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Efficiency analysis of Serbian banking sector using data envelopment analysis: prepandemic era

Analiza efikasnosti bankarskog sektora pomoću metode analize obavijenosti podataka: pre pandemijski period

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Sažetak

Abstract

During the last decade, the question of measuring efficiency became inevitable factor of successful management of banks. Therefore, the basic goal of this paper is to apply DEA - Data Envelopment Analysis in comparing unit efficiency of Serbian banking sector for three years period of time (2017, 2018 and 2019). DEA is a technique of mathematical programming which uses analysis and comparation of efficiency of different entities and it is suitable to solve problems of different number and nature of inputs and outputs. The presented paper employs an output-oriented data envelopment analysis with a constant return to scale assumption. The selected input variables are total assets and total capital, while net operating income and net profit were used as output variables. Results showed that the average efficiency of Serbian banking sector is about 50% in the observed years. Opportunity Bank achieved the maximal efficiency score in the entire time period while Mirabank had the lowest efficiency score.

Keywords: data envelopment analysis, efficiency, banking sector

Tokom poslednje decenije, pitanje merenja efikasnosti postalo je neizbežan faktor uspešnog upravljanja bankama. Stoga je osnovni cilj ovog rada da se primeni DEA - Data Envelopment Analysis u poređenju efikasnosti bankarskog sektora Srbije za trogodišnji period (2017, 2018. i 2019.). DEA je tehnika matematičkog programiranja koja koristi analizu i poređenje efikasnosti različitih entiteta i pogodna je za rešavanje problema različitog broja i vrste ulaznih i izlaznih jedinica. Predstavljeni rad koristi analizu obuhvata podataka orijentisanu na izlaz. Odabrane ulazne varijable su ukupna aktiva i ukupan kapital, dok su neto poslovni prihod i neto dobit korišćeni kao izlazne varijable. Rezultati su pokazali da je prosečna efikasnosti bankarskog sektora Srbije oko 50% u posmatranim godinama. Opportuniti banka je ostvarila maksimalnu ocenu efikasnosti u čitavom vremenskom periodu, dok je Mirabanka imala najnižu ocenu efikasnosti.

Ključne reči: metode analize obavijenosti podataka, efikasnost, bankarski sektor

1. Introduction

Banks, like any other organization, strive to achieve the highest possible results with as little investment as possible. Unlike other organizations, banks have a special role in the economy of one country, given that they are the driver of economic development. Banks have come a long way in their development from traditional to industrial and finally international. Consequently, today's banks are significantly different from traditional ones - they used to deal mainly with lending, deposit and exchange operations, while today banks are mostly universal and offer a wide range of services, which greatly facilitate business and offer numerous legal benefits. There are 26 banks operating in Serbia, most of which are foreign private banks. The directions in which this sector is moving are concentration, technological innovation and deregulation. One of the trends in the development of

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banking in Serbia is the concentration of capital, most often through mergers and acquisitions. The number of banks is decreasing from year to year, and bearing in mind that some of the banks are operating at a loss, there is speculation about the need to further reduce this number. When performing mergers and acquisitions, owners are most interested in the end results, which often show improvement. On the other hand, employees in banks that participate in integration most often suffer the consequences due to the fact that a large number of workers are fired due to overlapping positions.

An important principle in the business of any company, including banks, is the principle of efficiency, which can be defined as the ability to achieve the desired goals with minimal use of available resources. A bank that operates efficiently invests in highly profitable investments, grants loans to clients of good creditworthiness and thus has a smaller number of non-performing loans. Also, such a bank uses modern technology and thanks to that employs fewer workers. On the other hand, the bank offers innovative and quality services for which it charges commissions that represent additional income. In this paper, banks are considered as financial organizations that have funds in the form of total assets and capital, on the basis of which they strive to achieve maximum net operating income and net profit. Therefore, the degree of efficiency was determined as the ratio of output and input variables, where total assets and capital were used as input variables, and net operating income and net profit as output variables. These data were analysed using Data Envelopment Analysis (DEA), which is a well-known non-parametric method for efficiency measurement and evaluation.

The main hypothesis is that a large number of banks in Serbia operate with an unsatisfactory level of efficiency. The problem that needs to be examined is the way in which banks can influence the increase of efficiency levels, and the subject of research is the degree of efficiency of their operations and the necessary correction of input and output values in each of the banks to operate efficiently.

The importance of the analysis is multiple and is reflected in the fact that it provides insight into the efficiency of the banking sector, as well as each bank individually. At the same time, the obtained coefficients are comparable with the coefficients of other banks. Given that the financial statements from which the necessary data are drawn are standardized, and the selected balance sheet items are common to each organization that makes them, the comparison of efficiency can be done between banking groups, countries and other non-banking companies or sectors.

2. Literature review

In recent years there has been a noticeable increase in the interest of researchers in quest for drivers of efficiency in the banking sector. Also, a broad methodology has been applied and a numerous combinations of determinants have been identified to examine the efficiency of banks. Feroz et al. (2003) indicate that data envelopment analysis can provide a consistent and reliable measure of managerial or operational efficiency. In terms of financial statement analysis, they demonstrated that DEA can augment the traditional ratio analysis. The estimated targets for improvement of the inefficient banks can help top management in better resource allocation and faster decision-making process that ultimately lead to greater financial stability (Cvetkoska et al., 2021).

Chiu et al. (2009) investigate 43 Taiwanese banks for the period 1998 to 2002. Their study results show that loan quality factors have an impact on a bank's efficiency and bank with a higher performance will be more secure. Titko et al. (2014) attempt to extract from the larger set of inputs and outputs only the most important ones by processing the data using correlation analysis, linear regression analysis, analysis of mean values and two-sample Kolmogorov-Smirnov test. The regression analysis was used to substantiate variables selection. The results show that in time of financial turbulence it is better to use profitoriented variables as outputs instead of loans. Kumar and Singh (2014) noticed that the major factor resulting in poor performance of large Indian banks is their huge amount of deposits and operating expenses. Alfiero et al. (2017) using the non-parametric Data Envelopment Analysis - Slack Based Model methodology, recognized general growth of technical efficiency, managerial efficiency and scale efficiency in case of Italian saving banks over the 2010-2015 period. Floros et al. (2020) using recent data from PIGS countries which suffer from debt issue. The study results confirm that the choice of inputs and outputs does matter in the case of European banking efficiency, especially because of the controversial role of deposits as input or output. Antunes at al. (2021) suggests a new DEA model to assess the efficiency of 39 Chinese commercial banks over the period 2010–2018. The results indicate that the state-owned banks have the highest efficiency, the rural banks are found to be least efficient, while the foreign banks experience the strongest volatility over the examined period.

If we take into account the application of DEA methods in the analysis of the efficiency of the banking sector of Serbia, it cannot be noticed a large number of papers on this topic. Mihailović et al. (2009) conducts an analysis of the efficiency of banks in Serbia in 2005 using as inputs total asset, total capital and number of employees and as outputs interest revenue and revenue before taxation. They used data envelopment analysis, as one of two contacted methods, to compare and rank banks in Serbia. Savić et al. (2012) applied DEA Window analysis to assess the efficiency of Serbian banks based on panel data sample from 2005 to 2011. The analysis consists of inputs such as number of employees, fixed assets and intangible investments, capital and deposit and outputs such as granted loans and deposits and non-interest income. Knežević et al. (2012), reffering to developing countries, remark the imporatance of monitoring the efficiency of banks and compare it with countries in the region. Lukić et al. (2017) emphasize the need for efficient management of assets and liabilities, human resources, capital, operating income and profit in most Serbian banks.

Marjanović et al. (2018) implement the input-oriented DEA CRS/VRS models based on financial performance of commercial banks in Serbia in period 2014-2016 by using as inputs total assets, interest expenditure, operating expenditure and number of employees and as outputs interest income and profit before tax. Their findings point to the fact that two-thirds of banks operate in conditions of inefficiency due to operational and managerial problems. Radojičić et al. (2018) present the results of indepth analysis of the efficiency of Serbian banks during the period 2005–2016. In terms of average efficiency, the banking sector of Serbia displays an improving trend over the observed period.

3. Methodology

Commonly used methodologies in efficiency measurement are the parametric approach and the nonparametric approach. Parametric approaches to efficiency analysis are based on measures of central tendency while the efficiency of each decision-making unit (DMU) is compared with average values. On the contrary, nonparametric approach (such as DEA) compares the efficiency of each individual entity with the highest achieved level of efficiency, rather than the average. It does not require a priori an assumption about the analytical form of the observed inputs and outputs. A significant advantage of the DEA method is that a different number of heterogeneous inputs and outputs can be used in the analysis, which can be represented by different types of metrics (Marcikic Horvat & Budimcevic, 2018).

There are four main phases in conducting an efficiency study using the DEA method:

- Defining and selecting decision-making units whose relative efficiency should be determined;
- Determining input and output variables that are relevant and suitable for assessing the relative efficiency of selected decision-making units;
- Selection of an adequate DEA model;
- Solving DEA models, analysis and interpretation of results.

In the first phase, it is actually decided what will be the subject of the analysis. In order to be able to perform the analysis, it is necessary to determine which data refer to investments and which to results. This data must be expressed numerically in order to be able to compare with each other. After that, it is necessary to determine the primary goal of decision-making units and, based on it, determine whether they strive to minimize input or maximize output variables. The previous phases create preconditions for solving the model and analysis of the solution, which should lead to certain conclusions regarding the level of efficiency and ways of improvement.

In order to determine the efficiency of the banks involved in the analysis, it is necessary to determine which input and output values are adequate in order to obtain the relevant parameters. Input variables are variables that are invested and consumed during the reproduction process in order to result in output values. Given the service character of banks, the process of reproduction consists of investing a certain amount of money, which is transformed and increased through work. The data were retrieved from the Annual Report published at the website of The National Bank of Serbia for the for-time period of three years (2017, 2018 and 2019). Based on the comprehensive analysis of previous researches and the main objective of this study, two output and two input variables were selected for DEA model. Net operating income and profit were used as output variables, while total assets and total capital were used as input variables. All variables are presented in national currency Serbian dinar (RSD). Input and output variables are presented in Table 1 in thousands RSD.

Table 1. Input and output variables in thousands RSD

Bank		Total	Net	Net
(DMU)	Total assets	Capital	operating	profit
2019 Addiko bank	94.756.920	22.386.667	income 5.291.238	1.107.695
AlK bank	223.600.839	57.319.673	12.357.305	8.230.318
API bank	9.443.082	1.413.215	2.097	-240.795
Sberbank	157.360.045	25.304.344	4.164.398	6.113.883
Srpska banka	13.744.154	3.593.816	493.955	306.155
Unicredit banka	441.577.253	85.846.985	13.420.579	7.590.686
Vojvođanska banka	215.048.577	31.842.377	7.609.313	5.982.870
Sberbank	157.360.045	25.304.344	4.164.398	6.113.883
Srpska banka	13.744.154	3.593.816	493.955	306.155
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Bank	m . 1 .	Total	Net	Net
(DMU) 2018	Total assets	Capital	operating income	profit
Addiko bank	100.118.520	21.397.933	5.370.547	1.299.958
AIK bank	206.558.803	56.288.856	10.399.446	5.642.948
API bank	10.256.179 571.075.179	1.637.651 106.601.186	523.890	30.358 12.604.110
Sberbank Srpska	27.398.407	1.903.525	27.037.379 531.786	12.604.110 84.191
banka Unicredit		1.905.325		04.191
banka	98.983.728	11.378.842	4.641.212	752.946
Vojvođanska banka	59.501.046	10.278.876	1.261.350	6.163.444
Sberbank	202.932.013	23.631.800	8.798.731	2.918.483
Srpska banka	169.532.493	52.534.819	8.484.326	2.246.813
Unicredit banka	15.509.148	3.882.374	876.104	139.561
Vojvođanska banka	56.905.090	11.151.685	2.394.904	361.725
Sberbank	15.623.162	3.894.268	726.124	465.279

Srpska banka	401.165.980	67.560.513	19.003.608	8.145.182
Unicredit banka	5.937.892	2.133.010	176.725	-322.469
Vojvođanska	15.930.961	1.494.829	840.347	-601.924
banka Sberbank	10.776.218	2.430.479	653.892	6.484
Srpska	57.459.595	8.082.158	3.218.647	620.816
banka Unicredit	51.457.575	0.002.150	5.210.047	020.010
banka	16.146.648	3.127.726	1.975.517	426.452
Vojvođanska banka	316.087.334	44.238.385	16.112.801	8.155.854
Sberbank	220.036.784	19.517.180	9.013.078	2.225.767
Srpska banka	111.057.393	15.917.579	3.783.586	996.082
Unicredit	291.924.334	57.908.928	16.496.139	6.628.209
banka Maina tamaka	291.924.334	57.908.928	10.490.139	0.028.209
Vojvođanska banka	121.659.363	24.688.423	4.903.776	1.309.826
Sberbank	12.887.132	3.340.172	361.847	21.696
Srpska banka	437.400.289	77.828.498	18.136.908	9.221.647
Unicredit	87.602.369	26.323.288	3.559.920	285.955
banka Bank	07.002.507		Net	
(DMU)	Total assets	Total Capital	operating	Net profit
2017	04 115 010		income	-
Addiko bank AIK bank	94.115.810 209.359.749	21.945.443 61.161.912	5.727.755 18.337.117	1.218.248 12.080.194
API bank	10.671.360	1.634.994	286.849	-173.756
Sberbank	564.859.578	119.006.802	27.017.436	11.847.694
Srpska banka	1.863.780	1.841.243	121.588	-2.644
Unicredit banka	86.933.902	9.091.129	1.936.203	345.102
Vojvođanska	26.870.287	3.949.539	3.170.464	1.902.074
banka Sberbank	161.921.004	20.894.563	1.932.260	2.632.240
Srpska	158.441.193	51.199.228	8.129.666	2.230.094
banka Unicredit	150.441.175	51.177.220	0.129.000	2.230.074
banka	14.999.131	3.793.039	2.903.031	1.807.884
Vojvođanska banka	41.449.177	6.119.707	1.880.519	194.631
Sberbank	11.706.404	3.489.499	585.756	62.546
Srpska banka	369.183.538	63.260.055	18.576.152	8.117.368
Unicredit	6.022.727	2.468.710	149.959	-485373
banka Maina tamaka	0.022.727	2.408.710	149.939	-405575
Vojvođanska banka	14.925.253	1.714.100	607.734	-1.090.752
Sberbank	7.756.956	2.145.329	523.746	7.113
Srpska banka	44.121.316	7.388.340	3.242.085	442.574
Unicredit banka	2.733.218	14.123.297	1.712.694	353.086
Vojvođanska	287.987.802	40.404.589	13.544.830	6.030.737
banka Sberbank	140.488.253	15.874.902	10.059.504	3.879.268
Srpska	96.682.054	14.833.361	4.735.689	1.189.346
banka Unicredit	90.082.034	14.055.501	4.755.089	1.109.540
banka	264.621.343	57.443.503	16.536.378	6.626.785
Vojvođanska banka	116.032.677	24.123.795	4.296.726	802.911
banka Sberbank	14.100.809	3.297.724	425.838	29.216
Srpska	362.748.552	69.340.495	14.498.628	6.633.319
banka Unicredit	73.255.431	27.199.182	2.657.022	-458.856
banka	,0.200.101	2	2.007.022	.20.050

What is clear at first glance, in Table 1, is that the input values significantly exceed the values of the output values, which in some cases are even negative. The largest value of input in 2019 has Banca Intesa (as much as 752,697,959 thousand RSD), followed by Unicredit Bank and Komercijalna Bank with significantly lower values (527,424,238 and 508,232,616 thousand RSD). Mirabank (only 8,340,603 thousand RSD), API Bank and Bank of China (10,856,297 and 12,253,481 thousand RSD) have the lowest input values. When it comes to output values in the same year, the highest value is also recorded by Banca Intesa (40,399,233 thousand RSD), as well as

Komercijalna banka and Raiffeisen banka. API Bank and Mirabanka have the lowest output values - these banks even reach negative exit values (-238,698 and -73,263 thousand RSD). In both cases, the downward trend in output is the result of a loss that is higher than net operating income. This is largely due to excessive reliance on interest income with significantly lower fee and commission income, which has an increasing share in the modern bank. Also, it is noticeable that these banks have extremely high labor costs, which are almost equal to the realized net operating profit.

Nowadays, various variants of DEA model are developed in different areas of application. The most popular are CCR (after Charnes, Cooper and Rhodes, developed in 1978) and BCC models (after Banker, Charnes and Cooper, developed in 1984) with the main difference in modelling return to scale. CCR model assumes constant return to scale, while BCC model allows variable return to scale. Efficiency measurement for each decision-making unit (DMU) in CCR model is attained as a maximum of a ratio of weighted sum of outputs to weighted sum of inputs. The solution of the following linear programming model provides the efficiency measure for DMU (Jemric & Vujcic, 2002):

$$max_u z_0 = \sum_{r=1}^{3} u_r y_{r0}$$

subject to:

$$\sum_{r=1}^{s} u_r y_{rj} - \sum_{\substack{i=1\\i=1}}^{m} v_i x_{ij} \le 0, j = 1, 2 \dots, n$$
$$\sum_{\substack{i=1\\i=1}}^{m} v_i x_{i0} = 1$$
$$u_r \ge 0, r = 1, 2, \dots s$$
$$v_i \ge 0, i = 1, 2, \dots m$$

where x_{ij} is the perceived input amount of the ith type of the jth DMU and y_{rj} is the perceived output amount of the rth type for the jth DMU.

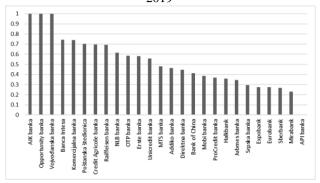
In our study, we applied the output-oriented CCR model to calculate and compare the relative technical efficiency of Serbian banking sector. The choice of DEA model orientation depends on whether decision-makers have more influence on improving input or output levels. The output-oriented model tries to determine the maximum possible proportional increase of outputs while keeping the levels of used inputs constant (Radovanov et al., 2020). Therefore, output oriented CCR DEA method is applied, because banks do not aim to reduce the total assets or capital. Instead, the bank will strive to achieve higher net operating income and profits in order to improve efficiency level.

4. Results and discussion

Achieved average efficiency scores are presented in Figures 1, 2 and 3. The results of efficiency scores lie between 0 and 1 i.e., between 0% and 100%. It is evident that when it comes to the absolute amount of total assets, capital, net operating income and profit, Banca Intesa,

UniCredit Bank and Komercijalna Banka dominate, while the other observed banks have significantly lower values. Results of applied DEA method provide a clear answer to the question of whether high absolute amounts of these balance sheet items imply high business efficiency.

Figure 1. Efficiency scores of Serbian banking sector in 2019



Results presented in Figure 1 show that in 2019 three operated efficiently: Vojvodjanska banka, banks Opportunity banka and AIK banka. The lowest efficiency score has API bank 0.0024 (0.24%). The average efficiency of the banking sector of Serbia in 2019 was 0.5196 (51.96%). This value is slighly higher 0.5403 (54.03%) if we exclude the efficiency score achieved by API bank, which is dramatically lower compared to the others. This indicates that improvement can be made in the given sector because it operates at a level that represents only slightly more than half of the possible capacities. In this period, 12 banks operated with aboveaverage efficiency, five banks operated with efficiency over 70%, and two banks operated with the efficiency below 30%.

Figure 2. Efficiency scores of Serbian banking sector in 2018

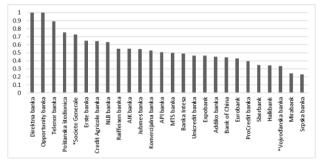


Figure 2 shows the obtained efficiency scores in 2018. Direct Bank and Opportunity Bank operate under maximal efficiency in this year. In the last place according to the level of efficiency is Srpska banka with a efficiency score of 0.2295 (22.95%). The average efficiency score in this period was 0.5421 (54.21%), which is 0.0225 (2.25%) higher than in 2019, which means that the average efficiency of the Serbian banking sector decreased in the observed period. Within the banking sector of Serbia, 11 banks operated in 2018 with the level of efficiency above the average. Efficiency over 70% is achieved by four banks, while two banks operate with the efficiency score below 30%.

Figure 3. Efficiency scores of Serbian banking sector in 2017

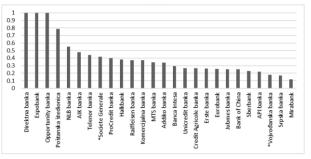


Figure 2 shows the obtained efficiency scores in 2017 where three banks achieved 100% efficiency: Direktna banka, Expobanka and Opportunity banka. The lowest efficiency score is achieved by Mirabanka with a value of 0.1201 (12.01%). The average efficiency score in 2017 is 0.4098 (40.98%), which is the lowest value in the observed three years. Efficiency above this average is achieved only by 8 banks. Efficiency score over 70% is achieved by four banks, while 12 banks operate with efficiency below 30%.

The following figure (Figure 4) refers to the distribution of average efficiency score of Serbian banking sector during observed period of time (2017, 2018 and 2019).

Figure 4. Average efficiency scores in the entire time period

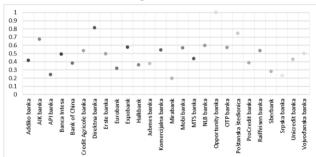


Figure 4 shows that only Opportunity Bank operated with 100% efficiency in the entire period of time. In second and third place are Direct Bank with an average efficiency of 0.8154 (81.54%) and Postanska Stedionica Bank with an average efficiency of 0.7483 (74.83%). With an average efficiency level below 30% operated Sberbank, API Bank, Srpska banka and finally Mirabanka, which operates with an average efficiency below 20% (0.1980 or 19.80%). Overall, efficiency of banks can change dramatically over observed period of time (Bulajić et al., 2013).

It is interesting to note that the banks that operate with the highest efficiency level are not the ones that have the highest absolute amounts of input, but vice versa. The best example of this is Opportunity Bank. Banks that have a significantly higher amount of capital and assets are located mainly in the group of banks with an efficiency of 50% to 70%. This suggests that large banks in Serbia generate a smaller percentage of outputs from available inputs compared to smaller banks. It can also be notices that the large number of banks with medium to high

average efficiency in this three years period are domestically owned. The reason for this may be the compulsion of domestic banks to make the best use of available resources because, unlike foreign banks, they cannot withdraw funds from the parent bank abroad. Also, smaller state-owned banks are less efficient than larger domestic banks who are successfully struggling with foreign competition (Bulajić et al., 2013).

The most efficient banks on the territory of Serbia in the three-year period were Opportunity Bank (operates efficiently in all three years), Direct Bank (operated efficiently in two of three years), AIK Bank, Vojvodjanska Bank and Expobank (operates efficiently in one of three years). From the above, it can be concluded that smaller banks that are oriented towards small clients - small businesses, individuals and households - operate far more efficiently in Serbia, than banks that work with larger companies and rely on investment banking business. Actually, the poor performance of larger banks mainly stems from their huge deposits and operating costs (Kumar & Singh, 2014).

5. Conclusion

The main goal of this paper was to evaluate the efficiency of Serbian banking sector during three years (2017, 2018 and 2019) using output-oriented CCR DEA model. Based on the presented results, it can be concluded that banks in Serbia operated most efficiently in 2018. In the next year the average efficiency decreased by 0.04%, which is a consequence of faster growth of input values compared to the growth of net operating income and a simultaneous decline of net profit. Although the decrease in efficiency in 2019 is a negative indicator of the movement of this score, the encouraging fact is that this score did not fall below the value achieved in 2017.

Banks that operated efficiently in Serbia during observed period of time are precisely in the category of banks with low or medium levels of input values. When it comes to output values, labor costs and depreciation in many cases distinguishes efficient from inefficient banks, which indicates the need to modernize and automate business processes. In fact, bank with a higher performance will be more secure (Chiu et al, 2009).

The presented methodology results explain the positioning of banks in Serbia in the pre-pandemic period. Undoubtedly, to better understand the long-term potential of the Serbian banking sector, the post-pandemic period should be taken into consideration. Also, according to Knežević et al. (2012), it would be of great importance to monitor the efficiency of banks in relation to banks from other countries in order to determine the position of the Serbian banking sector in the region.

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