

THE LINEAR PROGRAMMING APPROACH ON A-P SUPER-EFFICIENCY DATA ENVELOPMENT ANALYSIS MODEL OF INFEASIBILITY OF SOLVING MODEL

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ABSTRACT

The traditional CCR or BCC model in DEA for DMU assessment often come out with multiple efficiency units, which cause problems in rank ordering. Therefore, we intended to improve level of discrimination among Decision Making Units (DMUs). The present study thus seeks to solve the above-mentioned problems of discrimination and infeasible solutions in A-P super-efficient mode by proposing a feasible linear programming approach. In addition, for the nonlinear problem, using $(1-\alpha)/(1+\beta)$, in the objective function and constrains of the general model when applying SE in DEA, an improved linear programming solution mode is proposed, along with the assessment model of dominated decision making units to solve the problem of infeasible solution. Through the results of proof and calculation on the data in the given example, the present study proposes linear programming solution mode suitable for input-oriented, output-oriented and non-oriented model, respectively and such a linear programming solution model can substantially improve and solve the above mentioned problems of discrimination and infeasible solutions.

Keywords: Data Envelopment Analysis (DEA), Super-Efficiency (SE), Infeasibility

1. INTRODUCTION

Data Envelopment Analysis (DEA) was first employed by Charnes *et al.* (1978) to evaluate the efficiency among Decision Making Units (DMUs) through a linear programming model. DEA features its capability of processing multiple input and output variables simultaneously. Nevertheless, such a superior mode, when in the early stage of development, will often come with the problem that more than one DMU are considered efficiency unit, thus resulting in a lack of discrimination on efficiency evaluation. In view of this Andersen and Petersen (1993) then proposed the A-P Super Efficiency Model (SE model), which greatly helps enhance the

discrimination of efficiency evaluation. The essence of A-P model lies in the fact that the data of a certain DMU, while being evaluated, is not included in the efficiency frontier and the efficiency values of certain DMU are often greater than "1", which is how the term "super efficiency" is derived.

A-P model is not only originally employed to reorder the efficiency of DMU (Andersen and Petersen, 1993; Xue and Hurker, 2002; Chen, 2005; Ray, 2008; Pan *et al.*, 2011; Ebadi, 2012), but it is also used to detect the "outlier point" (Banker and Chang, 2005). Moreover, there are discussions on the sensitivity analysis as well as stability radius (Charnes *et al.*, 1992; Zhu, 1996; 2001; Seiford and Zhu, 1998; Leopold, 2003).

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Thrall (1996) points out that the A-P Super Efficiency (SE) mode under Constant Returns to Scale (CCR) often results in problems of infeasible solution. Zhu (1996) proves that SE and CCR mode is the sufficient condition of infeasible solution when the values of some of the data in the input or output variable are "0". In addition, even if the data values are not "0", as in other modes of the SE and the DEA, infeasible solution will also appear. Seiford and Zhu (1999) also put forward the necessary and sufficient conditions of infeasible solution in the SE model of DEA for input oriented and output-oriented situations. When applying SE in DEA, Cheng *et al.* (2011) proved that when any output values in DMU_k , $y_{rk} > y_{rj}$, $j = 1, \dots, n$, $j \neq k$, under VRS conditions, an infeasible solution will appear in A-P super-efficient model. To solve the shortcoming of the A-P, Cheng *et al.* (2011), without being confined to any oriented model, adopt $(1-\alpha)/(1+\beta)$ as the objective function and add some nonlinear formula to the constrains of the general model in order to solve the problem of infeasible solution. Though their discussions are comprehensive, their proposed model is filled with nonlinear mode, making the computation of efficiency evaluation complicated.

Based on the above analysis, the purpose of the study is to propose an improved linear programming model, based on the spirits of Cheng *et al.* (2011), to solve the problem of infeasible solution in SE of DEA. This study consists of five sections: (1) introduction (background description of research questions), (2) A-P mode and the mode proposed by Cheng *et al.* (2011), (3) SE-BCC model (as proposed by the present study, with the elaboration of solution feasibility with the aid of a theorem), (4) illustration by examples and proofs of such a model with its application and (5) conclusion.

2. MATERIALS AND METHODS

A-P Mode and the Mode Proposed by Cheng *et al.* (2011): (1) Input Oriented Mode-BCC Model:

$$\begin{aligned} & \text{Min. } \theta \\ & \text{s.t. } \sum \lambda_j X_{ij} \leq \theta X_{ik}, j \neq k, i = 1, \dots, m \\ & \sum \lambda_j y_{rj} \geq y_{rk}, r = 1, \dots, s \\ & \sum \lambda_j = 1, \lambda_j \geq 0, j = 1, \dots, n \end{aligned} \quad (1)$$

Cheng *et al.* (2011) proved that when applying A-P super efficiency mode of Equation (1), among all the output variables of DMU_k , if there is at least one y_{rk} , $r = 1, \dots, s$ which is greater than the y_{rj} of any other DMU_k , then there is no feasible solution in Equation (1).

2.1. Output Oriented Mode-BCC Model

$$\begin{aligned} & \text{Max. } \phi \\ & \text{s.t. } \sum \lambda_j X_{ij} \leq \theta X_{ik}, j \neq k, i = 1, \dots, m \\ & \sum \lambda_j y_{rj} \geq \phi y_{rk}, r = 1, \dots, s \\ & \sum \lambda_j = 1, \lambda_j \geq 0, j = 1, \dots, n \end{aligned} \quad (2)$$

Cheng *et al.* (2011) proved that when applying A-P super efficiency mode of Equation (2), among all the output variables of DMU_k , if there is at least one y_{rk} , $r = 1, \dots, s$ which is smaller than the y_{rj} of any other DMU_k , then there is no feasible solution in Equation (2).

2.2. Non-Oriented Mode

To summarize the above conditions where there is no feasible solution in Equation (1) and (2), Cheng *et al.* (2011) proposed a non-oriented mode using Equation (2) to solve infeasible solution in Equation (1) and (2):

$$\begin{aligned} & \text{Min } (1-\alpha)/(1+\beta) \\ & \text{s.t. } \sum \lambda_j X_{ij} \leq (1+\beta)X_{ik}, j \neq k, i = 1, \dots, m \\ & \sum \lambda_j y_{rj} \geq (1-\alpha)y_{rk}, r = 1, \dots, s \\ & \sum \lambda_j = 1, \lambda_j \geq 0, j = 1, \dots, n \end{aligned} \quad (3)$$

From the perspectives of the present study, the shortcomings of the improved mode by Cheng *et al.* (2011) include:

- The input-oriented and output-oriented models with certain functions were all excluded and were revised as Non-oriented Mode, which will somehow lose the spirit of DEA mode
- The non-oriented Mode in Equation (3) in its objective and constrains formula has become nonlinear programming, which causes difficulties in computation when solving the Equation

Based on the above mentioned shortcomings of SE-BCC mode, the present study thus proposes the following modes.

2.3. Input-Oriented Mode-BCC Model

In the formula (1), if the output values in DMU_k are multiplied by a "shrinking factor, $(1-\alpha)$ " in advance and make it a necessary condition, then the efficiency value of the given DMU_k can be acquired under the least $\alpha+\theta$ in the objective formula. The model after revision of Equation (1) is:

$$\begin{aligned}
 & \text{Min } \theta + \alpha \\
 & \text{s.t. } \sum \lambda_j X_{ij} \leq \theta X_{ik}, j=1, \dots, n, j \neq k, i=1, \dots, m \\
 & \sum \lambda_j y_{rj} \geq (1-\alpha) y_{rk}, r=1, \dots, s \\
 & \sum \lambda_j = 1, \lambda_j \geq 0, j=1, \dots, n
 \end{aligned} \tag{4}$$

In Equation (4), the purpose of adding “ α ” is to minimize “ α ” and the acquired “ θ ” value is not the true efficiency value “ θ^* ” of DMU_k . The value of θ^* should be $(\theta/(1-\alpha))$.

Proof:

The true value of “ θ^* ” in $DUM_k = (\sum u_r y_{rk} / \sum v_i X_{ik}) = (\sum (u_r(1-\alpha)y_{rk} / \sum (v_i(1-\alpha)X_{ik})) = \theta/(1-\alpha) \neq$

2.4. Output-Oriented Mode-Bcc Model

$$\begin{aligned}
 & \text{Max } \phi - \beta \\
 & \text{s.t. } \sum \lambda_j X_{ij} \leq (1+\beta) X_{ik}, j \neq k, i=1, \dots, m \\
 & \sum \lambda_j y_{rj} \geq \phi y_{rk}, r=1, \dots, s \\
 & \sum \lambda_j = 1, \lambda_j \geq 0, j=1, \dots, n
 \end{aligned} \tag{5}$$

In Equation (5), the purpose of subtracting “ β ” in the objective formula is to minimize “ β ” and the acquired “ β ” value is not the true efficiency value $\theta^*=1/\phi$ of DMU_k . The value of “ θ^* ” should be $((1+\beta) / \phi)$. Its proof is similar to that in Equation (4), thus omitted here.

2.5. Non-Oriented Model

$$\begin{aligned}
 & \text{Min. } \alpha + \beta \\
 & \text{s.t. } \sum \lambda_j X_{ij} \leq (1+\beta) X_{ik}, j \neq k, i=1, \dots, m \\
 & \sum \lambda_j y_{rj} \geq (1-\alpha) y_{rk}, r=1, \dots, s \\
 & \sum \lambda_j = 1, j \neq k, \lambda_j \geq 0, j=1, \dots, n
 \end{aligned} \tag{6}$$

Note that the objective formula of Equation (3) in the study of Cheng *et al.* (2011) is non-linear, but its function is similar to that of Equation (6) in the present study. If we put the results of Equation (6) in the objective formula of Equation (3), with other necessary conditions being constant, then the results of these two approaches should be the same, but the mode suggested by the present study is more feasible. $\theta^* = (1+\beta) / (1-\alpha)$.

3. RESULTS

An Illustration: Assume that there are 5 DMUs with 2 input variables and 2 output variables, data of which are specified below.

3.1. Input-Oriented Mode-BCC Mode

Since y_1 is greater than the first output variable of all the DMU, DMU_a in the given example is chosen and we put the data of **Table 1** into Equation (1), then:

$$\begin{aligned}
 & \text{Min } \theta + \alpha \\
 & \text{s.t. } 23 * \lambda_2 + 26 * \lambda_3 + 36 * \lambda_4 + 18 * \lambda_5 < 31 * \theta \\
 & 34 * \lambda_2 + 45 * \lambda_3 + 61 * \lambda_4 + 76 * \lambda_5 < 42 * \theta \\
 & 42 * \lambda_2 + 32 * \lambda_3 + 66 * \lambda_4 + 43 * \lambda_5 < 74 * \alpha > 74 \\
 & 25 * \lambda_2 + 65 * \lambda_3 + 35 * \lambda_4 + 55 * \lambda_5 + 55 * \alpha > 55 \\
 & \lambda_2 + \lambda_3 + \lambda_4 + \lambda_5 = 1
 \end{aligned} \tag{7}$$

The optimal solution to Equation (7) is: $\theta = 0.8438, \alpha = 0.4501, \theta^* = 0.8438 / (1-0.4501) = 1.5345$.

3.2. Output-Oriented Mode-BCC Mode

Though x_1, x_2 in DMU_a are not smaller than the first and second input variables of all the DMU, we still choose DMU_a as an example for explanation and put the data of **Table 1** into Equation (2), then:

$$\begin{aligned}
 & \text{Max } \Phi - \beta \\
 & \text{s.t. } 23 * \lambda_2 + 26 * \lambda_3 + 36 * \lambda_4 + 18 * \lambda_5 - 31 * \beta < 31 \\
 & 34 * \lambda_2 + 45 * \lambda_3 + 61 * \lambda_4 + 76 * \lambda_5 - 42 * \beta < 42 \\
 & 42 * \lambda_2 + 32 * \lambda_3 + 66 * \lambda_4 + 43 * \lambda_5 > 74 * \Phi \\
 & 25 * \lambda_2 + 65 * \lambda_3 + 35 * \lambda_4 + 55 * \lambda_5 > 55 * \Phi \\
 & \lambda_2 + \lambda_3 + \lambda_4 + \lambda_5 = 1
 \end{aligned} \tag{8}$$

The optimal solution to Equation (8) is: $\Phi = 0.6226, \beta = 0, \theta^* = 1/0.6226 = 1.6062$. Owing to the x_1 and x_2 of the DMU_a has not less than all DMUs’ first and second term of input variables. Since the optimal solution to Equation (8), $\beta = 0$, the model used in the present study does not distort the meaning of the original mode.

3.3. Non-Oriented Model

We still choose DMU_a as an example and put the data of **Table 1** into Equation (3), then:

$$\begin{aligned}
 & \text{Min. } \alpha + \beta \\
 & \text{s.t. } 23 * \lambda_2 + 26 * \lambda_3 + 36 * \lambda_4 + 18 * \lambda_5 - 31 * \theta * \beta < 31 \\
 & 34 * \lambda_2 + 45 * \lambda_3 + 61 * \lambda_4 + 76 * \lambda_5 - 42 * \theta * \beta < 42 \\
 & 42 * \lambda_2 + 32 * \lambda_3 + 66 * \lambda_4 + 43 * \lambda_5 > 74 * \alpha > 74 \\
 & 25 * \lambda_2 + 65 * \lambda_3 + 35 * \lambda_4 + 55 * \lambda_5 > 55 * \alpha > 55 \\
 & \lambda_2 + \lambda_3 + \lambda_4 + \lambda_5 = 1
 \end{aligned} \tag{9}$$

Table 1. Data of the illustration

DMU	Input		Output	
	X ₁	X ₂	Y ₁	Y ₂
A	31	42	74	45
B	23	34	42	25
C	26	45	32	65
D	36	61	66	35
E	18	76	43	55

The optimal solution to Equation (9) is: $\alpha = 0.3814$, $\beta = 0.0$, $\theta^* = (1+\beta)/(1-\alpha) = 1.6166$.

4. DISCUSSION

Efficiency Evaluation of the Dominated Decision Making Units: It is indicated from solution of non-oriented mode that when $\alpha = \beta = 0.0$, the efficiency value of DMU is “1” under the non-oriented mode of SE-BCC. If none of the output variables of DMU_k in a certain unit are greater than those of another unit and none of the input variables of DMU_k are smaller than those of DMU_j, then in Equation (6) $\alpha = \beta = 0.0$. To specify the relationships involved, we thus give the following definitions: When none of the output variables y_{rk} in DMU_k are greater than $\{y_{rj}\}$ (i.e., $y_{rk} \leq \{y_{rj}\}$) and also when $x_{jk} \leq \{x_{ij}\}$, $j = 1, \dots, n$, $j \neq k$, then DMU_k is called “dominated decision making units”.

Since there must be a lot of dominated decision making units among DMU, the adoption of Equation (3) (Cheng *et al.*, 2011) or Equation (6) of the present study will definitely result in another shortcoming of “low discrimination”. As the problem of infeasible solution in SE-BCC model does not necessarily exist, the present study thus suggests not to employ non-oriented mode to the efficiency evaluation of dominated decision making units. Though there will be two evaluation criteria in a system, the study claims that when in relative efficiency assessment, rank ordering of efficiency is more important than mere efficiency values and efficiency is of comparative nature.

Through the discussion in this study, we conceive that there should be different solutions to the problem of infeasible solution in SE-BCC mode. To take the data of individual DMU as example:

- y_{a1} of DMU_a $> \{y_{b1}, y_{c1}, y_{d1}, y_{e1}\}$; y_{c2} of DMU_c $> \{y_{a2}, y_{b2}, y_{d2}, y_{e2}\}$, thus, DMU_a and DMU_c fit in “input-oriented”, “output-oriented” and “non-oriented mode”
- x_{e1} of DMU_e $< \{x_{a1}, x_{b1}, x_{c1}, x_{d1}\}$ x_{b2} ; of DMU_b $< \{x_{a2}, x_{c2}, x_{d2}, x_{e2}\}$, thus, DMU_e and DMU_b fit it “input-oriented”, “output-oriented” and “non-oriented mode”

- DMU_d dominated by DMU_a, thus these two DMU do not fit in “non-oriented mode”

5. CONCLUSION

The A-P super efficiency model in DEA contributes greatly to enhancing level of discrimination. Such a powerful tool for efficiency evaluation often comes with infeasible solution under the conditions of variable returns to scale and a large number of scholars have been striving for finding solutions to the problem of infeasible solution. The present study revises the model of Cheng *et al.* (2011) by transforming their non-linear mode into linear mode, as well as providing a proof of the reasonable efficient value of DMU_k, through the use of a theorem. Also in this study, an example is given to delineate the application with extensive adoption of the improved SE-BCC model; that is, Dominated Decision Making Units is defined and the efficiency value is “1” under non-oriented conditions. In order not to lower discrimination level of efficiency evaluation, the study suggests that non-oriented model not be adopted.

6. REFERENCES

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