efficient departments, as BCC-efficiency scores are not less than that of the corresponding CCR models. The BCC-efficient DMUs that were reported to be inefficient in CCR models were DMUs 6, 8, 29, 32, and 37 in 2013, DMUs 5, 30, 33, 37, and 50 in 2014, and DMUs 23 and 57 in 2015. We can also see how departments can improve their performance by analyzing their slacks (output shortfalls in this output-oriented approach). For instance, if DMU 19 wanted to be fully efficient in 2015, it would have to raise its teaching load by approximately 1712 periods in that year (See Table 5). Moreover, both CCR and BCC models measure the "technical efficiency". The average technical efficiency values of the 57 departments were 0.657, 0.694, and 0.688 in 2013, 2014, and 2015, respectively.

	* CCR-0					BO	C-O			SBM	юс		SBM-O-V				
DMU*	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	
1	0.549	0.00	0.00	0.00	0.562	0.00	0.00	0.00	0.328	234.63	335.83	0.00	0.333	972.89	325.01	0.00	
2	0.332	0.00	0.00	0.00	0.368	0.00	1435	0.00	0.076	1827.41	470.10	1908.69	0.077	1920.54	466.11	1940.66	
3	0.194	0.00	4.05	0.00	0.207	0.00	650	0.00	0.036	1997.03	170.62	996.77	0.066	985.37	88.87	743.20	
4	0.317	0.00	0.00	0.00	0.332	0.00	0.00	0.00	0.189	1705.11	170.19	607.59	0.264	3754.60	82.39	1311.16	
5	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
6	0.868	0.00	0.00	0.00	1.000	0.00	0.00	0.00	0.401	0.00	26.91	0.00	1.000	0.00	0.00	0.00	
7	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
8	0.929	683.51	0.00	0.00	1.000	0.00	0.00	0.00	0.884	2796.57	0.00	974.57	1.000	0.00	0.00	0.00	
9	0.539	0.00	0.00	0.00	0.562	0.00	0.00	0.00	0.400	1034.36	387.85	0.00	0.400	111751	386.65	0.00	
10	0.766	0.00	0.00	0.00	0.779	0.00	0.00	0.00	0.443	0.00	89.29	0.00	0.490	0.00	73.92	0.00	
11	0.573	0.00	0.00	0.00	0.607	0.00	0.00	0.00	0.424	0.00	356.02	350.56	0.429	0.00	345.61	472.90	
12	0.636	0.00	0.00	0.00	0.656	0.00	0.00	0.00	0.210	0.00	249.63	0.00	0.211	0.00	248.64	0.00	
13	0.837	0.00	0.00	0.00	0.881	0.00	0.00	0.00	0.785	0.00	165.65	0.00	0.827	0.00	126.43	0.00	
14	0.520	0.00	0.00	0.00	0.522	0.00	0.00	0.00	0.215	0.00	234.32	259.53	0.237	0.00	200.32	770.81	
15	0.268	0.00	0.00	0.00	0.327	0.00	33.86	1340.14	0.046	3101.09	503.53	2821.31	0.047	1777.17	489.31	3116.67	
16	0.809	0.00	0.00	0.00	0.844	0.00	0.00	0.00	0.646	0.00	145.64	0.00	0.671	0.00	129.92	0.00	
17	0.541	0.00	0.00	0.00	0.561	0.00	0.00	0.00	0.208	0.00	327.01	0.00	0.211	0.00	322.05	0.00	
18	0.653	0.00	0.00	0.00	0.679	0.00	0.00	0.00	0.552	0.00	282.37	0.00	0.553	0.00	281.58	0.00	
19	0.849	0.00	0.00	0.00	0.920	0.00	0.00	0.00	0.763	0.00	153.67	0.00	0.853	0.00	82.12	93.50	
20	0.515	0.00	0.00	0.00	0.591	0.00	0.00	551.17	0.288	0.00	126.96	558.87	0.375	779.67	59.85	1333.32	
21	0.347	0.00	0.00	000	0.348	0.00	0.00	000	0.121	1436.63	25/83	801.47	0.148	2927.15	193.98	1313.16	
22	0538	0.00	338	0.00	0.646	0.00	80.09	0.00	0.024	0.00	5/6.83	0.00	0.037	0.00	377.05	1114.83	
23	0.672	0.00	0.00	0.00	0.696	0.00	0.00	0.00	0.073	0.00	258.86	0.00	0.074	0.00	255.05	0.00	
24	0.160	0.00	0.00	0.00	0.162	0.00	0.00	0.00	0.059	5153.63	264.15	000	0.071	7057.55	200.30	1639.16	
25	0.330	0.00	0.00	0.00	0.338	0.00	690	0.00	0.132	10905.75	390.24	0.00	0.170	/05/35	229.50	0.00	
20	0.706	0.00	0.00	0.00	0./82	0.00	0.00	1182.55	0.505	527.26	298.62	9090.59	0.550	690.92	238.50	9021.20	
2/	1,000	0.00	0.00	0.00	1,000	0.00	0.00	000	1,000	000	0.00	000	1,000	000	0.00	002129	
20	0.968	7030.99	0.00	0.00	1,000	0.00	0.00	0.00	0.694	8904.05	0.00	0.00	1.000	0.00	0.00	0.00	
30	0.530	000	0.00	0.00	0.596	0.00	1629	0.00	0.104	1130.05	573.05	0.00	0.155	0.00	365.08	0.00	
31	0.073	0.00	1083	000	0.390	0.00	4495	147.53	0.025	000	262.60	59.92	0.031	0.00	20895	690.74	
32	0.982	0.00	000	0.00	1,000	0.00	000	000	0.946	000	13.37	000	1000	0.00	000	000	
33	0.599	0.00	0.00	0.00	0728	0.00	3.28	542.16	0.065	000	94.89	189.77	0141	0.00	3781	104797	
34	0.883	0.00	0.00	0.00	0.940	2219.03	0.00	000	0.822	3945.47	0.00	000	0.826	384460	000	000	
35	0,308	0.00	0.00	0.00	0.319	0.00	0.00	0.00	0.110	5030.66	422.56	686.91	0.113	5403.26	406.60	814.82	
36	0.443	0.00	0.00	0.00	0.451	0.00	0.00	0.00	0.200	2466.49	332.58	0.00	0.209	3399.46	311.14	0.00	
37	0.875	0.00	0.00	0.00	1.000	0.00	0.00	0.00	0.671	0.00	196.58	0.00	1.000	0.00	0.00	0.00	
38	0.429	0.00	0.00	0.00	0.432	0.00	0.00	0.00	0.095	186.71	274.71	323.79	0.116	0.00	219.89	306.76	
39	0.857	0.00	0.00	0.00	0.892	0.00	0.00	0.00	0.846	2347.90	0.00	2861.02	0.871	0.00	0.00	3281.67	
40	0.418	0.00	0.00	0.00	0.437	1482.86	0.00	0.00	0.246	4285.53	230.20	0.00	0.261	5909.15	192.52	0.00	
41	0.671	0.00	0.00	0.00	0.779	2077.87	0.00	0.00	0.642	6592.62	0.00	0.00	0.668	5881.40	0.00	0.00	
42	0.576	3349.16	0.00	0.00	0.607	0.00	0.00	0.00	0.407	27127.65	0.00	0.00	0.458	21034.21	0.00	856.53	
43	0.644	0.00	0.00	0.00	0.656	0.00	0.00	0.00	0.602	0.00	163.12	0.00	0.613	976.74	12.54	3493.21	
44	0.497	0.00	2.27	0.00	0.505	0.00	3.03	0.00	0.036	0.00	179.26	0.00	0.053	1443.38	116.92	0.00	
45	0.508	0.00	0.00	0.00	0.513	0.00	0.00	0.00	0.306	0.00	295.52	121.42	0.330	915.28	253.61	476.32	
46	0.346	0.00	0.00	0.00	0.370	0.00	0.00	521.19	0.087	0.00	297.86	1798.12	0.098	955.60	250.60	2221.16	
47	0.786	0.00	0.00	0.00	0.803	0.00	0.00	0.00	0.125	0.00	118.97	0.00	0.128	0.00	115.30	0.00	
48	0.613	0.00	0.00	0.00	0.631	0.00	0.00	0.00	0.291	0.00	245.81	6.05	0.300	0.00	233.32	193.83	
49	0.455	0.00	0.00	0.00	0.473	0.00	0.00	0.00	0.147	6257.76	483.25	0.00	0.167	4302.24	420.06	0.00	
50	0.244	0.00	1.26	0.00	1.000	0.06	0.00	0.00	0.060	1231.60	99.53	308.70	1.000	0.02	0.00	0.00	
51	0.363	0.00	0.00	0.00	0.382	0.00	0.00	0.00	0.047	7864.04	557.04	0.00	0.055	0.00	481.66	0.00	
52	0.579	0.00	7.69	0.00	0.659	0.00	50.06	0.00	0.015	0.00	443.38	0.00	0.016	0.00	405.29	0.00	
53	0.542	0.00	0.00	0.00	0.584	0.00	7.48	0.00	0.076	3674.78	51639	0.00	0.105	0.00	367.77	0.00	
54	0.407	0.00	0.00	0.00	0.412	0.00	0.00	0.00	0.143	0.00	266.48	1333.69	0.158	414.00	226.73	1806.82	
55	0.640	0.00	0.00	0.00	0.669	0.00	0.00	1.27	0.204	0.00	110.99	209.41	0.286	0.00	65.14	898.86	
56	0.693	0.00	1.13	0.00	0.698	0.00	1.51	0.00	0.057	0.00	110.01	0.00	0.080	1076.47	75.63	0.00	
57	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	

Table 3: Estimated efficiency scores and output slacks for DMUs in 2013

Notes: \*DMU I to 57 are Insurance, Information Technology, Population, Valuation, Political Revolution Roadmap of the Communist Party of Vietnam, Management Information Systems, Managerial Accounting, Financial Accounting, Auditing, Real Estate Business, International Business, Public Economics, Investment Economics, Human Resource Economics, Agricultural Economics and Rural Development, Development Economics, Commercial Economics and Business, Real Estate Business, and European Economics, and Rural Development, Development Economics, Commercial Economics, Investment Economics, Matural Development, Conomics, Guatarata, Administration, Urban economics, Microeconomics, Macroeconomics, Economic History, Monetary and Financial Theories, Marketing, Commercial Bank, Non-specialized Foreign Language, Accounting Principles, Basic principles of Marxism – Leninism, Basic Law, Business Law, Guanagement, Fravel and Tourism Management, Interprise Management, Hospitality Management, General Business Management, Human Resource Management, Public Finance, Corporate Finance, International Finance, Sociology, Stock Market, Business Statistics, Socio-Economic Statistics, International trade, Business English, Vietnamese and Linguistic Theories, Economic Informatica, Basic Mathematica, Mathematical Economics, Mathematical Finance, Narketing Communications, Ho Chi Minh Ideology, Business Culture Department, respectively.

S+(1), S+(2), and S+(3) are shortage of research hours, graduates number, and teaching load output, respectively.

	CCRO					BC	co			SBM	юс		SBM-O-V				
DMU	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	
1	0.457	0.00	0.00	0.00	0.481	0.00	0.00	000	0.338	124.13	32796	000	0.338	124.13	32796	0.00	
2	0334	0.00	45.45	439930	0.402	0.00	24.17	3131.46	0.034	147225	446.65	2310.79	0.034	1472.25	446.65	2310.79	
3	0,590	0.00	21.06	206927	0.623	000	9.70	101826	0.015	0.00	6561	54561	0.015	000	6561	545.61	
4	0714	000	000	237798	0.779	000	000	128581	0468	000	51.39	74488	0468	000	51.39	74488	
5	0903	000	2239	000	1000	000	000	000	0999	000	000	000	0999	000	000	000	
6	0576	000	300	000	0721	000	000	000	0340	000	4847	000	0340	000	4847	000	
7	1000	000	000	000	1000	000	000	000	1000	000	000	000	1000	000	000	000	
,	0965	000	000	000	0047	22808	000	202014	0787	100806	000	2782.24	0787	100896	000	2782.24	
0	0517	000	000	000	0.047	000	000	200014	0.787	000	26082	071 22	0.787	000	36082	07122	
10	0517	000	000	000	0.000	000	000	000	0512	216266	7015	9/135	0512	216266	7015	000	
10	0.523	000	000	222054	0.00	000	000	2020.10	0.313	21000	20272	207004	0.315	21000	202.72	207004	
10	0.0010	000	2762	323034	0.010	000	25.44	250279	0.30/	000	70.17	2079.04	0.307	000	70.17	2079.04	
12	0.510	000	2/02	2023.79	0.50	000	2.3:++	200./8	0.594	1044.29	75.17	2450.54	0.594	104428	75.17	2450.54	
14	0.00	000	000	172572	0.032	000	000	156051	0.304	000	2/99/	75496	0.304	000	2/99/	75496	
14	0.000	000	6190	202200	0.644	000	2512	252802	0.555	1220	452.04	202020	0.555	1220	452.04	7,54,60	
15	0.080	000	51.89	395800	04/8	000	23.12	232895	0002	12.20	45504	200808	0002	1220	45504	200000	
16	0.676	000	14.00	150217	08/9	000	000	145803	0./38	000	01.04	118962	0./38	000	01.04	118962	
1/	0454	000	1400	1595.06	0529	000	208	90080	0.13/	000	36/34	83622	0.03/	000	36/34	83622	
18	0.820	000	000	30938/	029/	000	000	339/60	0632	000	/533	404368	0632	000	/533	4043.68	
20	0,400	000	000	151990	0.75	000	000	2014.10	0.510	000	2824	21004	0.510	000	2824	21004	
20	0.407	000	000	2244.41	0.//5	000	000	10// 00	0100	000	21029	21004	0.100	000	21029	21004	
21	0407	000	6200	201795	0742	000	11.42	10495	0.199	000	21050	201724	0.199	000	21000	201724	
22	0.317	000	4002	209720	0./42	000	1521	19460	0.100	000	2//30	142690	0.100	000	2//30	142690	
20	0.192	000	4095	000	0226	000	000	000	0.072	2025 10	15220	12024	0.072	2025 10	152.20	12024	
24	0.260	000	35.58	000	0.483	000	000	000	0.083	118500	45904	764333	0.083	118500	45904	264333	
26	0651	000	000	301318	0832	000	000	205982	0449	000	15728	311696	0449	000	15728	311696	
20	0622	248022	000	000	0660	257159	000	33084	0525	7493.09	000	852605	0525	7493.09	000	852605	
28	1000	000	000	000	1000	000	000	000	1000	000	000	000	1000	000	000	000	
29	0917	494295	000	000	0983	460979	000	000	0847	471853	000	62160	0847	471853	000	62160	
30	0.689	000	5020	1920.78	1,000	000	000	000	1,000	000	000	000	1,000	000	000	000	
31	0558	000	4913	129823	0586	000	4380	112170	0004	000	27240	56488	0004	000	27240	56488	
32	1,000	000	000	000	1,000	000	000	000	1,000	000	000	000	1,000	000	000	000	
33	0914	0.00	1360	125033	1.000	000	000	002	1.000	000	0.00	002	1.000	000	000	0.02	
34	0877	0.00	0.00	000	0965	000	000	000	0.925	213232	0.00	000	0.925	213232	000	0.00	
35	0378	0.00	0.00	3934.17	0,422	0.00	0.00	3508.03	0.162	20645	373.42	1559.67	0.162	20645	373.42	1559.67	
36	0.773	0.00	000	380829	0.806	000	000	375663	0.452	0.00	12695	2913,44	0.452	000	12695	2913.44	
37	0.712	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
38	0.842	0.00	11.64	1764.64	0.863	0.00	4.78	1117.10	0.466	0.00	47.63	82759	0.466	0.00	47.63	82759	
39	1.000	0.00	0.00	0.00	1.000	0.00	0.00	000	1.000	0.00	0.00	000	1.000	000	0.00	0.00	
40	0342	0.00	000	0.00	0.368	0.00	000	0.00	0.263	561055	233.54	000	0.263	561055	233.54	0.00	
41	0.688	0.00	000	0.00	0.885	0.00	000	0.00	0.759	5071.57	0.00	000	0.759	5071.57	000	0.00	
42	0.605	0.00	0.00	0.00	0.755	000	000	759.07	0579	0.00	0.00	1219590	0579	000	0.00	1219590	
43	0518	1097.74	0.00	0.00	0.677	1608.77	000	0.00	0.384	10643.61	0.00	953.14	0.384	10643.61	0.00	953.14	
44	0509	0.00	1886	000	0.657	000	193	000	0.025	0.00	38,44	000	0.025	000	38,44	0.00	
45	0529	0.00	000	3933.81	0535	0.00	000	3678.18	0233	0.00	20658	1624.88	0233	000	20658	1624.88	
46	0.384	0.00	37.74	43858	0.388	000	3450	166.05	0.004	000	246.79	52.83	0.004	000	246.79	52.83	
47	0.499	0.00	193	0.00	0.529	0.00	000	000	0.157	0.00	19827	000	0.157	0.00	19827	0.00	
48	0.380	0.00	0.00	0.00	0.401	000	000	0.00	0302	3982.39	232.07	000	0.302	3982.39	232.07	0.00	
49	0352	0.00	0.00	0.00	0.408	0.00	0.00	0.00	0.137	6069.00	44855	56433	0.137	6069.00	44855	56433	
50	0298	0.00	1228	428.79	1.000	000	000	000	0.999	0.00	0.00	000	0.999	000	000	0.00	
51	0.384	0.00	0.00	000	0.419	000	000	000	0.139	502550	442.49	614.75	0.139	502550	442.49	614.75	
52	0497	0.00	4925	0.00	0569	000	3338	000	0.002	0.00	420.86	000	0.002	000	420.86	0.00	
53	0414	0.00	000	0.00	0500	000	000	000	0.146	3621.49	423.42	000	0.146	3621.49	423.42	0.00	
54	0362	0.00	000	224331	0378	000	000	155087	0.283	303646	0.00	5381.84	0.283	303646	000	5381.84	
55	0.635	0.00	655	2261.39	0.674	000	0.00	1250.79	0.293	0.00	6735	61733	0.293	000	6735	61733	
56	0569	0.00	1733	0.00	0.664	000	1.44	000	0.014	0.00	6836	000	0.014	000	6836	0.00	
57	0.729	0.00	1469	295921	0.744	0.00	854	2361.56	0259	0.00	93.01	1513.60	0259	0.00	93.01	1513.60	

Table 4: Estimated efficiency scores and output slacks for DMUs in 2014

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		α	RO			BC	<b>x</b> •0			SBM	ос		SBM-O-V				
DMU*	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	Score	S+(1)	S+(2)	S+(3)	
1	0.486	0.00	0.00	0.00	0.488	0.00	0.00	0.00	0.411	0.00	161.59	2829.50	0.413	0.00	159.08	2895.67	
2	0.254	0.00	0.00	0.00	0.266	0.00	0.00	0.00	0.129	1663.06	224.99	3473.68	0.132	2402.04	217.67	3536.92	
3	0.194	0.00	10.58	0.00	0.200	0.00	17.90	0.00	0.063	1019.63	94.92	1680.47	0.074	2952.38	75.77	1845.88	
4	0545	0.00	0.00	0.00	0.557	0.00	0.00	0.00	0.439	0.00	40.32	1583.62	0.446	0.00	41.18	1455.75	
6	0.894	0.00	933	0.00	0.896	0.00	6.88	0.00	0.377	0.00	28.05	0.00	0.404	0.00	25.09	0.00	
7	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
8	0.971	0.00	0.00	0.00	0.972	0.00	0.00	0.00	0.937	1034.61	0.00	91592	0.938	848.84	0.00	948.59	
9	0.658	0.00	0.00	0.00	0.662	0.00	0.00	0.00	0.639	0.00	135.94	945.03	0.640	0.00	133.81	1001.24	
10	0.589	0.00	0.00	0.00	0.595	0.00	0.00	0.00	0.484	0.00	76.07	420.77	0.497	0.00	7036	571.46	
11	0.550	0.00	0.00	0.00	0.551	0.00	0.00	0.00	0.533	0.00	126.54	1587.10	0536	0.00	123.13	1677.10	
13	0.693	0.00	0.00	0.00	0.695	0.00	0.00	0.00	0.682	0.00	81.41	3603.51	0.683	0.00	102.56	2732.38	
14	0.403	0.00	0.00	0.00	0.405	0.00	0.00	0.00	0.395	12789.00	43.33	1917.53	0.398	13101.85	36.05	2054.62	
15	0.415	0.00	0.00	0.00	0.460	0.00	0.00	0.00	0.209	0.00	206.34	3687.19	0.210	0.00	204.33	3740.25	
17	0.643	0.00	0.00	0.00	0.714	0.00	0.00	0.00	0.380	0.00	11735	2350.96	0.384	0.00	113.93	2441.04	
18	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
19	0.562	0.00	0.00	0.00	0.575	0.00	0.00	171231	0.446	16641.70	0.00	6907.11	0.448	22754.98	0.00	5227.62	
20	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
21	0.376	0.00	0.00	0.00	0.378	0.00	0.00	0.00	0.291	0.00	11091	2222.82	0.294	1327	105.86	2353.56	
22	0.588	0.00	42.59	0.00	0.670	0.00	0.00	0.00	0.092	0.00	257.80	0.00	0.092	0.00	255.01	474.70	
23	0.911	0.00	30.77	0.00	1.000	0.00	0.00	0.00	0.210	0.00	97.80	0.00	1.000	0.00	0.00	0.00	
24	0.487	0.00	1.67	0.00	0.501	0.00	10.46	0.00	0.154	0.00	85.95	43634	0.161	0.00	81.74	419.31	
25	0.934	0.00	0.00	0.00	0.945	0.00	0.00	0.00	0.820	0.00	15.79	0.00	0.847	0.00	13.06	0.00	
27	0.707	0.00	0.00	0.00	0.708	0.00	0.00	0.00	0.682	16049.11	0.00	1789.40	0.683	15783.03	0.00	1824.53	
28	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
29	0.865	97.83	0.00	0.00	0.865	128.78	0.00	0.00	0.674	14908.59	0.00	0.00	0.682	14375.66	0.00	0.00	
31	0.699	0.00	19.43	0.00	0.735	0.00	2858	0.00	0.053	0.00	125.12	0.00	0.054	0.00	121.77	0.00	
32	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
33	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
34	0.785	0.00	0.00	0.00	0.788	0.00	0.00	0.00	0.607	15852.19	0.00	0.00	0.608	1578431	0.00	0.00	
35	0.398	0.00	0.00	0.00	0.415	0.00	0.00	0.00	0.270	0.00	147.80	3272.75	0.272	0.00	144.16	3368.86	
36	0.496	0.00	0.00	0.00	0.496	0.00	0.00	0.00	0.388	0.00	141.42	1833.73	0.392	0.00	137.90	1926.66	
37	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
39	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	1.000	0.00	0.00	0.00	
40	0.693	0.00	0.00	0.00	0.699	0.00	0.00	0.00	0.544	12508.42	17.22	0.00	0.546	1276629	13.72	0.00	
41	0.531	0.00	0.00	0.00	0.533	0.00	0.00	0.00	0.495	13245.00	41.76	1130.60	0.499	13531.78	35.08	1256.26	
42	0.787	0.00	0.00	0.00	0.855	0.00	0.00	398.13	0.744	15056.45	0.00	2200.68	0.748	14735.34	0.00	2161.01	
43	0.737	0.00	0.00	0.00	0.751	0.00	0.00	0.00	0.685	3223.45	0.00	2273.54	0.688	3267.15	0.00	2212.78	
45	0.530	0.00	0.00	0.00	0.543	0.00	0.00	0.00	0.374	0.00	90.56	1909.11	0.380	0.00	85.53	2041.95	
47	0.402	0.00	15.67	0.00	0.402	0.00	15.16	0.00	0.101	0.00	16733	58721	0.105	7247	15757	1585.83	
49	0553	0.00	26.81	0.00	0.620	0.00	0.00	0.00	0.161	0.00	271.06	0.00	0.162	0.00	267.46	613.32	
50	0.546	0.00	6.79	0.00	0.549	0.00	7.48	0.00	0.158	0.00	36.03	28831	0.202	42850	2436	705.19	
51	0.476	0.00	28.71	0.00	0.491	0.00	17.68	0.00	0.029	0.00	26627	100,47	0.029	313.45	264.32	0.00	
52	0.673	0.00	34.88	0.00	0.697	0.00	21.84	0.00	0.033	0.00	201.66	0.00	0.033	0.00	201.56	0.00	
53	0.672	0.00	1554	0.00	0.722	0.00	0.00	0.00	0.219	0.00	21424	0.00	0.224	0.00	20/90	0.00	
54	0.259	0.00	1604	0.00	0.264	0.00	0.00	0.00	0.239	1460.50	140.42	2/92.6/	0.242	2824.79	12690	2909:42	
30 54	0./30	0.00	10.04	0.00	0.762	0.00	8.80	0.00	0.181	0.00	4995	1307.42	0.18/	0.00	20.95	300.92	
50	0.000	0.00	14:44	542.24	1,000	0.00	000	0.00	0.075	0.00	15.50	60000	1,000	0.00	000	0.00	
3/	0.200	0.00	1,500	JHJ24	1.000	0.00	0.00	0.00	0.0/0	0.00	15.50	00250	1,000	0.00	0.00	0.00	

Table 5: Estimated efficiency scores and output slacks for DMUs in 2015

Notes: \*There are only 48 DMUs in the sample of the year 2015. Nine missing departments due to incomplete data consist of DMUs 5, 12, 16, 26, 30, 38, 44, 46, and 48.

There are, in fact, two types of measures or approaches in DEA: radial and non-radial. The two previous models, under weak efficiency, evaluate only the radial or proportional efficiency, but do not take account of the output shortfalls that are represented by non-zero slacks. On the contrary, the non-radial approach, which is represented by SBM, deals with slacks directly in the objective function and reflects non-zero slacks in outputs when they are present. The SBM model is perhaps more suitable and practical in this case when all the variables employed seem to be non-radial. Furthermore, SBM accounts for all inefficiencies instead of accounting only for purely technical inefficiencies as did the previous models, therefore it measures "mix efficiency". We ran the output-oriented SBM (which puts emphasis on the output shortfalls) under both the constant and variable returns-to-scale assumptions.

It is apparent from the results that a DMU is SBM-efficient under constant returns-to-scale if and only if it is CCR-efficient, and similarly, a DMU is SBM-efficient under variable returnsto-scale if and only if it is BCC-efficient, following strictly Tone's theorem (Tone and Cooper, 1997). In the SBM, the efficiency scores of the less efficient DMUs, however, were significantly lower than those in the CCR or BCC model, which leads to large differences among DMUs. This is because the SBM model takes into account slacks for the less efficient DMUs. The general efficiency score in the SBM, thus, was very low; for example, the figures for SBM models under constant and variable returns-toscale in 2015 were respectively just 0.479 and 0.509. Nonetheless, in the output-oriented SBM models, we can analyze the output slacks of the less efficient *DMU*s in order to explain why they did not reach the efficient positions and how they could improve their positions. *DMU* 1, for instance, performed well in research in 2015, but proved to be less efficient in training under both the CRS and VRS assumptions. In fact, it would have to increase its teaching load by 2896 periods a year and produce 159 graduating students more to reach the highest efficiency level in 2015 (according to the result of the SBM (VRS) model). Meanwhile, *DMU* 2, however, needed to expand all of their outputs to be an efficient one in 2015, especially the research and teaching load outputs (see Table 5).

The Malmquist results shown in Table 6 allow us to evaluate the change in efficiency scores and the technological change as well as the total factor productivity (TFP) change of the DMUs over the period. Due to some missing data in 2015 as mentioned above, there were 48 DMUs being analyzed using the output-oriented Malmquist DEA model. In each year (period), the first column ('firm') shows the DMUs' names, the second ('effch') reflects technical efficiency change (catch-up effect), the third ('techch') reflects technology change (frontier-shift effect), the fourth ('pech') reflects pure technical efficiency change, the fifth ('sech') reflects scale efficiency change, and the last ('tfpch') indicates total factor productivity change or Malmquist index, which is a combination of the catch-up and the frontier-shift effects. As can be seen from the results, in terms of catch-up effect, a significant proportion of DMUs (66.67%) made progress in efficiency, that is their catch-up value is higher than 1, from 2013 to 2015. The average improvement rate in the technical efficiency of 48 de-

Table 6: Results of the output-oriented Malmquist DEA model applied to the data set in
three years, 2013-2015

	MALMQUIST INDEX SUMMARY												MALMQUIST INDEX SUMMARY OF						
year = 2 year = 3 FIRM MEANS																			
firm	effch	techch	pech	sech	tfpch	firm	effch	techch	pech	sech	tfpch	firm	effch	techch	pech	sech 1	tfpch		
1	0.819	1.126	0.856	0.957	0.922	1	1.063	0.864	1.013	1.049	0.918	1	0.933	0.986	0.931	1.002	0.920		
2	0.997	0.970	1.182	0.843	0.967	2	0.760	0.954	0.613	1.241	0.725	2	0.871	0.962	0.851	1.023	0.837		
3	3.004	0.896	2.998	1.002	2.692	3	0.329	1.051	0.321	1.028	0.346	3	0.995	0.971	0.980	1.015	0.965		
4	2.231	1.002	2.306	0.967	2.235	4	0.763	1.004	0.715	1.067	0.766	4	1.304	1.003	1.284	1.016	1.308		
5	0.652	1.144	0.721	0.904	0.746	5	1.552	0.811	1.243	1.248	1.258	5	1.006	0.963	0.947	1.062	0.968		
6	1.000	0.933	1.000	1.000	0.933	6	1.000	0.579	1.000	1.000	0.579	6	1.000	0.735	1.000	1.000	0.735		
7	0.931	0.953	0.947	0.983	0.886	7	1.123	0.643	1.026	1.094	0.722	7	1.022	0.783	0.986	1.037	0.800		
8	0.946	1.106	1.071	0.883	1.046	8	1.272	0.782	1.099	1.158	0.994	8	1.097	0.930	1.085	1.011	1.020		
9	0.677	1.158	0.782	0.866	0.784	9	1.121	0.803	0.944	1.187	0.900	9	0.871	0.964	0.859	1.013	0.840		
10	0.920	1.027	1.017	0.905	0.945	10	1.036	0.867	0.894	1.159	0.898	10	0.976	0.944	0.953	1.024	0.921		
11	0.691	1.070	0.740	0.934	0.739	11	1.183	0.802	1.066	1.110	0.949	11	0.904	0.926	0.888	1.018	0.838		
12	1.217	1.062	1.216	1.000	1.292	12	0.631	0.844	0.629	1.004	0.533	12	0.876	0.947	0.875	1.002	0.830		
13	1.422	0.934	1.600	0.889	1.328	13	1.083	1.026	0.878	1.233	1.111	13	1.241	0.979	1.186	1.047	1.215		
14	0.830	1.123	1.001	0.829	0.932	14	1.415	0.986	1.270	1.114	1.395	14	1.084	1.052	1.127	0.961	1.140		
15	1.246	1.024	1.323	0.942	1.275	15	1.219	0.940	1.114	1.094	1.145	15	1.232	0.981	1.214	1.015	1.209		
16	0.692	1.020	0.812	0.852	0.706	16	0.951	0.755	0.770	1.235	0.718	16	0.811	0.878	0.791	1.026	0.712		
17	1.355	0.952	1.312	1.033	1.290	17	1.430	1.063	1.290	1.109	1.521	17	1.392	1.006	1.301	1.070	1.400		
18	1.162	1.055	1.169	0.993	1.226	18	0.925	0.981	0.911	1.015	0.907	18	1.037	1.017	1.032	1.004	1.054		
19	0.942	1.077	1.363	0.691	1.014	19	1.138	0.901	0.760	1.496	1.025	19	1.035	0.985	1.018	1.017	1.020		
20	0.700	1.113	0.948	0.739	0.779	20	1.867	0.922	1.510	1.236	1.721	20	1.143	1.013	1.196	0.956	1.158		
21	1.716	1.121	1.924	0.892	1.924	21	1.724	0.891	1.532	1.125	1.535	21	1.720	0.999	1.717	1.002	1.719		
22	1.005	1.103	1.596	0.630	1.109	22	2.769	0.805	1.750	1.582	2.229	22	1.668	0.942	1.671	0.998	1.572		
23	0.919	0.952	0.868	1.058	0.874	23	1.137	0.675	1.073	1.060	0.767	23	1.022	0.801	0.965	1.059	0.819		
24	1.000	1.062	1.000	1.000	1.062	24	1.000	0.671	1.000	1.000	0.671	24	1.000	0.844	1.000	1.000	0.844		
25	0.948	0.957	0.983	0.964	0.907	25	0.943	0.666	0.880	1.072	0.628	25	0.945	0.798	0.930	1.016	0.755		
26	0.823	0.985	0.775	1.061	0.811	26	1.252	0.939	1.232	1.016	1.176	26	1.015	0.962	0.977	1.039	0.976		
27	1.000	1.152	1.000	1.000	1.152	27	1.000	0.847	1.000	1.000	0.847	27	1.000	0.988	1.000	1.000	0.988		
28	1.517	0.909	1.373	1.104	1.379	28	1.094	1.008	1.000	1.094	1.103	28	1.288	0.957	1.172	1.099	1.233		
29	0.990	1.024	0.979	1.011	1.014	29	0.895	0.715	0.817	1.096	0.640	29	0.942	0.856	0.895	1.053	0.806		
30	1.198	1.053	1.353	0.885	1.261	30	1.055	1.009	0.962	1.096	1.064	30	1.124	1.031	1.141	0.985	1.159		
31	1.707	1.035	1.769	0.965	1.766	31	0.642	0.948	0.616	1.042	0.608	31	1.046	0.990	1.044	1.003	1.036		
32	0.806	1.085	1.000	0.806	0.874	32	1.405	0.836	1.000	1.405	1.174	32	1.064	0.952	1.000	1.064	1.013		
33	1.155	0.997	1.121	1.030	1.151	33	1.000	0.648	1.000	1.000	0.648	33	1.075	0.804	1.059	1.015	0.864		
34	0.808	1.063	0.794	1.017	0.859	34	2.028	0.733	1.899	1.068	1.487	34	1.280	0.883	1.228	1.042	1.130		
35	1.011	1.033	1.100	0.920	1.045	35	0.772	0.745	0.603	1.281	0.576	35	0.884	0.878	0.814	1.085	0.775		
36	1.050	0.990	1.244	0.844	1.040	36	1.302	0.704	1.132	1.150	0.916	36	1.169	0.835	1.187	0.985	0.976		
37	0.799	1.022	1.014	0.788	0.816	37	1.422	0.741	1.109	1.282	1.054	37	1.066	0.870	1.060	1.005	0.928		
38	1.030	1.016	1.038	0.992	1.047	38	1.001	1.034	1.014	0.987	1.035	38	1.016	1.025	1.026	0.990	1.041		
39	0.612	1.110	0.635	0.964	0.680	39	0.806	0.805	0.760	1.060	0.649	39	0.703	0.945	0.695	1.011	0.664		
40	0.753	1.091	0.890	0.846	0.821	40	1.570	0.788	1.472	1.066	1.237	40	1.087	0.927	1.144	0.950	1.008		
41	1.187	1.077	1.000	1.187	1.279	41	1.833	1.004	0.549	3.342	1.841	41	1.475	1.040	0.741	1.992	1.535		
42	1.021	1.105	1.107	0.922	1.128	42	1.240	0.777	1.161	1.068	0.964	42	1.125	0.927	1.133	0.992	1.042		
43	0.837	1.115	0.899	0.932	0.933	43	1.353	0.798	1.178	1.149	1.080	43	1.065	0.943	1.029	1.035	1.004		
44	0.736	1.097	0.900	0.818	0.807	44	1.622	0.772	1.374	1.181	1.252	44	1.093	0.920	1.112	0.983	1.005		
45	0.889	0.933	0.916	0.970	0.829	45	0.714	0.908	0.698	1.024	0.649	45	0.797	0.921	0.800	0.997	0.734		
46	0.987	0.911	1.008	0.979	0.899	46	1.159	1.116	1.130	1.025	1.293	46	1.070	1.008	1.068	1.002	1.078		
47	0.792	1.097	0.872	0.908	0.869	47	1.064	0.797	0.907	1.173	0.848	47	0.918	0.935	0.889	1.032	0.858		
48	0.729	0.854	0.744	0.980	0.623	48	1.353	1.118	1.344	1.007	1.513	48	0.993	0.977	1.000	0.993	0.970		
mean	0.994	1.033	1.075	0.925	1.027	mean	1.122	0.845	0.979	1.147	0.949	mear	1.056	0.934	1.026	1.030	0.987		

Notes: \*Firm (DMU) 1 to 48 are Insurance, Information Technology, Population, Valuation, Management Information Systems, Managerial Accounting, Financial Accounting, Auditing, Real Estate Business, International Business, Investment Economics, Human Resource Economics, Agricultural Economics and Rural Development, (Natural Resources and) Environmental Economics and Management, International Economics, Commercial Economics and Business, Real Estate Business and Land Administration, Urban economics, Microeconomics, Macroeconomics, Economic History, Monetary and Financial Theories, Commercial Bank, Non-specialized Foreign Language, Accounting Principles, Basic Law, Business Law, Management of Technology, Economic Management, Social Management, Travel and Tourism Management, Enterprise Management, General Business Management, Human Resource Management, Public Finance, Corporate Finance, International Finance, Stock Market, Socio-Economic Statistics, Business English, Vietnamese and Linguistic Theories, Economic Informatics, Basic Mathematical Economics, Mathematical Finance, Marketing Communications, Ho Chi Minh Ideology, Business Culture Department, respectively. partments over the period was 5.6%. However, regarding the frontier-shift effect, there were only 9 *DMUs* that improved their frontier technology between 2013 and 2015, while 39 other *DMUs* demonstrated frontier-shift values less than 1, leading to the mean technological contribution to the *DMUs*' efficiency declined by 6.6% during the period. As a result, the average Malmquist index, which is a more comprehensive indicator (as mentioned above), was merely 0.987, with just 23 out of 48 DMUs, that is, nearly a half showed TFP growth in the period.

## 5. Discussion and concluding remarks

While the existing literature on the application of DEA in evaluating the efficiency of educational institutions is rich and plenteous, this method is still new for the education sector in Vietnam. Actually, we have found two studies applying this method to evaluating efficiency of Vietnamese higher education institutions, but so far this study is still the first one applying this method to examining the efficiency of academic departments within the same university.

As an illustrative example, this paper eval-

uates the performance of 57 departments of NEU. The data set used consists of one input (number of academic staff) and three outputs (number of research hours, number of graduates, and teaching load). We ran four different DEA models to investigate efficiency of the departments, then computed the Malmquist index to examine the improvement in efficiency of the departments from 2013 to 2015.

This study reveals some clear policy-making implications. First, the results provide a deeper insight into the current status of teaching and research activities of the department (reflected in the efficiency scores, output-slacks, and Malmquist index). Second, the information about output-slacks is especially useful for departments to improve their efficiency position (in terms of output expansion). Thus, such information helps departments adjust their development plan in a more appropriate way. Last but not least, to fully exploit the benefits of this method for the purpose of efficient resource allocation, we wish the data on all activities of the institution and its departments should be available and up-to-date.

### Notes:

1. "Weak efficiency" satisfies the condition  $\theta^* = 1$ ; "strong efficiency" satisfies two conditions:  $\theta^* = 1$  and all slacks are zero (Cooper et al., 2007).

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