

A new model for evaluating cost efficiency using output slack variables in Data Envelopment Analysis

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Abstract

In previous methods to estimate cost efficiency using data envelopment analysis, a dominated decision-making unit (DMU) in the production possibility set (PPS) can be introduced as a cost efficiency unit. That is, a DMU may be cost-efficient, but not performance efficient. In this paper, at first, we provide a new definition for a cost-efficient unit, and then a model is presented in which the necessary condition for introducing a cost efficiency of DMU is its parato efficiency.

1 Introduction

Data envelopment analysis is a non-parametric method for evaluating the efficiency of similar Decision-Making Units (DMU). It is to use the same inputs to produce the same outputs [3, 1]. Cost efficiency (CE) is one of the concepts of data envelopment analysis that measures the ability of a DMU to produce current output at the lowest cost. In such cases, the price of each input is available. Cost efficiency was first proposed by Farrell [3] and then developed by Farr et al. [5], using linear programming. The proposed model requires the input and output values of each DMU and also the price of each input. Tone [7] expressed the shortcomings of the cost efficiency model when the price of each input is different for each DMU, and proposed a new method. The proposed method creates a set called Cost Space (CS) by multiplying the inputs of each DMU in its corresponding price and then evaluates the DMUs in this space. In all previous methods for evaluating cost-efficiency, a dominated DMU in the production possibility set (PPS) can also be introduced as a cost-efficient unit. In other words, may a DMU be cost-efficient, but not efficient in performance. It seems logical that a DMU should be called cost-efficient, which is also performance efficient. It means, not only the cost-efficient DMU works well, but it also produces its outputs at the lowest cost.

In this paper, we provide an example to illustrate this point and then propose a new definition of cost-efficiency. Using this definition and the models that follow, we show that Parato efficiency is a necessary condition for the cost efficiency of each DMU.

At first, we state the Farr et al [5] method to evaluate cost efficiency and its shortcomings. In the second section, we define a new definition and method for evaluating cost efficiency. Then, using the stated approach, a model for evaluating cost efficiency is presented. The fourth section will be included with a numerical example and finally, the results will be shown.

2 Farr et. al [5] method for evaluating CE

Suppose there are n DMUs under evaluation with the input and output (X_j, Y_j) , $(j = 1, 2, \dots, n)$ respectively, that $X_j = (x_{1j}, \dots, x_{mj}) \in \mathfrak{R}^m$ and $Y_j = (y_{1j}, \dots, y_{sj}) \in \mathfrak{R}^s$.

The Production Possibility Set with variable returns to scale (PPS_V), is expressed as follows:

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$$PPS_V = \{(X, Y) : X \geq \sum_{j=1}^n \lambda_j X_j, Y \leq \sum_{j=1}^n \lambda_j Y_j, \sum_{j=1}^n \lambda_j = 1, \lambda_j \geq 0 (j = 1, \dots, n)\}$$

The BCC input-oriented (BCC-I) model evaluates the efficiency of DMU_p , by solving the following linear program:

$$\begin{aligned} & \text{Min } \theta_p \\ & \text{s.t. } \sum_{j=1}^n \lambda_j x_{ij} \leq \theta_p x_{ip}, i = 1, \dots, m, \\ & \sum_{j=1}^n \lambda_j y_{rj} \geq y_{rp}, \quad r = 1, \dots, s, \\ & \sum_{j=1}^n \lambda_j = 1, \\ & \lambda_j \geq 0, \quad j = 1, \dots, n. \end{aligned} \tag{1}$$

where x_{ij} and y_{rj} (all nonnegative) are the inputs and outputs of the j th DMU.

Definition 2.1. DMU_p is called technical efficient, if in model (1) we have $\theta_p^* = 1$, where (*) is the optimum sign.

Definition 2.2. DMU_p is called BCC-strong efficient, if it is technical efficient and in any optimal solution of model (1), all slack variables are equal to zero.

Definition 2.3. DMU_p is Parato efficient in PPS_V , if it is not dominated by any other DMU. That means there is no other DMU in PPS_V with the property that all inputs are less than or equal to the inputs of DMU_p and all outputs are greater than or equal to the outputs of DMU_p and the inequality is strictly in at least one component.

Theorem 2.4. DMU_p is Parato-efficient in PPS_V if and only if it is BCCefficient.

To evaluate the cost-efficiency of the $DMU_p(CE_p)$, Farr et al. [5] proposed the following linear programming model. By solving this model for each DMU, the lowest cost consumed to produce outputs at least as much as current outputs is obtained, by knowing the price of inputs.

$$\begin{aligned} & \text{Min } \sum_{i=1}^m C_i x_i \\ & \text{s.t. } \sum_{j=1}^n \lambda_j x_{ij} \leq x_i, \quad i = 1, \dots, m, \\ & \sum_{j=1}^n \lambda_j y_{rj} \geq y_{rp}, \quad r = 1, \dots, s, \\ & \sum_{j=1}^n \lambda_j = 1, \\ & \lambda_j \geq 0, j = 1, \dots, n, \quad x_i \geq 0, \quad i = 1, \dots, m. \end{aligned} \tag{2}$$

where C_i is input price of i (x_i). The optimal solution of model (2), x_i^* shows the amount of input i required to produce outputs at least as much as current outputs at the lowest cost. Finally, the cost-efficiency of DMU_p is obtained by using cost ratio from the following relation:

$$CE_p = \frac{\sum_{i=1}^m C_i x_i^*}{\sum_{i=1}^m C_i x_{ip}}$$

Where (*) indicates the optimization in model (2). This measure of cost efficiency is called Farrell cost efficiency.

Definition 2.5. DMU_p is cost-efficient, if $CE_p = 1$.

In other words DMU_p is cost-efficient, if $\sum_{i=1}^m C_i x_i^* = \sum_{i=1}^m C_i x_{ip}$. Now we will discuss the shortcoming of the mentioned method and explain the matter with an example. We show that a DMU may be cost-efficient, but not strong efficient.

Example 1. Consider five decision making units with one input and one output according to Table 1. Figure 1 also shows the PPS made by them. Their cost-efficiency, evaluated by Model (2) and then Equation (3), is also given in the fourth column of Table 1. The input price of these DMUs is $C_1 = 2$. Consider three DMUs 1, 2, and 3 that consume one unit in input. DMU_2 produces three times as much output as DMU_1 , and DMU_3 produces $\frac{5}{3}$ times as much output as DMU_2 . That means DMU_3 dominates both DMU_1 and DMU_2 . As can be seen, the cost-efficiency of these three DMUs are equal to one, while DMU_3 is just strong-efficient. This is also significant from a management point of view, where a DMU can be introduced as a cost efficient unit, that also is efficient in performance.

Example.1 data			
DMU	Input x	Output y	Farell Cost Efficient
DMU_1	1	1	1
DMU_2	1	3	1
DMU_3	1	5	1
DMU_4	2	7	1
DMU_5	5	6	0.3

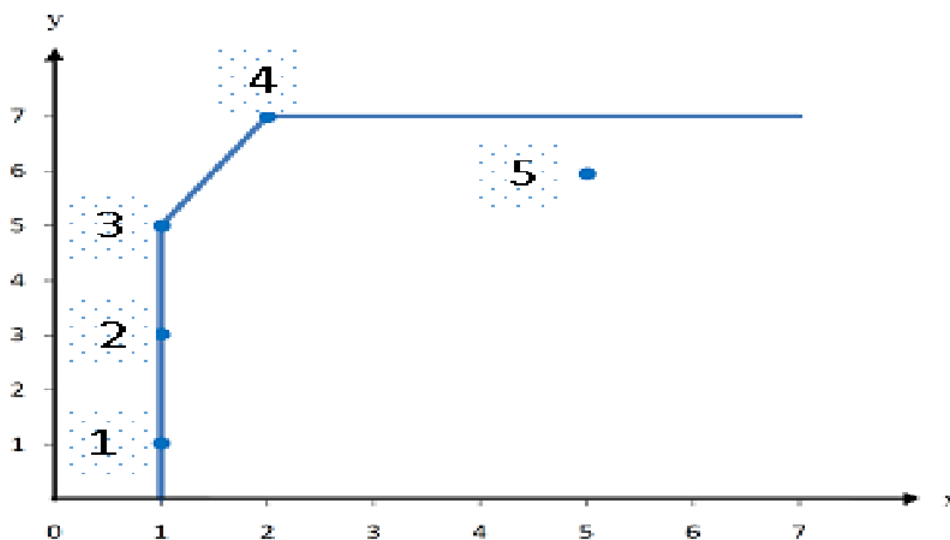


Figure 1: The Production Possibility Set of example 1

3 Proposed method for evaluating CE

In this section, we provide a new definition of cost efficiency and then present an approach to modify previous methods to evaluate cost-efficiency.

Definition 3.1. Suppose $CE_p = 1$. We say DMU_p is strong cost efficient if in each optimal solution of model (2) all slack variables are equal to zero. Otherwise DMU_p is weak cost efficient.

According to the above definition, to evaluate the cost-efficiency of a DMU, first, model (2) must be solved, and then using the optimal solution of that, the following model must be solved:

Max $\vec{1}S$

$$\begin{aligned}
 & \text{s.t. } \sum_{j=1}^n \lambda_j x_{ij} = x_i^*, i = 1, \dots, m, \\
 & \sum_{j=1}^n \lambda_j y_{rj} - s_r = y_{rp}, r = 1, \dots, s, \\
 & \sum_{j=1}^n \lambda_j = 1, \\
 & \lambda_j \geq 0, j = 1, \dots, n, s_r \geq 0, r = 1, \dots, s,
 \end{aligned} \tag{4}$$

Where x^* is the optimal solution of model (2) and $\vec{1} = (1, 1, \dots, 1)$.
 If in optimality of model (4), the value of objective function is equal to zero, then we say DMU_p is strong cost-efficient. Otherwise it will be weak cost-efficient.

Theorem 3.2. *If $CE_p = 1$, then DMU_p has a technical efficiency.*

Proof. Suppose $CE_p = 1$, then $(\lambda, X) = (e_p, X_p)$ is an optimal solution for model (2) in which e_p is a vector that all its components are zero except the p-th component which is one. Suppose to the contrary that $\theta_p^* < 1$. Since $(\theta_p^* X_p, Y_p) \in PPS$, there exists $\bar{\lambda} \geq 0$ with $\sum_{j=1}^n \bar{\lambda}_j = 1$, $\sum_{j=1}^n \bar{\lambda}_j X_j \leq \theta_p^* X_p$ and $\sum_{j=1}^n \bar{\lambda}_j Y_j \geq Y_p$. Therefore $(\bar{\lambda} \theta_p^* X_p)$ is a possible solution of model (2) where $C(\theta_p^* X_p) = \theta_p^* CX_p < CX_p$. This contradicts the optimality of $(\lambda, X) = (e_p, X_p)$.

Theorem 3.3. *If DMU_p is a strong cost-efficient, then it is strong-efficient.*

proof. Suppose DMU_p is Strong cost-efficient, then $CE_p = 1$ and in optimality of model (4), the optimal value of the objective function is zero. Since $CE_p = 1$, $(\lambda, X) = (e_p, X_p)$ is an optimal solution for model (2). Now suppose to the contrary that DMU_p is not strong-efficient. Thus there is a $(\bar{X}, \bar{Y}) \in PPS$ that dominates $DMU_{(p)}$ i.e. $(-\bar{X}, \bar{Y}) \geq (-X_p, Y_p)$ and $(-\bar{X}, \bar{Y}) \neq (-X_p, Y_p)$. Since $(\bar{X}, \bar{Y}) \in PPS$, so there is $\bar{\lambda} \geq 0$ where $\sum_{j=1}^n \bar{\lambda}_j X_j \leq \bar{X}$, $\sum_{j=1}^n \bar{\lambda}_j Y_j \geq \bar{Y}$ and $\sum_{j=1}^n \bar{\lambda}_j = 1$. If $\bar{X} \neq X_p$, then $(\bar{\lambda}, \bar{X})$ will be a feasible solution of model (2) and $\sum_{i=1}^m c_i \bar{x}_i < \sum_{i=1}^m c_i x_{ip}$, and this contradicts the optimality of $(\lambda, X) = (e_p, X_p)$. Also, if $\bar{Y} \neq Y_p$, then $(\lambda, S) = (\bar{\lambda}, \bar{S})$ is a possible solution for model (4) where $\bar{S} = \sum_{j=1}^n \bar{\lambda}_j Y_j - Y_p$. Obviously in this answer $\vec{1}\bar{S} > 0$ and this is also a contradiction.

Now we combine models (2) and (3) and present the following model:

$$\begin{aligned}
 & \text{Min } \sum_{i=1}^m c_i x_i - \varepsilon(\vec{1}S) \\
 & \text{s.t. } \sum_{j=1}^n \lambda_j x_{ij} = x_i, i = 1 \dots, m, \\
 & \sum_{j=1}^n \lambda_j y_{rj} - s_r = y_{rp}, r = 1, \dots, s, \\
 & \sum_{j=1}^n \lambda_j = 1, \\
 & \lambda_j \geq 0, j = 1, \dots, n, x_i \geq 0, i = 1, \dots, m,
 \end{aligned} \tag{5}$$

Where ε is a non-Archimedean number.

Using the above, we note that DMU_p is strong cost-efficient if and only if in each optimal solution (x^*, y^*) in model (5), we have $\sum_{i=1}^m c_i x_i^* = \sum_{i=1}^m c_i x_{ip}$ and $\vec{1}S^* = 0$.
 This section describes a method for evaluating the performance of DMUs from the cost point of view. According to the mentioned method, the output slack variables are also included in the evaluation of cost-efficiency. In this section,

this is done directly by presenting a model. Note the following model:

$$SCE_p = \text{Min } \rho_p = \frac{\frac{CX}{CX_p}}{1 + \frac{1}{s} \sum_{r=1}^s \frac{s_r}{y_{rp}}}$$

$$\text{s.t. } \sum_{j=1}^n \lambda_j x_{ij} = x_i, \quad i = 1, \dots, m,$$

$$x_i \leq x_{ip}, \quad i = 1, \dots, m,$$

$$\sum_{j=1}^n \lambda_j y_{rj} - s_r = y_{rp}, \quad r = 1, \dots, s, \tag{6}$$

$$\begin{aligned} \sum_{j=1}^n \lambda_j &= 1, \\ \lambda_j &\geq 0, \quad j = 1, \dots, n, \quad x_i \geq 0, \quad i = 1, \dots, m, \quad s_r \geq 0, \quad r = 1, \dots, s, \end{aligned}$$

Where $X = (x_1, x_2, \dots, x_m)$.

Theorem 3.4. *DMU_p is strong cost-efficient, if and only if SCE_p = 1.*

Proof. Assume DMU_p is a strong cost-efficiency, according to theorem (3) it also has strong efficiency. By definition 5 we have $CE_p = 1$ and all slack variables are equal to zero in optimal solution of model (2). By a way of contradiction suppose $SCE_p < 1$. Two cases may occur in model (6).

In the first case; there is optima solution such as $(\tilde{\lambda}, \tilde{X}, \tilde{S})$ in which there are some slack variables \tilde{s}_r that are positive. In this case $\tilde{X} \leq X_p$ and $\sum_{j=1}^n \tilde{\lambda}_j y_j - S \geq Y_p$ and at least one of the inequalities is strict. It means that DMU_p is dominated and it's not strong-efficient. It contradicts the assumption.

In the second case; There is an optimal answer for a model such as $(\lambda, \tilde{X}, \tilde{S})$, where $\frac{C\tilde{X}}{C\tilde{X}_p} < 1$. This also contradicts the assumption.

Conversely, suppose $SCE_p = 1$. Using the "proof by contradiction method", let DMU_p is not strong cost-efficient. two cases may occur; In the first case, there is an optimal solution for model (2) such as $(\tilde{\lambda}, \tilde{X})$ that $CX < CX_p$. Obviously, $(\tilde{\lambda}, \tilde{X}, 0)$ is a possible solution for model (6) with $SCE_p < 1$ and this contradicts the assumption. In the second case, $(\tilde{\lambda}^*, X^*)$ is the optimal solution of model (2) that $CX^* = CX_p$ i.e., $CE_p = 1$ and the optimal solution such as $(\tilde{\lambda}, \tilde{X})$ for model (4) exists for which $1\tilde{S} > 0$. Using these two optimal solutions, a possible solution such as $(\tilde{\lambda}, X^*, \tilde{S})$ can be constructed for model (6). Clearly for this solution, we have $SCE_p < 1$ and this contradicts the assumption.

According to Model (6), if we have $SCE_p < 1$ for DMU_p , it means that there is a virtual DMU such as $(\tilde{X}, \tilde{Y}) \in PPS$ that dominates DMU_p and its cost is less than DMU_p . It means $C\tilde{X} < CX_p$. The existence of such virtual DMU is easily verified by Model (6). Conversely, if there is no DMU located on strong efficiency boundary that produces output the same as DMU_p with lower cost, then it will have $SCE_p = 1$. Model (6) is a fractional linear programming model that is transformed into the following linear programming model using the Charnes and Cooper transformations [2]:

$$SCE_p = \text{Min } \bar{\rho}_p = \frac{C\bar{X}}{C\bar{X}_p}$$

$$\text{s.t } \sum_{j=1}^n \bar{\lambda}_j x_{ij} = \bar{x}_i, \quad i = 1, \dots, m, \quad \bar{x}_i \leq t x_{ip}, \quad i = 1, \dots, m,$$

$$\sum_{j=1}^n \bar{\lambda}_j y_{rj} - \bar{s}_r = t y_{rp}, \quad r = 1, \dots, s, \tag{6}$$

$$t + \frac{1}{s} \sum_{r=1}^s \frac{\bar{s}_r}{y_{rp}} = 1,$$

$$\sum_{j=1}^n \bar{\lambda}_j = t,$$

$$\bar{\lambda}_j \geq 0, \quad j = 1, \dots, n, \quad \bar{x}_i \geq 0, \quad i = 1, \dots, m, \quad \bar{s}_r \geq 0, \quad r = 1, \dots, s, \quad t \geq 0$$

4 Numerical examples

Example 2. Consider fifteen DMUs with one input and one output according to Table 2. The technical efficiency of these units is evaluated by model (1) and is shown in the fourth column of Table 2. Also, the fifth column of this table includes the cost efficiency of these units, which is obtained by solving model (2). In this example, decision making units 1, 2, 3, and 4 use the same inputs. But their outputs are different and the first three units are dominated by DMU_4 . This means that units 1, 2 and 3 are not cost-efficient. Using model (2), the cost efficiency of these four units is equal to one. Model (6) also gives the SCE, which is given in the sixth column of Table 2.

It is observed that in technical inefficient DMUs, $SCE < 1$. Also, not all DMUs with necessarily have an $SCE = 1$. This is due to the presence of positive slack variables in the optimal solution of model (2).

Exmaple.2 data and results of solving models 1, 2 and 6

DMU	Input x	Output y	Technical efficiency	Farell Cost Efficient	SCE
1	2	1	1	1	0.167
2	2	3	1	1	0.5
3	2	5	1	1	0.833
4	2	6	1	1	1
5	3	8	1	1	1
6	4	10	1	1	1
7	9	13	1	1	1
8	19	16	1	1	1
9	21	16	0.905	0.905	0.905
10	7	7	0.357	0.357	0.357
11	10	1	0.2	0.2	0.033
12	9	12	0.815	0.815	0.815
13	15	6	0.133	0.133	0.133
14	12	4	0.165	0.167	0.111
15	20	10	0.2	0.2	0.2

Example 3. Consider fifteen DMUs with two inputs and two outputs according to Table 3. The technical efficiency of these units is evaluated by model (1) and is shown in the sixth column of Table 3. Also, the seventh column of this table includes the cost efficiency of these units, which is obtained by solving model (2). Model (6) also gives the SCE, which is shown in the eighth column of Table 3. In this example, we also see that DMU_5 dominates DMU_{15} . These two units have the same technical efficiency and farell cost efficiency. But in these two units, only the SCE of unit number 5 is equal to one.

Exmaple.3 data and results of solving models 1, 2 and 6

DMU	Input x_1	Input x_2	Output y_1	Output y_2	Technical efficiency	Farell Cost Efficient	SCE
1	1	12	1	10	1	0.684	0.173
2	1	10	3	10	1	0.813	0.614
3	1	8	4	10	1	1	1
4	2	6	6	9	1	1	1
5	4	3	8	6	1	1	1
6	9	1	9	3	1	1	1
7	12	1	9	2	1	0.778	0.513
8	3	7	7	1	1	0.63	0.1
9	6	8	3	2	0.5294	0.472	0.104
10	12	2	6	4	0.6765	0.567	0.297
11	2	12	5	5	0.75	0.425	0.286
12	10	10	1	7	0.3722	0.367	0.05
13	6.5	2	7	7.5	1	1	1
14	9.1	1	1	1	1	0.802	0.09
15	4	3	6	6	1	1	0.75

5 Conclusion

This paper presents a method for evaluating cost efficiency. Previous methods had a problem that by solving a model, the DMU may be introduced as cost-efficient, but not efficient. It is not true in the management point of view. Therefore, it seems logical that in addition to checking the cost-efficiency out, the performance of DMUs should also be considered. The model presented in this paper introduces the necessary condition for the cost efficiency of a DMU to be its Parato efficiency. To this purpose, the output constraint slack variables are being considered to evaluate cost-efficiency. If $SCE_p = 1$, it means that there is no DMU in the PPS whose outputs are as much as that of DMU_p And its production cost is less than that of DMU_p .

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