An investigation of the undesirable outputs congestion in data envelopment analysis

Una investigación de la congestión de productos no deseados en el análisis de envolvente de datos

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ABSTRACT

Congestion is one of the most important subjects in Data Envelopment Analysis (DEA) which helps the Decision Maker (DM) to decide about changing the size of units. The estimation of congestion has attractive advantages from different perspectives. For example, the total cost of a particular DMU, in which the congestion occurs, can be reduced by the decreases in inputs. On the other hand, the output of units can be increased by the recognizing and eliminating the congestion of DMUs and so, the total profit of decision making units can be increased. Hence, the management is eager to know how to recognize and eliminate the congestion of units. Most of the existing methods to estimation of the congestion in the literature consider only the desirable outputs. This study focuses on the evaluation of congestion in the presence of undesirable outputs and proposes an approach to recognize the congestion of units. The method is demonstrated on a numerical example to illustrate the validity of the proposed method.

Keyword: Data Envelopment Analysis; Congestion; Undesirable outputs; Decision making units.

RESUMEN

La congestión es uno de los temas más importantes en el análisis envolvente de datos (DEA) que ayuda al responsable de la toma de decisiones (DM) a decidir sobre el cambio de tamaño de las unidades. La estimación de la congestión tiene atractivas ventajas desde diferentes perspectivas. Por ejemplo, el costo total de una DMU en particular, en la que ocurre la congestión, puede reducirse mediante la disminución de los insumos. Por otro lado, la producción de unidades se puede incrementar reconociendo y eliminando la congestión de las DMU y así, se puede incrementar el beneficio total de las unidades de toma de decisiones. Por lo tanto, la gerencia está ansiosa por saber cómo reconocer y eliminar la congestión de unidades. La mayoría de los métodos existentes para estimar la congestión en la literatura consideran solo los resultados deseables. Este estudio se centra en la evaluación de la congestión en presencia de salidas no deseadas.
indeseables y propone un enfoque para reconocer la congestión de unidades. El método se demuestra en un ejemplo numérico para ilustrar la validez del método propuesto.

**Palabra clave:** Análisis envolvente de datos; Congestión; Salidas indeseables; Unidades de toma de decisiones.

### 1. INTRODUCCIÓN

Data Envelopment Analysis (DEA) is a non-parametric methodology for assessing the relative efficiency of Decision Making Units (DMUs) with multiple inputs and multiple outputs (Charnes et al. [1], Banker et al. [2]). The estimation of congestion is an attractive issue in DEA. If the increases (decreases) in inputs leads to the decreases (increases) in outputs, then congestion occurs. Many scholars have been attracted to the evaluation of congestion. Färe and Grosskopf [3] introduced the economic concept of congestion. Färe et al. [4] suggested a radial model to recognize the congestion. Cooper et al. [5] proposed a slack-based method (CTT model) to evaluate the congestion of units which distinguished the congested inputs and also, obtained the amount of congestion of inputs as the difference between the observed value and the expected value. Brocket et al. [6] developed CTT model to identify the congestion of units in Chinese industry.

Cooper et al. [7] proposed an additive model to evaluate the congestion of DMUs. Cooper et al. [8] presented a single-model method to estimate the congestion. Jahanshaloo and Khodabakhshi [9] proposed a method based on the combination of inputs. Khodabakhshi [10] developed an approach to identify the congestion of units in the presence of stochastic data. Wei and Yan [11] used the output-oriented additive model to propose the necessary and sufficient conditions about the existence of congestion. Tone and Sahoo [12] presented a slack-based method to evaluate the scale elasticity in the presence of congestion. They introduced two basic concepts, namely, weak congestion and strong congestion and demonstrated that the congestion could be recognized by the status with having negative production elasticity. Sueyoshi and Sekitani [13] proposed a modified method to measure the congestion of units in the presence of the existing of the alternative optimal solutions.

On the other hand, the traditional DEA considers each DMU as a black box. This means that, each unit has a single process, however, there are also many applications that consider each DMU as a network of processes (for more studies about Network DEA models, see Kao [14], Tone and Tsutsui [15], Lozano [16], Lozano and Khezri [17]). Wang [18] considered the two-stage process in banking industry to estimate the congestion of units. For more studies about congestion, see Asgharian et al. [19], Flegg and Allen [20]. Odeck [21] considered the impact of the congestion of inputs, such as fuel consumption and the total number of workers on the efficiency in the Norwegian bus industry.

Quan and Yan [22] pointed out that the heavy industry emits many perilous pollutants such as sulfur dioxide, nitrogen oxides, carbon monoxide and hydrocarbons which are as the undesirable outputs of production process. The final costs of reducing the emissions of pollutants such as sulfur dioxide, nitrogen oxides, carbon monoxide and hydrocarbons may be changed according to the different environmental activities. The current environmental regulations do not provide any conditions for achieving to the minimum total cost. Murty et al. [23] showed that there exists a significant relationship between the environmental efficiency and the protection of the climate resources.

It should be noted that, most of the existing methods for the estimation of congestion in the literature consider the situation that the production process only produces the desirable outputs and do not consider the situation that the undesirable outputs can be produced (see Cooper et al. [24]), Brocket et al. [25], Khoveyni et al. [26], Wei et al. [27] and Hossein zadeh Lotfi et al. [28] for more details). Wu et al.[29] applied the method of Seiford and Zhu [30] to address the undesirable outputs and then formulated an approach Combined with the study of Wei and Yan [27] to measure the congestion of units in the presence of the desirable and undesirable outputs. Given that, the reduction in the undesirable outputs can result in the increases in the desirable outputs, therefore, it is important to include the undesirable outputs in the evaluation of decision making units. This study focuses on the situation that the production process generates the undesirable outputs and proposes a method to identify the congestion of units.
The rest of this paper is organized as follows: Section 2 provides the preliminaries and reviews the single-model method of Khodabakhshi [10]. Section 3 proposes a method to recognize the congestion of units in the presence of undesirable outputs. Numerical example is provided in section 4. Section 5 concludes the paper.

2. PRELIMINARIES

One of the most attractive issues in DEA is the estimation of congestion of DMUs. The concept of congestion is widely used in various sciences such as medicine, traffic engineering, transportation and population. In economics, congestion occurs when the increases in inputs lead to the decreases in outputs, without improving other inputs or outputs. In reverse, congestion occurs when the decreases in inputs lead to the increases in outputs, without worsening other inputs or outputs. The congestion estimation and the suggestion of methods to eliminate the congestion of units have attractive advantages from different perspectives.

1. The total cost of a particular DMU, in which the congestion occurs, can be reduced by the decreases in congested inputs.
2. The output of units can be increased by the recognizing and eliminating the congestion of DMUs and so, the total profit of decision making units can be increased.

Therefore, the congestion assessment has many economic benefits and so, the DM is eager to know how to recognize and eliminate the congestion of units. In the following, we review the single-model method of Khodabakhshi [10].

2.1. The single-model method of Khodabakhshi [10]

Khodabakhshi [10] proposed a single-model method based on the input relaxation model (jahanshahloo and Khodabakhshi [31]) to identify the input congestion of units as follows:

Maximize \[ \phi_0 + \epsilon \left( \sum_{i=1}^{n} s_{i1}^- + \sum_{r=1}^{s} s_{r}^+ - \sum_{i=1}^{m} s_{i2}^+ \right) \]

\[ \text{S.t.} \quad \sum_{j=1}^{n} \lambda_j x_{ij} + s_{i1}^- - s_{i2}^+ = x_{i0}, \quad i = 1, \ldots, m \]

\[ \sum_{j=1}^{n} \lambda_j y_{rj} - \phi_0 y_{r0} - s_{r}^+ = 0, \quad r = 1, \ldots, s \quad (1) \]

\[ \sum_{j=1}^{n} \lambda_j = 1 \]

\[ s_{i1}^-, s_{i2}^+, \lambda_j, s_{r}^+ \geq 0, \quad i = 1, \ldots, m; j = 1, \ldots, n; r = 1, \ldots, s. \]

The method of Khodabakhshi [10] reduces the number of required models for recognizing the congestion, which is certainly important from the perspective of computational time.

It should be noted that, the traditional DEA models only consider the situation that the production process generates the desirable outputs. However, in many real world applications, the undesirable outputs, such as CO2 emissions and any production waste, can be produced during the production process of the industrial activities. Therefore, an important issue in DEA is how to include the undesirable outputs in the production process. For more studies about the undesirable outputs in DEA, see Fare et al. [32], Seiford and Zhu [33], Yang et al. [34], Yang and Pollitt [35], Halkos and Petrou [36]).
3. THE PROPOSED METHOD TO RECOGNIZE THE COGESTION

In this section, we consider the situation that the production process produces the undesirable outputs and propose a model to recognize the congestion of units. The conventional DEA models for estimation of congestion only considers the inputs congestion. Given that, the reduction in the undesirable outputs can result in the increases in the desirable outputs, therefore, it is important to include the undesirable outputs in the evaluation of decision making units. In this paper, we suggest a new model to estimate the inputs congestion and the undesirable outputs congestion. For this purpose, we formulate model (2) as follows:

$$Maximize \quad \phi_0 + \varepsilon \left( \sum_{i=1}^{m} s_i^{-c} + \sum_{r=1}^{s} s_r^+ - \sum_{i=1}^{m} s_i^+ + \sum_{r=1}^{s} -s_i^{-bc} - \sum_{r=1}^{s} s_{rb}^+ \right)$$

$$S.t \quad \sum_{j=1}^{n} \lambda_j x_{ij} + s_{i}^{-c} - s_{i}^{+} = x_{i0} \quad i = 1, \ldots, m$$

$$\sum_{j=1}^{n} \lambda_j y_{rj} - \phi_0 y_{r0} - s_{r}^{+} = 0 \quad r = 1, \ldots, s$$

$$\sum_{j=1}^{n} \lambda_j b_{jt} + s_{t}^{-bc} - s_{t}^{+} = b_{t0} \quad t = 1, \ldots, p$$

$$\sum_{j=1}^{n} \lambda_j = 1$$

$$s_{i}^{-c}, s_{i}^{+}, \lambda_j, s_{r}^{+}, b_{t0}, \lambda_j \geq 0$$

Where, $s_{i}^{-c}$ and $s_{t}^{-bc}$ show the amount of congestion in the $i^{th}$ input and the $t^{th}$ undesirable output, respectively. Model (2) maximizes the desirable outputs and minimizes the inputs and the undesirable outputs. This model not only distinguishes the congested inputs and the undesirable outputs, but also measures the amount of congestion of each input and undesirable output.

The proposed method has the main properties:
1. Detecting the inputs congestion.
2. Recognizing the undesirable outputs congestion.
3. Proposing a model to identify the congestion of universities.
4. Investigating the impact of congestion on the inefficiency of units and determining the factors that lead to the congestion.
5. Investigating the reasons for the decreases in outputs due to the increases in inputs.

4. NUMERICAL EXAMPLE

In this section, the results of applying the proposed model to assess the performance and the congestion of 21 colleges of Zahedan branch of the Islamic Azad University in Iran, are reported. Each college is considered as a DMU with three inputs (the number of enrolled students ($x_1$), the number of faculty members ($x_2$), the number of courses provided by visiting lecturer ($x_3$)) and three desirable outputs (the number of graduates ($y_1$), the number of students admitted to higher level ($y_2$) and the number of research projects ($y_3$)) and one undesirable output (the number of expelled students ($y_4$)). Table 1 reports the data of units. The data has been collected based on the documents available in the faculties. It should
be noted that, the amount of $y_3$ that includes the number of research papers, e.g. ISI papers, ISC papers and etc, and also other research works can be considered as the research score for the faculty members according to the faculty members promotion regulations.

Table 1. The data of colleges.

| DMU | $x_1$ | $x_2$ | $x_3$ | $y_1$ | $y_2$ | $y_3$ | $y_4$ |
|-----|-------|-------|-------|-------|-------|-------|-------|
| 1   | 433   | 5     | 88    | 56    | 1     | 45    | 15    |
| 2   | 242   | 1     | 55    | 35    | 2     | 10    | 8     |
| 3   | 233   | 2     | 38    | 46    | 6     | 40    | 7     |
| 4   | 406   | 1     | 65    | 51    | 6     | 0     | 12    |
| 5   | 989   | 10    | 140   | 75    | 2     | 60    | 18    |
| 6   | 1276  | 9     | 145   | 148   | 2     | 60    | 22    |
| 7   | 2190  | 6     | 165   | 189   | 2     | 20    | 25    |
| 8   | 634   | 6     | 82    | 68    | 4     | 40    | 18    |
| 9   | 757   | 4     | 78    | 103   | 2     | 10    | 12    |
| 10  | 650   | 12    | 64    | 25    | 2     | 130   | 20    |
| 11  | 525   | 6     | 90    | 43    | 8     | 120   | 21    |
| 12  | 1020  | 2     | 100   | 116   | 1     | 0     | 24    |
| 13  | 1718  | 11    | 140   | 111   | 4     | 50    | 24    |
| 14  | 1025  | 6     | 120   | 80    | 3     | 90    | 17    |
| 15  | 655   | 11    | 88    | 30    | 10    | 50    | 16    |
| 16  | 924   | 5     | 92    | 39    | 1     | 20    | 21    |
| 17  | 360   | 6     | 72    | 31    | 1     | 20    | 10    |
| 18  | 779   | 8     | 102   | 146   | 1     | 10    | 14    |
| 19  | 589   | 5     | 92    | 115   | 1     | 0     | 15    |
| 20  | 441   | 4     | 80    | 35    | 1     | 20    | 13    |
| 21  | 258   | 5     | 50    | 11    | 1     | 20    | 12    |

Now, we use the proposed method to recognize the congested inputs and the congested undesirable outputs. The results are reported in Table 2.

Table 2. The results of model (2).

| DMU | $1/\phi^*$ | $s_1^{-c}$ | $s_2^{-c}$ | $s_3^{-c}$ | $s_t^{-hc}$ |
|-----|------------|------------|------------|------------|------------|
| 1   | 0.5327     | 0          | 0          | 0          | 6.5918     |
| 2   | 0.3269     | 0          | 0          | 0          | 12.3645    |
| 3   | 0.7354     | 0          | 0          | 0          | 11.467     |
| 4   | 0.7446     | 0          | 0          | 0          | 6.1804     |
| 5   | 0.7114     | 17.5722    | 2.2166     | 17.3055    | 3.5944     |
| 6   | 1          | 0          | 0          | 0          | 0          |
| 7   | 1          | 0          | 0          | 0          | 0          |
| 8   | 0.670      | 0          | 0          | 0          | 3.4132     |
| 9   | 0.6245     | 0          | 0          | 0          | 11.6444    |
| 10  | 1          | 0          | 0          | 0          | 0          |
The second column of Table 2 shows the efficiency of units. As we see, 5 colleges are efficient and 16 colleges are inefficient units. The last four columns of Table 2 show the amount of inputs congestion and the undesirable outputs congestion, respectively. The inefficient units have the congestion in some inputs and all of them have the congestion in the undesirable output. Therefore, the congested inputs and the congested undesirable output lead to the inefficiency of DMUs. It should be noted that, the increases in the number of expelled students \(y_4\) can reduce other outputs, such as the number of graduates and the number of students admitted at higher levels. Therefore, the undesirable output has the largest intensity for creating the inefficiency of units. The obtained results can help the DM to decide about increasing or decreasing the size of a particular DMU.

### 5. CONCLUSION

This study considered the production process in the presence of the undesirable outputs and proposed a method to recognize the inputs congestion and the undesirable outputs congestion. In this paper, we showed that, if we only increase the number of enrolled students and keep the other inputs, such as the faculty members, unchanged, then, the outputs, such as the number of research projects will be reduced, because of the faculty members have the limited time. In other words, the increases in the number of postgraduate students will increase the number of research projects, while the increases in the number of students at the bachelor’s degree level may reduce the number of research projects. Because of the faculty members can not be able to spend enough time for students as in the past. Also, the increases in the number of expelled students can reduce other outputs, such as the number of graduates and the number of students admitted at higher levels. Therefore, the congested undesirable outputs lead to the decreases in desirable outputs and the efficiency of units.

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