

# A Research Framework for Data Envelopment Analysis with Upper Bound on Output to Measure Efficiency Performance of Higher Learning Institution in Aceh Province

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**Abstract**— The higher education system in Indonesia can be considered not only as an important source of developing knowledge in the country but also could create favourable living conditions for the country. Therefore, it is not surprising that enrollments in higher education continue to expand. Data envelopment analysis (DEA) is a method to evaluate the technical efficiency of production units which have multiple input and output. The higher learning institution considered in this paper is Min Aceh province of Indonesia. This research framework of this research is DEA, with the bounded output. Accordingly, we present some important differences in efficiency performance of higher education institute. Finally, we will discuss the effort from these departments to become efficient.

**Keywords**— data envelopment analysis; efficiency; higher education; performance measurement; upper bound

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## I. INTRODUCTION

Analysis of efficiency in Engineering and Economics traditionally done using the classical efficiency ratio based on input and output weights for all Decision Making Units (DMU) that is being evaluated [1]. The benchmarking process is a process for measuring a performance that is expected to improve efficiency in the implementation of activity [2] and can also be used to formulate performance standards that can support organizational growth [3]. Farrel in [4] proposes a method that can measure efficiency by using multiple outputs and inputs called relative efficiency, by calculating the ratio of the weight of the output sum and the weight of the input sum. Charnes, Cooper, and Rhodes in [5] developed the Farrel [4] idea to develop a method for evaluating the relative efficiency of the Decision Making Unit (DMU) set. Charnes, Cooper, and Rhodes have proposed The Model for Measurement of DEA Efficiency [5], known as DEA Charnes, Cooper, and Rhodes (DEA CCR) and has been used by some researchers for the measurement of relative efficiency by using DEA. Eskelinen [6] uses the DEA CCR Model for efficiency measurement in a retail bank. Lin and Chen [7] have used DEA CCR to measure the relative efficiency of capital and resource placement. Researchers have applied the DEA model to address corporate, regional and national sustainability issues as well as those related to the supply chain [8].

DEA has become one of the most appropriate methods for comparing the various Decision Making Units (DMU) associated with public services such as universities [9]. There is currently no performance measurement process undertaken either by the government for the higher education level in Aceh Province. Higher Education has a huge role in improving the quality of human life. In Indonesia, universities as institutions of higher education have a great responsibility to improve the quality of life of the community, to develop the welfare of the state and to increase scientific knowledge. For this reason, the Government of Indonesia has sought to raise funds to support the education process. It is common in the higher education sector in Indonesia to gain some financial support from public funding, especially from the cost of student education. The same situation also occurred in the existing universities in the province of Aceh, Indonesia.

Measuring the efficiency of college performance is very important to do, but it is difficult to do considering the characteristics of each different college especially if the college is viewed as a nonprofit on organization with multiple outputs generated from multiple inputs. A study by Hussein, Noordin, and Ishak [10] has two main outputs that can be used to measure college performance: the number of graduates and the number of publications. On the other hand, research conducted by Etzkowitz and Zhou [11] that the main inputs of measuring the efficiency of college performance are the number of teaching staff and the

number of students. Similar research has previously been conducted by Johnes and Yu [12] using DEA method to measure relative efficiency by researching 109 colleges in China in 2003 and 2004. In their research, Impact measurements and research productivity were made as output Variables and as input variables used the number of teaching staff, the number of graduates, capital, and resources.

In general Data Envelopment Analysis, the data, in general, is assumed to be non-negative with no specific limitations. However, some cases require some restrictions on the data [13]. Cooper et al. assumed that each DMU has a range of values for each input and output measurement. The rationale is that each DMU has limits for input and output [14]. It uses lower bounds input and upper bounds output to define the limits of efficiency measurement [14]. Branda suggests a DEA-oriented method of studying the relationship between input and output [15].

Wu et al. have examined that an unexpected number of outputs may arise in the form of a deviation of the resulting output, based on the existing input parameters, if there is no limit to the output [16]. This is seen in the study by Smirlis, Maragos, Despotis conducted a study on the efficiency of school performance in Greece using the lower bound on the input and upper bound on the output, where an inefficient Decision Making Unit (DMU) can occur [17]. The primary constraint when establishing bounded on input and output based on the results of research conducted by Wei and Wang is the presence of imprecise data [18].

Research conducted by experts so far there has been no research that discusses the upper bound output to fixed input. Where, for reasons of the quality of college performance efficiency measurement, the input bound is assumed to be fixed, and the output will increase proportionally along with the utilization of the existing input. Fixed input is caused by the addition of resources at the university especially for the input of the teaching staff, and the number of students has a quota set by the Ministry of Research, Technology and Higher Education which cannot be added, but who want to increase is an increase in output (Upper bound on output). The Data Envelopment Analysis (DEA) with Upper Bound on Output is aimed at measuring the efficiency of university performance in Aceh using two main inputs: Number of Lecturers and Number of Students as well as 2 (two) main outputs: Number of Research And Number of Graduates. This research is needed to describe the performance efficiency level of each university.

## II. MATERIAL AND METHOD

Measurements and Comparison are necessary for the Benchmarking process to achieve higher performance. Benchmarking in this research is to measure the relative performance efficiency. Relative efficiency is the measurement of efficiency for multiple inputs and multiple outputs, which can be defined as the ratio of total weighted output divided by total weighted input / total weighted input. [16] have examined that an unexpected number of outputs may arise in the form of the resulting output deviation, based on the existing input parameters, if there is no limit to the output. Research conducted by experts so far there has been

no research that discusses the upper bound output to fixed input.

Research conducted by experts so far there has been no research that discusses the upper bound output to fixed input. Where, for the reason of the quality of college performance efficiency measurement, the input bound is assumed to be fixed, and the output will increase proportionally along with the increase of the explored input. Data Envelopment Analysis (DEA) with Upper Bound on Output is aimed at measuring the efficiency of the performance of universities in Aceh using two main inputs: Number of Teaching Staff and Number of Students as well as 2 (two) main outputs: Number of Research and Number of Graduates.

### A. Data

The data source for benchmarking in this research is data of university in Aceh Province. There are several variables observed for performance measurement, namely: Number of Lecturers, Number of Students, Number of Research, and Number of Graduates.

### B. Research Process

The phase of the research process can be seen in Fig. 1.

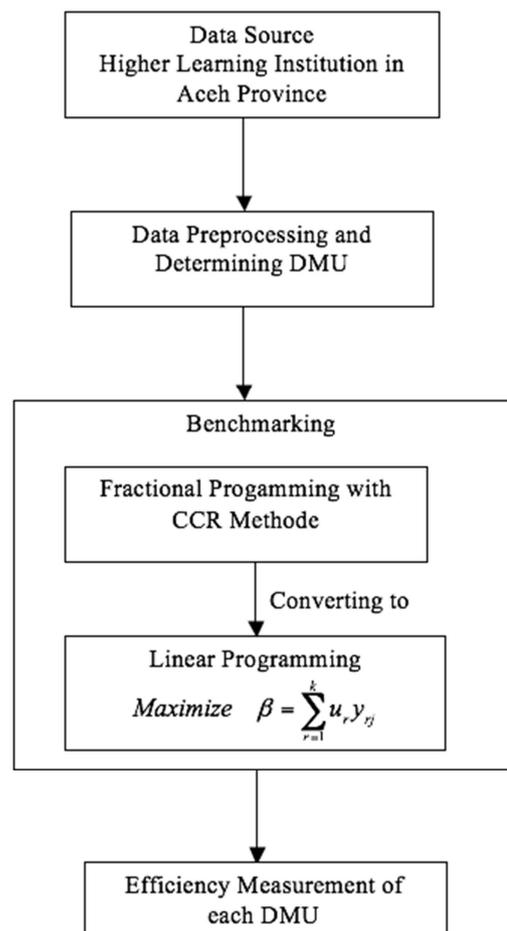


Fig. 1 Phases of the research process

### C. DEA Process

The process stages of DEA can be seen in Fig. 2 [19].

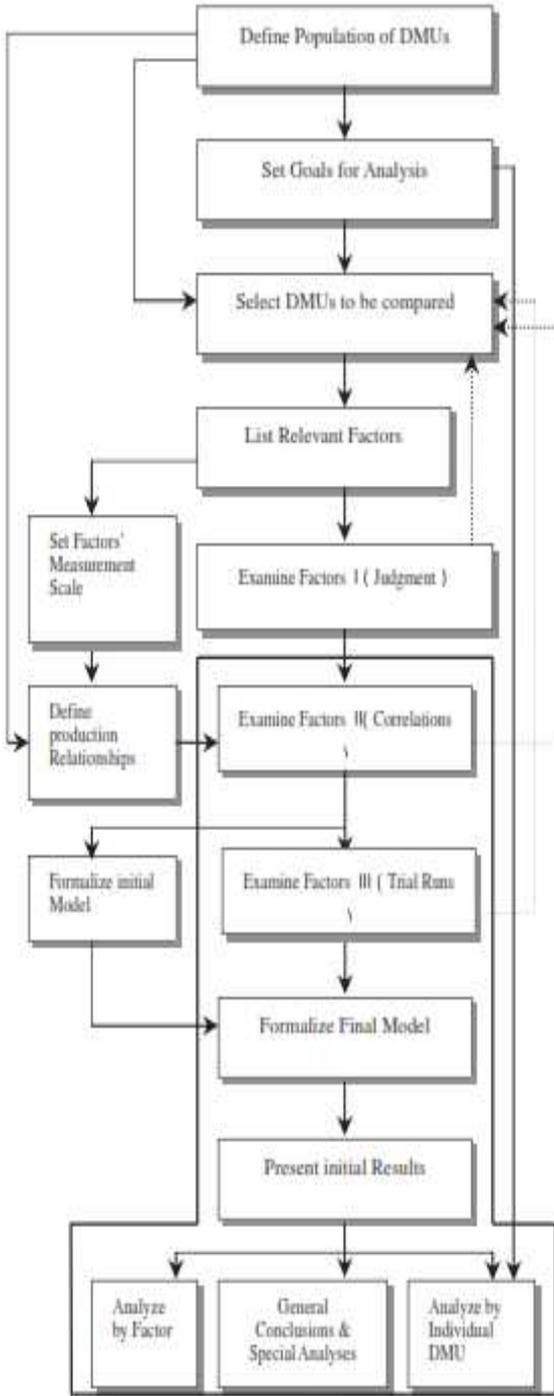


Fig. 2 Process stages of DEA

### D. Fractional Programming with CCR Method

Fractional Programming with CCR Method can be modeled by the following equation.

$$\text{Maximize } \alpha = \frac{\sum_{r=1}^k u_r y_{r0}}{\sum_{s=1}^l v_s x_{s0}} \quad (1)$$

Limit or constraint function:

$$u_r, v_s \geq 0; \quad r = 1, \dots, k; \quad s = 1, \dots, l \quad (2)$$

Where:

$\alpha$  = Efficiency of object  $s$

$k$  = output of object  $s$  observed

$y_{is}$  = the number of outputs  $i$  produced by object  $s$

$x_{js}$  = number of inputs  $i$  used by object  $s$

$u_i$  = the output weight  $i$  produced by object  $s$

$v_j$  = the input weight  $i$  given by object  $s$

In the model, there are  $n$  number of DMUs with  $k$  number of outputs resulting from  $l$  number of inputs.  $y_{rj}$  ( $>0$ ) are the number of output of the  $j^{\text{th}}$  DMU and ( $>0$ ) are the number of input of the  $j^{\text{th}}$  DMU,  $v_s$  are the variable weights to be determined after solving the model.

### E. Linear Programming with CCR Method

The model (1) is in the form of fractional programming, it would be computationally intractable particularly when the number of DMUs is large. Therefore it is necessarily to convert the model (1) into a linear programming problem, as proposed by [4], which can be written as follows (output oriented).

$$\text{Maximize } \beta = \sum_{r=1}^k u_r y_{rj} \quad (3)$$

Subject to

$$\sum_{s=1}^l v_s x_{sj} = 1;$$

$$\sum_{r=1}^k u_r y_{rj} - \sum_{s=1}^l v_s x_{sj} \leq 0; \quad \forall j$$

$$u_r, v_s \geq 0 \quad \forall r, \forall s$$

## III. FRAMEWORK OF THE OPTIMIZATION MODEL

### A. Mathematical Model Formulation

Seresht et al. [20] have proposed linear programming models to generate bounded intervals as can be seen in Equations 4 and 5.

$$\text{Maximize } \beta = \sum_{r=1}^k u_r y_{rj}$$

$$\sum_{i=1}^m w_{id} x_{id}^l - \sum_{r=1}^s u_{rd} y_{rd}^u \geq 0, j = 1, 2, \dots, n$$

$$\sum_{i=1}^m w_{id} x_{id}^u = 1$$

$$w_{id}, u_{rd} \geq \epsilon, \forall id, rd \quad (4)$$

It can be seen that in the Equation 4, DMU is in an under evaluation state in the sense that the issued input is still greater than the maximum output produced. This equation can be refined to produce an output boulder by using linear programming as can be seen in Equation 5.

$$\text{Maximize } \beta = \sum_{r=1}^k u_r y_{rj}$$

$$\sum_{i=1}^m w_{id}x_{id}^l - \sum_{r=1}^s u_{rd}y_{rd}^u \geq 0, j = 1, 2, \dots, n$$

$$\theta_d^l * \sum_{r=1}^s u_{rd}y_{rd}^l - \sum_{i=1}^m w_{id}x_{id}^u = 0$$

$$\sum_{i=1}^m w_{id}x_{id}^u = 1$$

$$w_{id}, u_{rd} \geq \epsilon, \forall_{id}, rd \quad (5)$$

DMU as can be seen in Equation 5, is in an under evaluation state in the sense that the issued input is still greater than the maximum output produced. This equation can be refined to produce an output bounder by using linear programming as can be seen in Equation 6.

$$\sum_{i=1}^m w_{id}x_{id}^u = 1$$

$$w_{id}, u_{rd} \geq \epsilon, \forall_{id}, rd \quad (6)$$

The value of  $\theta$  is essentially the result of a linear equation for the sum of the input values of each DMU. Where  $\theta$  is the upper bound interval.  $\theta$  itself can be obtained by using Equation 7.

$$\sum_{i=1}^m w_{id}x_{id}^u = 1$$

$$w_{id}, u_{rd} \geq \epsilon, \forall_{id}, rd \quad (7)$$

Where  $\theta$  is essentially the result of a linear equation for the sum of the input values of each DMU and will use as an upper bound interval.

### B. Feasible Neighborhood Heuristic Search Algorithm

The Branch-and-bound approach is a general method for solving linear integer programming problem. However, for large-scale problems such a procedure would be prohibitively expensive in terms of total computing time, and frequently the algorithm terminates without solving the problem. In reducing the problem, we assume that the integer variables is constant and only varying in the discrete step. The procedure may be summarized as follows [21] :

1) *Step 1:* Solve the problem to get the optimal solution and ignore the integral problem

2) *Step 2:* Obtain a (sub-optimal) integer feasible solution, using heuristic rounding of the continuous solution.

3) *Step 3:* Divide the set  $S$  of integer variables into the set  $S_1$  and give all the variables in their bounds, where nonbasic at the continuous solution  $S_2$  and  $S = S_1 + S_2$

4) *Step 4:* Perform a search on the objective function, maintaining the variables in  $S_1$  at their bounds and  $S_2$  as continuous variables only been allowed in discrete change

5) *Step 5:* At the solution obtained in step 4, examine the reduced input and increased output of the variables in  $S_1$  and if there are the variable that must be released in the continuous variable, released them and put them into set  $S_2$ .

For more details, suppose there are DMUs with inputs and outputs as can be seen in Table 1.

TABLE I  
LIST OF DMU WITH INPUT AND OUTPUT

No.	DMU	Input		Output	
		Number of Teaching Staff	Number of Students	Number of Research	Number of Graduates
1	Information Technology	17	588	5	610
2	Civil Engineering	26	747	5	533
3	Architectural Engineering	15	396	5	195
4	Industrial Engineering	17	467	5	300
5	Chemical Engineering	25	348	5	252
6	Mechanical Engineering	23	499	5	224
7	Electrical Engineering	19	420	5	326
8	Agribusiness	17	689	5	273
9	Agro-Technology	34	822	5	284
10	Aquaculture	10	501	5	204
11	Communication Science	11	719	5	273
12	Political Science	11	262	5	183
13	Sociology	13	487	5	204
14	Anthropology	9	173	5	116
15	Jurisprudence	50	1096	10	467
16	Medicine	30	278	4	257
17	Management	48	1265	5	1302
18	Economic Development	11	853	5	290
19	Accounting	23	1127	5	417

To perform the Upper Bound process of output with the Feasible Neighborhood Heuristic Search Algorithm can be done using Software LINDO Release 6.1 Demo Version. In the LINDO this problem can be solved using Equation (8).

Maximize  $610 U_1 + 5 U_2$   
Subject to

$$17 V_1 + 588 V_2 = 1 \quad (8)$$

$$610 U_1 + 5 U_2 - 17 V_1 - 588 V_2 \leq 0$$

$$533 U_1 + 5 U_2 - 26 V_1 - 747 V_2 \leq 0$$

$$195 U_1 + 5 U_2 - 15 V_1 - 396 V_2 \leq 0$$

$$300 U_1 + 5 U_2 - 17 V_1 - 467 V_2 \leq 0$$

$$252 U_1 + 5 U_2 - 25 V_1 - 348 V_2 \leq 0$$

$$224 U_1 + 5 U_2 - 23 V_1 - 499 V_2 \leq 0$$

$$326 U_1 + 5 U_2 - 19 V_1 - 420 V_2 \leq 0$$

$$273 U_1 + 5 U_2 - 17 V_1 - 689 V_2 \leq 0$$

$$284 U_1 + 5 U_2 - 34 V_1 - 822 V_2 \leq 0$$

$$204 U_1 + 5 U_2 - 10 V_1 - 501 V_2 \leq 0$$

$$273 U_1 + 5 U_2 - 11 V_1 - 719 V_2 \leq 0$$

$$183 U_1 + 5 U_2 - 11 V_1 - 262 V_2 \leq 0$$

$204 U_1 + 5 U_2 - 13 V_1 - 487 V_2 \leq 0$   
 $116 U_1 + 5 U_2 - 9 V_1 - 173 V_2 \leq 0$   
 $467 U_1 + 5 U_2 - 50 V_1 - 1096 V_2 \leq 0$   
 $257 U_1 + 5 U_2 - 30 V_1 - 278 V_2 \leq 0$   
 $1302 U_1 + 5 U_2 - 48 V_1 - 1265 V_2 \leq 0$   
 $290 U_1 + 5 U_2 - 11 V_1 - 852 V_2 \leq 0$   
 $417 U_1 + 5 U_2 - 23 V_1 - 1127 V_2 \leq 0$   
 $U_1, U_2, V_1, V_2 \geq 0$   
 END

We use software LINDO Release 6.1 Demo Version. The expression (3) is in LINDO format.

The result is as follows.

OBJECTIVE FUNCTION VALUE

1) 1.000000

VARIABLE	VALUE	REDUCED COST
U1	0.001639	0.000000
U2	0.000000	0.000000
V1	0.058824	0.000000
V2	0.000000	0.000000

It can be seen that DMU 1 is efficient, as the value of  $\beta$  is 1.0. The score of efficiency for all DMUs can be found in Table 2.

TABLE II  
RESULT OF EFFICIENCIES FOR EACH DMU USING OUTPUT-ORIENTED DEA

No	DMU	DEA Score
1	Information Technology	1,0
2	Civil Engineering	0,6982436
3	Architectural Engineering	0,6818709
4	Industrial Engineering	0,7045490
5	Chemical Engineering	0,8069085
6	Mechanical Engineering	0,5265533
7	Electrical Engineering	0,8263003
8	Agribusiness	0,6639550
9	Agrotechnology	0,3810771
10	Aquaculture	1,0
11	Communication Science	0,9912544
12	Political Science	0,9152225
13	Sociology	0,7845375
14	Anthropology	1,0
15	Jurisprudence	0,4226586
16	Medical	1,0
17	Management	0,9921286
18	Economic Development	1,0
19	Accounting	0,5871874

From Table 2 we would be able to observe that DMU1, DMU10, DMU14, DMU16, and DMU18 are efficient. However, if the output variable the number of research ( $y_2$ ) is increased significantly, then the corresponding DMU would be efficient. For example, DMU2 is inefficient. If  $y_2$

is increased from 5 to 15, then it would be efficient. DMU3 is inefficient formerly if we increase  $y_2$  from 5 to 10, it would be efficient. Need to know of staff for DMU2 and DMU3, respectively, are 26 and 15.

In order to control the flexibility of output variables, it is necessary to add the constraints in the model (2) with the bounded output.

$$LB_{rj} \leq u_r y_{rj} \leq UB_{rj} \quad (9)$$

Where LB is lower bound, and UB is upper bound for DMU j.

IV. CONCLUSION

The efficiency assessment of all Higher Education Institute in Aceh Province is carried out using CCR output for oriented model. The analysis of efficiency can be done by varying the output variables, while the input variables are kept fixed. Nevertheless, the output variables should be upper bounded. It is found out that the main important to be efficient is to have more number of research. This finding is synchronized with the urge from Indonesia Government about to have more research scheme.

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