

DATA ENVELOPMENT ANALYSIS OF BANKING SECTOR IN BANGLADESH

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ABSTRACT

Banking sector of Bangladesh is flourishing and contributing to its economy. In this aspect measuring efficiency is important. Data Envelopment Analysis technique is used for this purpose. The data are collected from the annual reports of twenty four different banks in Bangladesh. Data Envelopment Analysis is mainly of two types – constant returns to scale and variable returns to scale. Since this study attempts to maximize output, so the output oriented Data Envelopment Analysis is used. The most efficient bank is one that obtains the highest efficiency score.

KEY WORDS

Operations research; Farm inputs; Statistics; Assets; Methods; Sampling; Costs; Performance.

Financial institutions around the world experienced substantial changes in the last few years. Technological progress, reduced information costs, fiercer competition among both bank and non-bank financial intermediaries and ongoing deregulation all led to substantial changes in numerous financial systems. Bank efficiency has been an important issue in this transition. There are two types of methods to measure comparative efficiency: parametric and non-parametric methods. The non-parametric approaches use mathematical programming techniques (Coelli, 1996); among those Data Envelopment Analysis (DEA) is widely used. The primary focus of DEA is to measure the production or performance function of DMUs (decision making unit).

DEA evaluates the inputs consumed and outputs produced by DMUs and identify those units that comprise an efficient frontier and those that lie below this frontier. The standard DEA models have an input and output orientation. An input orientation identifies the efficient consumption of resources while holding outputs constant. An output orientation identifies the efficient level of output given existing resource consumption. The output orientation provides estimates of the amount by which outputs could be proportionally expanded given existing input levels. In addition, DEA models can be either constant or variable returns to scale (Banker et al., 1984). DEA model

can be used with very small data precisely because it is a non-parametric approach.

Efficiency of firm is measured in terms of its relative performance that is, efficiency of a firm relative to the efficiencies of firms in a sample. Data Envelopment Analysis (DEA) has used to identify banks that are on the output frontier given the various inputs at their disposal. Jackson and Fethi (2000) study on Turkish banks found that the profitable banks are more likely to operate at higher levels of technical efficiency. Seiford and Thrall (1990) found that mathematical programming procedure used by DEA for efficient frontier estimation is comparatively robust. DEA is a linear programming model introduced by Charnes et al. (1978) to measure efficiency under the assumption of constant returns to scale and extended by Banker et al. (1984) to allow variable returns to scale. A large number of papers have extended and applied the DEA methodology (Coelli, 1996).

Bhattacharyya et al. (1997) examined the productive efficiency of 70 Indian commercial banks during early stages (1986-1991) prior to liberalization. They used DEA to calculate radial technical efficiency scores. Sathye (2003) measured the productive efficiency of banks in India using DEA. The study shows that the mean efficiency score of Indian banks compares well with the world mean efficiency score. Sufian (2007) has employed the DEA method to investigate the

effects of mergers and acquisitions on the efficiency of Malaysian banks. DEA has become increasingly popular in measuring efficiency in different national banking institutes. Dwivedi and Charyulu (2011) seek to determine the impact of various market and regulatory initiatives on efficiency improvements of Indian banks.

DATA AND METHODOLOGY

The data for this study are collected from 24 banks through the annual reports of the year 2010. Some of the annual reports are obtained from respective bank's websites and some others are collected from the respective bank's headquarters. This study mainly emphasis on five variables, these are: operation profit, operation income, operation cost, total assets and deposits. For the purpose of efficiency analysis operation profit is considered as output variable that is to be maximized and the other four variables are considered as input variables.

A linear program is applied to create a virtually efficient DMU that sits on the efficiency frontier, in which each DMU has a hundred (100) percent efficiency relative to every other DMU. The linear program uses two constraints. The first constraint forces the virtual DMU to produce at least as many outputs as the studied DMU. The second constraint finds out how much input the virtual DMU would need.

Let us begin with input oriented constant returns to scale (CRS) model. Let us consider there are data on K inputs and M outputs on each of N DMU's. For the i -th DMU these are represented by the vectors x_i and y_i , respectively. The $K \times N$ input matrix X , and the $M \times N$ output matrix Y , represent the data of all DMU's. For each DMU we would like to obtain a measure of the ratio of all outputs over all inputs such as $u'y_i/v'x_i$ as efficiency score, where u is an $M \times 1$ vector of output weights and v is a $K \times 1$ vector of input weights.

Now for selecting optimal weights we write the mathematical programming problem as follows:

$$\begin{aligned} & \max_{u,v} (u'y_i/v'x_i), \\ & \text{such that } u'y_j/v'x_j \leq 1, \quad j = 1, 2, \dots, N, \\ & u, v \geq 0 \end{aligned} \quad (1)$$

This involves finding values for u and v such that the efficiency measure of the i -th DMU

is maximized, subject to the constraint that all efficiency measures must be less than or equal to one. A great problem with this particular ratio formulation is that it has an infinite number of solutions. To avoid this we can impose the restriction $v'x_i = 1$, which gives:

$$\begin{aligned} & \max_{\mu,v} (\mu'y_i), \\ & \text{such that } v'x_i = 1, \\ & \mu'y_j - v'x_j \leq 0, \quad j = 1, 2, \dots, N, \\ & \mu, v \geq 0, \end{aligned} \quad (2)$$

where the notation change from v to μ and v is reflecting the transformation here. This form is called the multiplier form of the linear programming problem. Using the duality in linear programming; one can derive an equivalent envelopment form of this problem:

$$\begin{aligned} & \min_{\theta,\lambda} \theta, \\ & \text{such that } -y_i + Y\lambda \geq 0, \\ & \theta x_i - X\lambda \geq 0, \\ & \lambda \geq 0, \end{aligned} \quad (3)$$

where θ is a scalar and λ is a $N \times 1$ vector of constants. This envelopment form involves fewer constraints than the multiplier form ($K+M < N+1$), and hence is generally the preferred form to solve. The value of θ obtained will be the efficiency score for the i -th DMU. The value of θ obtained will be the efficiency score for the i -th DMU. It will satisfy $\theta \leq 1$, with a value of 1 indicating a point on the frontier and thus a technically efficient DMU is obtained according to the Farrell (1957) definition.

The CRS linear programming problem can be easily modified to account for variable returns to scale (VRS) by adding the convexity constraint: $N1'\lambda = 1$ to (3) to provide:

$$\begin{aligned} & \min_{\theta,\lambda} \theta, \\ & \text{such that } -y_i + Y\lambda \geq 0, \\ & \theta x_i - X\lambda \geq 0, \\ & N1'\lambda = 1 \\ & \lambda \geq 0, \end{aligned} \quad (4)$$

where $N1$ is an $N \times 1$ vector of ones. This procedure provides TE (technical efficiency) scores which are greater than or equal to those obtained using the CRS model.

Scale efficiency is calculated as follows:

$$\text{Scale Efficiency} = (\text{TE obtained from CRS} / \text{TE obtained from VRS})$$

One limitation of this measure of scale efficiency is that the value does not indicate whether the DMU is operating in the area of increasing or the decreasing returns to scale. This may be determined by running an addition DEA problem with non-increasing returns to scale (NIRS) imposed. This can be done by altering the DEA model in equation 4 by substituting the $N1'\lambda = 1$ restriction with $N1'\lambda \leq 1$, to provide:

$$\begin{aligned} & \min_{\theta, \lambda} \theta, \\ & \text{such that } -y_i + Y\lambda \geq 0, \\ & \quad \theta x_i - X\lambda \geq 0, \\ & \quad N1'\lambda \leq 1 \\ & \quad \lambda \geq 0, \end{aligned} \quad (5)$$

The nature or characteristics of the scale inefficiencies (due to increasing or decreasing returns to scale) for a particular DMU can be determined by seeing whether the NIRS TE score is equal to the VRS TE score. The output-oriented models are very similar to their input-oriented

counterparts. Let us assume the example of the following output-oriented VRS model:

$$\begin{aligned} & \max_{\phi, \lambda} \phi, \\ & \text{st } -\phi y_i + Y\lambda \geq 0, \\ & \quad x_i - Y\lambda \geq 0, \\ & \quad N1'\lambda = 1 \\ & \quad \lambda \geq 0, \end{aligned} \quad (6)$$

where $1 \leq \phi < \infty$, and $\phi - 1$ is the proportional increase in outputs that could be achieved by the i -th DMU, with input quantities held constant. Note that, $1/\phi$ defines a TE score which ranges between zero and one. One point that should be noted is that the output- and input-oriented models will estimate the same frontier and therefore, identify the same set of DMU's as being efficient.

RESULTS AND INTERPRETATIONS

The following table shows the descriptive statistics of the sample $n=24$ banks.

Table 1 – Descriptive Statistics of the Banks*

Variables	Mean	Median	Minimum	Maximum	Standard deviation
Operation profit	5030.10	4164.50	994.50	12080.80	2987.90
Operation income	8201.00	6880.00	2633.00	18742.00	4545.00
Operation cost	3172.00	2635.00	1280.40	7862.00	1893.90
Assets	123860.00	98050.00	55170.00	345710.00	79604.50
Deposits	99680.00	78670.00	44850.00	291930.00	67749.90

*The above table is figured in units of million BDT taka (currency of Bangladesh with 1 USD = 81.8 BDT)

Under the CRS assumption both the output and input oriented technical efficiency scores are same. Here, all the efficiency scores of DEA are obtained using the DEAP- xp1 software developed by Tim Coelli (1996). All the CRS (output), VRS (output) and scale efficiency scores of the banks along with the peers are given in the table 2.

Table 2 shows the technical efficiency scores for all the DMUs. Here we see that under the CRS output results only three banks- SOUTHEAST BANK LIMITED (DMU 12), NATIONAL BANK LIMITED (DMU 19) and AB BANK LIMITED (DMU 22) are technically efficient because they have the technical efficiency scores equal to one. We note that the technical efficiency (TE) of DMU 1 is 0.879. That is DMU 1 should be able to increase the operating profit by 12.1%

without increasing inputs. Similar interpretation holds for the other DMUs. A remarkable thing is that DMU 12 is recognized as most efficient bank because the linear combination of DMU 12 is more used than DMU 19 or DMU 22 as peer. So, using CRS output oriented multi stage DEA, the DMU 12 is most efficient though all of DMUs 12, 19 and 22 have technical efficiency score equal to one.

Under the VRS output results the 12 banks- EXIM BANK (DMU 7), MUTUAL TRUST BANK LIMITED (DMU 9), ONE BANK LIMITED (DMU 10), TRUST BANK LIMITED (DMU 11), SOUTHEAST BANK LIMITED (DMU 12), JAMUNA BANK LIMITED (DMU 14), SOCIAL ISLAMI BANK LIMITED (DMU 17), EASTERN BANK LIMITED (DMU 18),

NATIONAL BANK LIMITED (DMU 19), JANATA BANK LIMITED (DMU 20), AB BANK LIMITED (DMU 22), and AGRANI BANK LIMITED (DMU 23) are technical efficient. Other 12 banks are technically inefficient as their efficiency scores are less than one. Technical efficient DMUs are peer of themselves only. We know that only the efficient DMUs form the linear combinations for the inefficient DMUs for effi-

ciency perspective. For example, DMU 15 is a linear combination of the DMUs 10, 18 and 19. That is, this linear combination of and 10, 18 and 19 determines the efficient output of DMU 15. The peer counts for DMU 12 is 9 whereas for DMU 10 it is 6 and for the other efficient DMUs 7, 18, 19, 20 and 22 the peer counts are 4, 4, 6, 2 and 1 are respectively. Since DMU 12 is most used, so the most efficient bank is DMU 12.

Table 2 – Efficiency Scores obtained using DEA*

DMUs	CRS TE	CRS Peers	VRS TE	VRS Peers	SCALE TE	–
1	0.879	22(0.041), 12(0.596)	0.952	12(0.270), 7(0.038), 10(0.693)	0.924	irs
2	0.829	12(0.180), 19(0.430)	0.873	10(0.699), 19(0.301)	0.949	irs
3	0.785	22(0.449), 12(0.232)	0.805	19(0.148), 10(0.190), 12(0.106), 18(0.555)	0.975	irs
4	0.606	19(0.783)	0.638	10(0.386), 19(0.614)	0.951	irs
5	0.759	19(0.228), 22(0.507), 12(0.156)	0.762	19(0.362), 22(0.222), 12(0.131), 18(0.285)	0.996	irs
6	0.844	19(0.119), 22(0.067), 12(0.990)	0.889	12(0.631), 20(0.105), 19(0.264)	0.949	irs
7	0.891	12(0.519)	1.000	7(1.000)	0.891	irs
8	0.691	19(0.125), 12(0.491)	0.731	10(0.597), 12(0.252), 18(0.151)	0.944	irs
9	0.740	12(0.400)	1.000	9(1.000)	0.740	irs
10	0.907	12(0.289), 19(0.154)	1.000	10(1.000)	0.907	irs
11	0.786	12(0.428)	1.000	11(1.000)	0.786	irs
12	1.000	12(1.000)	1.000	12(1.000)	1.000	–
13	0.749	12(0.561)	0.844	12(0.082), 7(0.412), 10(0.506)	0.887	irs
14	0.809	12(0.440)	1.000	14(1.000)	0.809	irs
15	0.859	12(0.122), 19(0.407)	0.994	10(0.638), 18(0.294), 19(0.068)	0.864	irs
16	0.768	12(1.840)	0.913	12(0.301), 20(0.699)	0.841	drs
17	0.475	12(0.310)	1.000	17(1.000)	0.475	irs
18	0.980	22(0.479), 12(0.099)	1.000	18(1.000)	0.980	irs
19	1.000	19(1.000)	1.000	19(1.000)	1.000	–
20	0.810	12(2.203)	1.000	20(1.000)	0.810	drs
21	0.642	12(0.563)	0.710	12(0.091), 7(0.794), 10(0.115)	0.904	irs
22	1.000	22(1.000)	1.000	22(1.000)	1.000	–
23	0.816	22(0.126), 12(1.812)	1.000	23(1.000)	0.816	drs
24	0.757	12(0.923)	0.764	12(0.841), 7(0.159)	0.990	irs
MEAN	0.808	–	0.911	–	0.891	–

* Each bank is a DMU and the peer weights are given within bracket with peers.

The VRS efficiency results also give output scale efficiency scores with VRS efficiency scores. A DMU is considered as scale efficient if its output scale efficiency score is equal to one. Only three banks- SOUTHEAST BANK LIMITED (DMU 12), NATIONAL BANK LIMITED (DMU 19) and AB BANK LIMITED (DMU 22) are scale efficient as their output scale efficiency scores are equal to one. So, in common we get three banks DMU 12, DMU 19 and DMU 22 which are efficient under both CRS and VRS assumption and they are scale efficient too.

Whether the DMU is operating in an area of increasing or decreasing returns to scale can be checked by running an additional DEA problem

with non-increasing returns to scale (NIRS) imposed. If the NIRS TE score and VRS TE score are unequal for a DMU, then increasing returns to scale (IRS) exist for that DMU. For our given data, DRS exist for DMU 6, 16, 20 and 23. IRS exists for the remaining 17 DMUs other than 12, 19 and 22.

The peer weights give the weights to construct a linear combination of the efficient banks to represent an inefficient one. The descriptive statistics of the technical efficiency scores obtained from these methods are given in table 3.

The mean and median of TE scores of CRS DEA is smaller than other two methods. Maximum TE score is one for all methods but mini-

imum score is not same for all the three methods. The range (maximum-minimum) is biggest for CRS DEA and smallest for VRS DEA. Standard

deviation of TE scores also reflects this. A graphical comparison of CRS TE scores, VRS TE scores and scale TE scores is shown in figure 1.

Table 3: Descriptive Statistics of TE Scores for n=24 Banks

Methods	Mean	Median	Maximum	Minimum	Standard Deviation
CRS DEA	0.8076	0.8095	1.0000	0.4750	0.1276
VRS DEA	0.9086	0.9970	1.0000	0.6380	0.1223
SCALE	0.8912	0.9155	1.0000	0.4750	0.1170

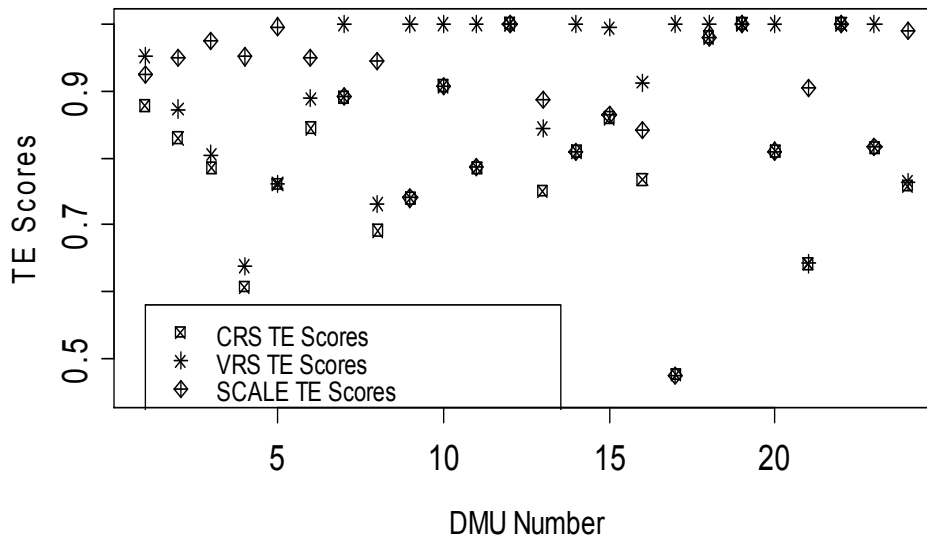


Figure 1 – Comparison Graph of CRS DEA, VRS DEA and Scale TE Scores

CONCLUSION

Here CRS DEA gives 3 efficient banks (DMUs 12, 19 and 22), VRS DEA gives 12 efficient banks (DMUs 7, 9, 10, 11, 12, 14, 17, 18, 19, 20, 22 and 23). Now we can rank the banks according to their efficiency scores. The banks with higher technical efficiency possess top ranks. Here few banks have efficiency scores equal to one and their ranking can be determined by considering peer counts. However, the most efficient bank is SOUTHEAST BANK LIMITED (DMU 12) which is valid for all methods applied. This is because in CRS DEA efficient DMU 12 has 20 peer counts while other efficient DMU 19 and 22 have only 7 and 6 respectively. DMU 12 possesses rank 1 and DMU 19 and DMU 22 have rank 2 and rank 3 respectively. Other rankings are done according to the decreasing value of the TE scores. Similar ranking can be provided in VRS DEA case. Here efficient DMUs 7, 10, 12, 18, 19, 20 and 22 possess ranking 4, 2, 1, 5, 3, 6 and 7 respectively.

The study found that CRS-DEA consists of 3 efficient banks and the range of the efficiency scores is too large whereas VRS-DEA consists of

12 efficient banks and the range of efficiency scores is smaller than CRS-DEA. So, it may be inappropriate to use CRS-DEA instead of VRS-DEA in this case. Again, VRS assumption overcomes the shortcoming of CRS assumption which supports the idea of Banker et al. (1984) and seems to be more appropriate.

However, as mentioned by Avkiran (1999), DEA provides insights on which areas need to be improved but it does not have information on how to improve. Further investigations are needed in order to identify approaches for each bank to increase operation profit by moving towards the efficient frontier.

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