



Study Of The Optimization Of Measurement Efficiency Model Of Fuzzy Data Envelopment Analysis

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ARTICLE INFO	ABSTRACT
<p>Article history: Received: Jun 30, 2022 Revised: Jul 17, 2022 Accepted: Jul 25, 2022</p>	<p>The performance of determination efficiency is affected by both situation certainty and uncertainty. The goal of this study is to develop fuzzy data envelopment analysis models for measuring efficiency performance with the Data envelopment analysis (DEA) method and also a fuzzy method that considers the inputs and outputs of the decision making unit (DMU) on intuitionistic triangular fuzzy numbers. On model development, a step procedure in measurement efficiency performance is given. For processing its efficiency, there is input and output representation on useful intuitionistic triangular fuzzy numbers as output and input estimation data. The application of the fuzzy model to the given case demonstrates that it can be used and is capable of measuring efficiency performance more efficiently.</p>
<p>Keywords: Efficiency, performance , Data envelopment analysis (DEA), Fuzzy DEA.</p>	<p>Copyright © 2022 Jurnal Mantik. All rights reserved.</p>

1. Introduction

In a business environment, companies must ensure all planned goals can be achieved with a good result. Therefore, measurement performance is required to determine the level of achievement destination. The company measured efficiency to find out where input is used to produce output, or, in other words, how much output is produced in relation to the input used. Measurement efficiency can also be used to know which inputs only those who don't know how to use efficiently or know the output that should be upgraded with available inputs can conduct repair on these inputs. This method can be used to measure efficiency through Data Envelopment Analysis (DEA). DEA is a non-parametric method with multifactor input and output that is used to evaluate level efficiency relative to Decision Making Units (DMUs).

In development, the DEA model has been widely used in various fields. However, the general DEA model uses input and output variables containing one score (crisp) or one score sure. Whereas in reality, that's a possible score, neither input nor output variables could be determined by certain so-introduced fuzzy concepts. Integration of fuzzy concepts into the DEA model has been conducted by a number of researchers previously with different applications. Various fields The application of fuzzy DEA includes efficiency service health [1], [2],[3] , measurement transportation system efficiency [4], as well as measurement efficiency cost, efficiency income[5], and efficiency profit [6]. The basic output characteristics of the fuzzy DEA model could be distinguished. namely, the output that is still fuzzy [7], the output that is crisp [8], [9] and the output that is fuzzy and crisp [10]. However, the situations in daily life, multiple inputs and outputs simultaneously have subjective, linguistic, and form uncertainty and have an intuitive fuzzy essence or even obscurity [11]

So, this study leads the creation of a fuzzy DEA model that is, in essence, institutionalistic, with inputs and outputs that are owned and shown by an intuitive triangular fuzzy number. The aim of this paper is to develop a fuzzy DEA model that represents the input and the output as an intuitionistic triangular fuzzy number. To show how fuzzy DEA's model used in this study, the case studies were

used to show how the model using to measure the company's efficiency by calculating the score of its own performance.

2. Method

2.1 Development of Fuzzy Data Envelopment Analysis Model

The Idea of This study developed a data environment analysis (DEA) model by [12] to evaluate performance [13] with a determined score efficiency, but with the inherent uncertainty nature of the inputs and outputs. In this study, the measurement efficiency performance was achieved through data envelopment analysis (DEA). This DEA model is capable of analyzing efficiently each take-up unit decision or a decision making unit (DMU) with multiple inputs and multiple outputs [14]. In the study, DEA model was developed. This is the classic DEA CCR model, with natural input and output uncertainty. The inputs and outputs are looked at with a rough set of fuzzy triangular numbers and an institutionalistic twist. In this study, the inputs and outputs are on defined fuzzy environment as triangular intuitionistic fuzzy number (TIFN).

Definition 1 : fuzzy triangular intuitionistics $\tilde{a} = (a^l, a^m, a^u)$ is the set of real numbers in R , with function membership as following :

$$\mu_{\tilde{a}}(x) = \begin{cases} \frac{x - a^m}{a^m - a^l} \mu_{\tilde{a}}, & \text{jika } a^l \leq x \leq a^m \\ \mu_{\tilde{a}}, & \text{jika } x = a^m \\ \frac{a^u - x}{a^u - a^m} \mu_{\tilde{a}}, & \text{jika } a^m < x \leq a^u \\ 0, & \text{jika } x < a^l \text{ atau } x > a^u \end{cases}$$

And there are non - membership function that can be defined as following :

$$\vartheta_{\tilde{a}}(x) = \begin{cases} \frac{a^m - x + (x - a^l)}{a^m - a^l} \vartheta_{\tilde{a}}, & \text{jika } a^l \leq x < a^m \\ \vartheta_{\tilde{a}}, & \text{jika } x = a^m \\ \frac{x - a^m + (a^u - x)}{a^u - a^m} \vartheta_{\tilde{a}}, & \text{jika } a^m \leq x < a^u \\ 1, & \text{jika } x < a^l \text{ atau } x > a^m \end{cases} \tag{2}$$

$\mu_{\tilde{a}}$ is the maximum membership degree and $\vartheta_{\tilde{a}}$ is the minimum non-membership degree that satisfies the following conditions: $0 \leq \mu_{\tilde{a}} \leq 1, 0 \leq \vartheta_{\tilde{a}} \leq 1, 0 \leq \mu_{\tilde{a}} + \vartheta_{\tilde{a}} \leq 1$. [11]

On this proposed model development, consider maximizing output with input and output on DMU to j defined intuitionistic triangular fuzzy number as following :

$\tilde{x}_{ij} = ((x_{ij}^l, x_{ij}^m, x_{ij}^u); \mu_{x_{ij}}, \vartheta_{x_{ij}})$ and $\tilde{y}_{rj} = ((y_{rj}^l, y_{rj}^m, y_{rj}^u); \mu_{y_{rj}}, \vartheta_{y_{rj}})$. At the $x_{ij}^l, x_{ij}^m, x_{ij}^u$ stated value that x_{ij}^l = value x input *lower* , x_{ij}^m = value x input *medium* , x_{ij}^u = value x input *upper* . Whereas for $(y_{rj}^l, y_{rj}^m, y_{rj}^u)$ states y_{rj}^l = value of y output *lower* , y_{rj}^m = value of y output *medium*, y_{rj}^u = value of y output *upper* .

Then the fuzzy DEA model can be denoted as following :

Maximize $\sum_{r=1}^s u_r \tilde{y}_{rp}$ (3)



Constraints :

$$\sum_{i=1}^m v_i \tilde{x}_{ip} = \tilde{1} \quad (4)$$

$$\sum_{r=1}^s u_r \tilde{y}_{rj} - \sum_{i=1}^m v_i \tilde{x}_{ij} \leq \tilde{0}, \quad (5)$$

$$u_r \geq 0; r = 1, 2, \dots, s$$

$$v_i \geq 0; i = 1, 2, \dots, m$$

Where :

u_r = r th output vector weight

v_i = i -th input vector weight

\tilde{x}_{ij} = i -th intuitive triangular fuzzy number on the j -th DMU .

\tilde{y}_{rj} = the output value which is triangular intuitionistic fuzzy number to r on the j -th DMU.

Definition 2 . suppose there is fuzzy triangular intuitionistics $\tilde{a}_i = ((a_i^l, a_i^m, a_i^u); \mu_i, \vartheta_i)$. Then the ranking aggregate function is as follows:

$$\bar{R}(\sum_{i=1}^n \tilde{a}_i) = (1 + \min \mu_i - \max \vartheta_i) \times \sum_{i=1}^n \frac{R(\tilde{a}_i)}{1 + \mu_i - \vartheta_i} = \frac{(1 + \min \mu_i - \max \vartheta_i)}{6} \sum_{i=1}^n (a_i^l + a_i^m + a_i^u) \quad (6)$$

Definition 2 above is ranking the aggregate will be converted on a DEA model containing fuzzy triangular intuitionistics on inputs and outputs.

Then the DEA model of triangular intuitionistics fuzzy numbers is used are :

$$\text{Maksimumkan} \quad \sum_{r=1}^s u_r \left((y_{rp}^l, y_{rp}^m, y_{rp}^u); \mu_{y_{rp}}, \vartheta_{y_{rp}} \right)$$

Constraints : (4)

$$\sum_{i=1}^m v_i \left((x_{ip}^l, x_{ip}^m, x_{ip}^u); \mu_{x_{ip}}, \vartheta_{x_{ip}} \right) = \tilde{1}$$

$$\sum_{r=1}^s u_r \left((y_{rj}^l, y_{rj}^m, y_{rj}^u); \mu_{y_{rj}}, \vartheta_{y_{rj}} \right) - \sum_{i=1}^m v_i \left((x_{ij}^l, x_{ij}^m, x_{ij}^u); \mu_{x_{ij}}, \vartheta_{x_{ij}} \right) \leq \tilde{0},$$

$$u_r \geq 0; r = 1, 2, \dots, s$$

$$v_i \geq 0; i = 1, 2, \dots, m$$

3. Results And Discussion

From the derivated formula in the previous section, the DEA model of triangular intuitionistics fuzzy numbers is used:

$$\text{Maksimumkan} \quad \sum_{r=1}^s u_r \left((y_{rp}^l, y_{rp}^m, y_{rp}^u); \mu_{y_{rp}}, \vartheta_{y_{rp}} \right) \quad (5)$$

Constraints :

$$\sum_{i=1}^m v_i \left((x_{ip}^l, x_{ip}^m, x_{ip}^u); \mu_{x_{ip}}, \vartheta_{x_{ip}} \right) = \tilde{1}$$

$$\sum_{r=1}^s u_r \left((y_{rj}^l, y_{rj}^m, y_{rj}^u); \mu_{y_{rj}}, \vartheta_{y_{rj}} \right) - \sum_{i=1}^m v_i \left((x_{ij}^l, x_{ij}^m, x_{ij}^u); \mu_{x_{ij}}, \vartheta_{x_{ij}} \right) \leq \tilde{0},$$



$$u_r \geq 0; r = 1, 2, \dots, s$$

$$v_i \geq 0; i = 1, 2, \dots, m$$

Here we developed the algorithm to measurement fuzzy DEA model efficiency:

- Step 1. Analyze and identify the input and output variables for every DMU owned. Then define estimation or estimation performance for each input and output variable on the DMU.
- Step 2. Estimate score the performance of the input and output variables of each DMU.
- Step 3. Define a DEA fuzzy model
- Step 4. Transformation equation (4) into form **definition 2** for the craps model.
- Step 5. Solve the DEA fuzzy model with the linear programming method. Completion of this DEA fuzzy model was conducted as many times as DMU owned.
- Step 6. Counting efficiency performance from step model 4th, the efficient DMU analyzed based on the ranking[15] that has been obtained.

3.1 Numerical Example

The fuzzy DEA model that has been proposed will be implemented next on a case by case basis in order to know that the proposed model could be implemented. Data used in the implementation used case on this study adapted from a previous study by [16]

Step 1: Identify the inputs and outputs of given case . In this case, there are 8 DMU with 4 inputs and 2 outputs. Input in study this is cost fare worker , (I_1), cost ingredient standard (I_2), Cost inventory (I_3), and time production (I_4). Output is profit (O_1), Revenue (O_2). Following procedure step fuzzy DEA solution that has been built .

Step 2. Estimating score performance Estimate score the performance of the input and output variables of each DMU is carried out by the fuzzification process to form triangular instuitionistic fuzzy number like in the table. 1 and table 2 explain about the output of data for fuzzy DEA.

Table 2
Estimated input data for DEA fuzzy solution

DMU	Input 1 ($l, m, u ; \mu, \vartheta$)	Input 2 ($l, m, u ; \mu, \vartheta$)	Input 3 ($l, m, u ; \mu, \vartheta$)	Input 4 ($l, m, u ; \mu, \vartheta$)
1	(71.6,72.35,73.10; 0.4,0.5)	(12.6,14.35,16.10; 0.7, 0.3)	(53.09,54.7,56.31; 1.0, 0.0)	(3.55, 5.2, 6.85; 0.8, 0.1)
2	(67.14,68.32,69.50; 0.7, 0.2)	(22.3,23.6,24.9; 0.6, 0.1)	(53.09, 54.7, 56.31; 0.9, 0.1)	(3.9, 6, 8.1; 0.5, 0.3)
3	(72.07,72.35, 72.63; 1.0, 0.0)	(8.5,10.5,12.5; 0.4, 0.2)	(53.09, 54.7, 56.31; 0.8,0.2)	(3.55, 5.2, 6.85; 0.7, 0.3)
4	(71.7, 72.35, 73.00; 0.9, 0.1)	(10.58,13.2,15.82; 0.8, 0.1)	53.09, 54.7, 56.31; 0.7, 0.1)	(3.5, 5.2, 6.85; 0.9, 0.1)
5	(67.44, 68.32, 69.2; 0.8, 0.2)	(6.64, 9.14, 11.64; 0.6, 0.3)	(53.09, 54.7, 56.31; 0.6, 0.2)	(3.9, 6, 8.1; 1.0, 0.0)
6	(67.83,68.32, 68.81; 0.7, 0.3)	(19.61, 22.31, 25.01; 0.8, 0.2)	(53.09, 54.7, 56.31; 0.5, 0.4)	(3.9, 6, 8.1; 0.8, 0.2)
7	(70.47,72.35, 74.47; 0.6, 0.1)	(4.73,7.23, 9.73; 1.0, 0.0)	(53.09, 54.7, 56.31; 0.6, 0.1)	(3.5, 5.2, 6.85; 0.6, 0.4)
8	(66.62,68.32,70.02; 0.5, 0.4)	(21.55, 24.3, 27.45; 0.8, 0.2)	(53.09, 54.7, 56.31; 0.7, 0.3)	(3.9, 6, 8.1; 0.7, 0.2)



Table 2
Estimated input data for DEA fuzzy solution

DMU	Output 1 (l,m,u ; μ, ϑ)	Output 2 (l,m,u ; μ, ϑ)
1	(132,240,348; 0.9,0.1)	(168, 200, 232; 0.1, 0.0)
2	(389, 410, 431; 0.6, 0.2)	(40, 80, 120; 0.8, 0.1)
3	(92, 158, 224; 0.9, 0.0)	(65, 95, 125; 0.7, 0.1)
4	(152, 235, 318; 0.7, 0.2)	(25, 50, 75; 0.5, 0.2)
5	(181, 198, 215; 0.8, 0.2)	(100, 112, 124; 0.4, 0.3)
6	(200, 300, 400; 0.6, 0.4)	(80, 98, 116; 0.7, 0.2)
7	(100, 250, 400; 0.4, 0.2)	(78, 102, 126; 0.6, 0.2)
8	(80, 150, 220; 1.0, 0.0)	(112, 152, 192; 0.9, 0.1)

Step 3 :Modelling fuzzy DEA accordingly with the proposed formulation model and on step to 4: transform the model on step 3 to the shape of the crisp model . So from the data obtained on then the DEA fuzzy model for DMU 1 could formulated as following :

$$\begin{aligned}
 & \text{Maximize} \quad \left[\frac{1 + \min(0.9; 1.0) - \text{maks} (0.1; 0.0)}{6} \right] (132 + 240 + 348) u_1 \\
 & \quad \quad \quad + (168 + 200 + 232) u_2 \\
 & \text{Subject to :} \\
 & \quad \quad \quad \left[\frac{1 + \min(0.4; 0.7; 1.0; 0.8) - \text{maks} (0.5; 0.3; 0.0; 0.1)}{6} \right] (71.6 + 72.35 + 73.10) v_1 \\
 & \quad \quad \quad + (12.6 + 14.35 + 16.10) v_2 + (53.09 + 54.7 + 56.31) v_3 \\
 & \quad \quad \quad + (3.55 + 5.2 + 6.85) v_4 = 1 \\
 & \left[\left(\frac{1 + \min(0.9; 1.0) - \text{maks} (0.1; 0.0)}{6} \right) (132 + 240 + 348) u_1 + (168 + 200 + 232) u_2 \right] \\
 & \quad - \left[\left(\frac{1 + \min(0.4; 0.7; 1.0; 0.8) - \text{maks} (0.5; 0.3; 0.0; 0.1)}{6} \right) (71.6 + 72.35 \right. \\
 & \quad \quad \quad \left. + 73.10) v_1 + (12.6 + 14.35 + 16.10) v_2 + (53.09 + 54.7 + 56.31) v_3 \right. \\
 & \quad \quad \quad \left. + (3.55 + 5.2 + 6.85) v_4 \right] \leq 0 \\
 & \left[\left(\frac{1 + \min(0.6; 0.8) - \text{maks} (0.1; 0.2)}{6} \right) (389 + 410 + 431) u_1 + (40 + 80 + 120) u_2 \right] \\
 & \quad - \left[\left(\frac{1 + \min(0.7; 0.6; 0.9; 0.5) - \text{maks} (0.2; 0.1; 0.1; 0.3)}{6} \right) (67.14 + 68.32 \right. \\
 & \quad \quad \quad \left. + 69.50) v_1 + (22.3 + 23.6 + 24.9) v_2 + (53.09 + 54.7 + 56.31) v_3 \right. \\
 & \quad \quad \quad \left. + (3.9 + 6 + 8.1) v_4 \right] \leq 0
 \end{aligned}$$



$$\left[\left(\frac{1 + \min(0.9; 0.7) - \text{maks} (0.0; 0.1)}{6} \right) (92 + 158 + 224)u_1 + (65 + 95 + 125)u_2 \right] \\ - \left[\left(\frac{1 + \min(1.0; 0.4; 0.8; 0.7) - \text{maks} (0.0; 0.2; 0.2; 0.3)}{6} \right) (72.07 + 72.35 \right. \\ \left. + 72.63)v_1 + (8.5 + 10.5 + 12.5)v_2 + (53.09 + 54.7 + 56.31)v_3 \right. \\ \left. + (3.55 + 5.2 + 6.85)v_4 \right] \leq 0$$

$$\left[\left(\frac{1 + \min(0.7; 0.5) - \text{maks} (0.2; 0.2)}{6} \right) (152 + 235 + 318)u_1 + (25 + 50 + 75)u_2 \right] \\ - \left[\left(\frac{1 + \min(0.9; 0.8; 0.7; 0.9) - \text{maks} (0.1; 0.1; 0.1; 0.1)}{6} \right) (71.7 + 72.35 \right. \\ \left. + 73.00)v_1 + (10.58 + 13.2 + 15.82)v_2 + (53.09 + 54.7 + 56.31)v_3 \right. \\ \left. + (3.5 + 5.2 + 6.85)v_4 \right] \leq 0$$

$$\left[\left(\frac{1 + \min(0.8; 0.4) - \text{maks} (0.2; 0.3)}{6} \right) (181 + 198 + 215)u_1 + (100 + 112 + 124)u_2 \right] \\ - \left[\left(\frac{1 + \min(0.8; 0.6; 0.6; 1.0) - \text{maks} (0.2; 0.3; 0.2; 0.0)}{6} \right) (67.44 + 68.32 \right. \\ \left. + 69.2)v_1 + (6.64 + 9.14 + 11.64)v_2 + (53.09 + 54.7 + 56.31)v_3 \right. \\ \left. + (3.9 + 6 + 8.1)v_4 \right] \leq 0$$

$$\left[\left(\frac{1 + \min(0.6; 0.7) - \text{maks} (0.4; 0.2)}{6} \right) (200 + 300 + 400)u_1 + (80 + 98 + 116)u_2 \right] \\ - \left[\left(\frac{1 + \min(0.7; 0.8; 0.5; 0.8) - \text{maks} (0.3; 0.2; 0.4; 0.2)}{6} \right) (67.83 + 68.32 \right. \\ \left. + 68.81)v_1 + (19.61 + 22.31 + 25.01)v_2 + (53.09 + 54.7 + 56.31)v_3 \right. \\ \left. + (3.9 + 6 + 8.1)v_4 \right] \leq 0$$

$$\left[\left(\frac{1 + \min(0.4; 0.6) - \text{maks} (0.2; 0.2)}{6} \right) (100 + 250 + 400)u_1 + (78 + 102 + 126)u_2 \right] \\ - \left[\left(\frac{1 + \min(0.6; 1.0; 0.6; 0.6) - \text{maks} (0.1; 0.0; 0.1; 0.4)}{6} \right) (70.47 + 72.35 \right. \\ \left. + 74.47)v_1 + (4.73 + 7.23 + 9.73)v_2 + (53.09 + 54.7 + 56.31)v_3 \right. \\ \left. + (3.5 + 5.2 + 6.85)v_4 \right] \leq 0$$



$$\left[\left(\frac{1 + \min(1.0; 0.9) - \max(0.0; 0.1)}{6} \right) (80 + 150 + 220)u_1 + (112 + 152 + 192)u_2 \right] - \left[\left(\frac{1 + \min(0.5; 0.8; 0.7; 0.7) - \max(0.4; 0.2; 0.3; 0.2)}{6} \right) (66.62 + 68.32 + 70.02)v_1 + (21.55 + 24.3 + 27.05)v_2 + (53.09 + 54.7 + 56.31)v_3 + (3.9 + 6 + 8.1)v_4 \right] \leq 0$$

$$u_r \geq 0; r = 1,2;$$

$$v_i \geq 0; i = 1,2,3$$

Step 5: The formulation solved with using with LINGO and this paper using LINGO student version, shown figure 1.

Variable	Value	Reduced Cost
U1	0.4209233E-02	0.000000
U2	0.8312731E-03	0.000000
V1	0.3067485E-01	0.000000
V2	0.000000	0.000000
V3	0.000000	0.000000
V4	0.000000	0.000000
V5	0.000000	0.000000
V6	0.000000	0.000000
V7	0.000000	0.000000
V8	0.000000	0.000000

Row	Slack or Surplus	Dual Price
1	1.000000	1.000000
2	0.000000	1.000000
3	0.000000	1.000000
4	0.000000	0.000000
5	0.6284351	0.000000
6	1.105483	0.000000
7	0.8524147	0.000000
8	0.3442199	0.000000
9	0.6508699	0.000000
10	0.4707961	0.000000
11	0.4209233E-02	0.000000
12	0.8312731E-03	0.000000
13	0.3067485E-01	0.000000
14	0.000000	0.000000
15	0.000000	0.000000
16	0.000000	0.000000
17	0.000000	0.000000
18	0.000000	0.000000
19	0.000000	0.000000
20	0.000000	0.000000

Figure 1. Output Fuzzy DEA DMU 1 Program LINGO student version

The solution with LINGO is as many as DMU owned. In this implementation case, owned DMU, there are 8, so there are 8 times processing with LINGO.

Step 6: Based on measurement efficiency with LINGO, we obtained that of 8 DMUs analyzed, there are 2 efficient DMUs, i.e., DMU 1 and DMU 2. Besides that, there are 6 DMUs that aren't efficient, namely DMU 3, DMU 4, DMU 5, DMU 6, DMU 7, and DMU 8.



4. Conclusion

Study this discuss development of data envelopment analysis (DEA) models in measure efficiency performance which performance input and the output have existing characteristics on environment uncertainty . In resolve uncertainty that approached with set intuitionistic triangular fuzzy numbers . Generally function defined fuzzy membership , however with triangular intuitionistic fuzzy number defined also non- membership function and calculated score aggregate rank efficiency each take- up unit decision (DMU). The proposed DEA fuzzy model implemented and the result of the proposed model could applied for measure level efficiency in a company for planning production could efficient.

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