"Financial institutions performance evaluation in a unique developing market using TOPSIS approach"

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NUMBER OF REFERENCES	NUMBER OF FIGURES	NUMBER OF TABLES
19	1	4

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Abstract

Using Technique for Order Performance by Similarity to Ideal Solutions (TOPSIS) approach for the data from 2011 to 2015, the authors investigate the financial performance of 16 different financial institutions in Oman that include nine commercial banks, three specialized banks, two investment companies, and two finance companies. They find that the one investment company, Dhofar International Development and Investment Holding Co., was more efficient in 2015 and 2011. Moreover, Oman Housing Bank was more efficient in 2013 and 2014, while Ahli Bank was more efficient in the year 2012. In contrast, Bank Muscat that has the largest total assets was ranked number 16 for the years 2013, 2014 and 2015. As a result of Spearman's Rho (Rank-Order) Correlation, all ranked results are related to other years. If a bank is at placement in level, it can be affected by year before or year after. But Oman banks' correlations shows that there are 2 different periods as effecting one year to the other.

Keywords: financial institutions performance, TOPSIS, emerging markets, efficiency, decision making criteria.
JEL Classification: G21, G23, L25.

Introduction

The performance evaluation of financial institutions has very important results for investors, creditors and stakeholders. Measuring the banks', investment and finance companies ability to compete is critical for future sustainability. Their large economic significance highlights the importance of evaluating the financial performance in order to help to improve and monitor their financial conditions. In this paper, we investigate 16 financial institutions: 1) Ahli Bank, 2) Alizz Islamic Bank, 3) Bank Dhofar, 4) Bank Muscat, 5) Bank Nizwa, 6) Bank Sohar, 7) National Bank of Oman, 8) Oman Arab Bank, 9) HSBC Bank Oman, 10) Central Bank of Oman, 11) Oman Development Bank, 12) Oman Housing Bank, 13) Dhofar International Development & Investment Holding Company, 14) Oman International Development and Investment Company, 15) Muscat Finance Company, 16) United Finance Company. The selected banks represent all national banks in Oman, Central Bank of Oman and two specialized banks. In addition, other investment and finance companies are selected based on their assets size and availability of information. The study is conducted using Technique for Order Performance by Similarity to Ideal Solutions (TOPSIS) approach for a recent period 2011-2015. This is a critical investigation for several reasons. Firstly, the changes in the Omani economy from fuel dependence economy into a more diversified economy. This strategic change means that private sector and banking will play a measure role in the coming few years. This is known as the triple PPP, which stands for Public and Private Partnership. Secondly, the testing of financial institutions can add value to strategic policy makers in Central Bank of Oman, investors and commercial banks executive management. The period between 2011 and 2015 is characterized by a period of both strong economic growth, investment and a decline in oil prices, which effected both government and private sector profitability.

This paper makes several contributions. This is the first study that uses TOPSIS approach to measure the financial performance of financial institutions in Oman. Also, this paper examines a critical very recent time frame that combines both increase and decrease of economic activities. Furthermore, this is the first study that combines not only the commercial banks, but also specialized banks, investment companies and finance companies.

This paper is organized as follows. Section 1 provides the literature review. Section 2 describes the data and methodology used to assess the bank branches effectiveness, followed by section 3, which summarizes the results and implications. Major conclusions are illustrated in final section.

1. Literature review

Over the last few decades, there is a growing number of literature on the efficiency and effectiveness of financial institutions. These large numbers of studies use different methods to measure banks' and other financial institutions performance. Both parametric and nonparametric techniques have been used as efficiency measures. Murillo-Zamorano (2004) addressed that both techniques have attracted debate and no method is strictly preferable. Some studies use traditional ratio analysis to measure banks' performance, but conclude that it could be not sufficient (Tozum, 2002). Traditional ratio analysis measures the relationship between two variables to provide some indication about the

performance of financial institutions. Such ratio analysis includes profitability, assets quality, capital adequacy, risk management and many others. Traditional ratio analysis is easy to understand and simple, however, it only examines part of the units' activity without a comprehensive measure of performance evaluation. Furthermore, regression analysis is another common method using input-output analysis to measure financial performance of financial institutions. This method can help to evaluate the influence of multiple independent variables on dependent variable. Other methods include: frontier efficiency, multivariate statistical analysis, balanced scorecard, analytic hierarchy process and gray relation analysis.

Li, Liu, Liu and Whitman (2001) employed nine different financial rates to measure the performance of Chinese banks. They identified two reasons for low profitability of state-owned commercial banks compared with joint-equity banks. First is the higher ratio for non-interest expenses and second is the lower interest margin. Mercan, Reisman, Yolalan and Emel (2003) documented the effect of financial crisis and the impact of financial liberalization on commercial banks in Turkey using a financial performance index. They used fundamental financial ratios from 1989 to 1999. Staub, Souza and Tabak (2009) studied the efficiency of Brazilian banks during the period 2000 to 2007 using Data Envelopment Analysis (DEA). They conclude that banks in Brazil have low levels of economic efficiency compared European and American banks. In addition, they find that state owned banks are more cost efficient compared to foreign and private banks in Brazil. Using similar approach, Data Envelopment Analysis, many studies have found that banking sector is inefficient in different countries (Saha and Ravisankar, 2000; Bhattacharyya, Lovell and Sahay, 1997). Furthermore, Tecles and Tabak (2010) suggested that large banks in Brazil are the most profitable and cost efficient. Both local and foreign banks achieved a good level of performance and improved their efficiency. The study used Bayesian stochastic frontier approach for the period between 2000 and 2007.

Another study by Lin and Zhang (2009) found that bank performance measure by profitability, efficiency, and assets quality is better for Chinese banks with foreign acquisition and public listing compared to the big four state owned commercial banks. Ravi, Kurniawan, Thai and Kumar (2008) measured the bank financial performance using soft computing prediction system. The model combined includes multi-layered feed forward neural network trained with back propagation and other statistical techniques. Another method to measure the bank performance is the CAMELS evaluation system (Capital adequacy, Assets, Management capabilities, Earnings, Liquidity and Sensitivity) used by Kaya (2001). She conducted her study on

Turkish banking sector between 1998 and 2000. The study concludes that Turkish banks were successful using a set of financial success measurements.

Kraft and Tırtıroğlu (1998) investigated the banking system in Croatia and found that new private banks are not as efficient as the old ones. Hence, they recommended that the government should create a competitive framework to govern the banking system. Mertens and Urga (2001) found that big banks are more efficient in terms of profit, but less efficient than small banks in terms of cost. The study conducted in analyzing the efficiency of commercial banks in Ukraine. In Hungary, Hasan and Marton (2003) found that domestic banks in developed countries experienced higher efficiency than foreign banks.

2. Data and methodology

This paper investigates the Oman's financial institutions between 2011 and 2015. The analysis includes nine local banks: Ahli Bank, Alizz Islamic Bank, Bank Dhofar, Bank Muscat, Bank Nizwa, Bank Sohar, National Bank of Oman, Oman Arab Bank, HSBC Bank Oman, three specialized banks: Central Bank of Oman, Oman Development Bank, Oman Housing Bank, two investment companies: Dhofar International Development and Investment Holding Company, Oman International Development and Investment Company, and two finance companies; Muscat Finance Company and United Finance Company. Table 1 lists these financial institutions based on their total assets in US dollars for the year 2015.

Table 1. Total assets of financial institutions in Oman

No.	Financial institutions	Total assets ($000)
1	Bank Muscat	32,625,488
2	Central Bank of Oman	19,736,801
3	Bank Dhofar	9,344,864
4	National Bank of Oman	8,487,386
5	Oman International Development and Investment Co.	5,848,635
6	Bank Sohar	5,741,482
7	HSBC Bank Oman	5,721,456
8	Oman Arab Bank	5,156,567
9	Ahli Bank	4,937,061
10	Oman Housing Bank	903,771
11	Bank Nizwa	900,115
12	Alizz Islamic Bank	657,291
13	Dhofar International Development and Investment Holding Co.	647,594
14	Oman Development Bank	446,294
15	Muscat Finance Company Limited	411,443
16	United Finance Company	320,170

We used financial and non-financial variables related to financial institutions for TOPSIS method. All variables and information are collected and obtained from Bankscope and financial

institutions financial statements, such as their balance sheets, income statements and their official web-sites.

Table 2. All variables for TOPSIS analysis

Variable	Using in TOPSIS
Deposits & short term funding	Thousand USD
Total assets	Thousand USD
Return on average assets (ROAA)	Calculating in %
Return on average equity (ROAE)	Calculating in %
Loan loss res / gross loans	Calculating in %
Cost to income ratio	Calculating in %
Number of employees	Sum of employees in the year

In this study, we have used Technique for Order Preference by Similarity to Ideal Solution (TOPSIS). It is called multicriteria decision analysis. It is one of the classical decision making methods. The method was, first, developed by Hwang and Yoon in 1981 and, then, modified by Yoon in 1987. Finally, Lai and Liu has finalized the TOPSIS method in 1993. This method can be described with concept of chosen alternatives. The alternatives are either having a short distance or a long distance from the decision point. These points may include positive or negative ideal solutions from the decision point. The method can calculate the weighted alternatives to determine the standard decision matrix and it includes 6 steps. The following part illustrates these steps in details.

STEP 1: Determination decision matrix

In first step, it is necessary to make decision matrix. In terms of matrix, it has rows that include decision point, which want to be ordered by success criteria and columns refer to evaluation factors that may be used in decision making process.

$$A_{ij} = \begin{bmatrix} a_{11} & a_{12} & ... & a_{1n} \\ a_{21} & a_{22} & ... & a_{2n} \\ . & & & . \\ . & & & . \\ . & & & . \\ a_{m1} & a_{m2} & ... & a_{mn} \end{bmatrix} .$$

In first step, A_{ij} matrix *m* refers to count of decision point, *n* refers to count of evaluating factor as well.

STEP 2: Standard decision matrix

In the Step 2, standard decision matrix can be calculated with using first step's A_{ij} matrix's components. It can be shown as follows:

$$r_{ij} = \frac{a_{ij}}{\sqrt{\sum_{k=1}^{m} a_{kj}^2}} .$$

Once r_{ij} matrix is calculated, we make a new matrix that is vector normalization with the new elements like below:

$$r_{ij} = \begin{bmatrix} r_{11} & r_{12} & ... & r_{1n} \\ r_{21} & r_{22} & ... & r_{2n} \\ . & & & . \\ . & & & . \\ . & & & . \\ r_{m1} & r_{m2} & ... & r_{mn} \end{bmatrix} .$$

STEP 3: Weighted standard decision matrix

In this step, we need to know all element's weight in the matrix. All weighted elements' sum must be 100 percent. Indeed, firstly, we can calculate criteria weights (w_i) related to evaluation factors. And, then, each element that is placed in each column must multiply with the w_i, therefore, we can make a new matrix. New modified matrix is called V Matrix as follows:

$$\sum_{i=1}^{n} w_i = 1 \qquad ,$$

$$V_{ij} = \begin{bmatrix} w_1 r_{11} & w_2 r_{12} & ... & w_n r_{1n} \\ w_1 r_{21} & w_2 r_{22} & ... & w_n r_{2n} \\ . & & & . \\ . & & & . \\ . & & & . \\ w_1 r_{m1} & w_2 r_{m2} & ... & w_n r_{mn} \end{bmatrix} .$$

STEP 4: Making solutions for ideal (A^+) and ideal (A^-)

In the step 4, from the *V* matrix, two different solutions are produced which set both positive ideal solution and negative ideal solution. It is related to weighted normal value largest (best) choice which can alternatively be at least (worst) value acceptable to determine alternative. Set of ideal solving can be calculated using maximum and minimum component to find best solution as follows:

$$A^+ = \left\{ (\max_i v_{ij} \mid j \in J), (\min_i v_{ij} \mid j \in J') \right\} .$$

The formula above is determined for A^+ ideal solution cloud. For A^-, we calculated ideal solution as follows:

$$A^- = \left\{ (\min_i v_{ij} \mid j \in J), (\max_i v_{ij} \mid j \in J') \right\} .$$

In both formulas, *J* shows benefit maximization, and *J'* shows lost minimization.

STEP 5: Calculating dimension of distinction

This step needs to calculate standard deviation of decision point among positive and negative dimen-

sions. There are two parts of solution for each positive and negative side. It can be calculated as follows using Euclidean distance formula:

$$S_i^+ = \sqrt{\sum_{j=1}^{n}(v_{ij} - v_j^*)^2} \quad ,$$

$$S_i^- = \sqrt{\sum_{j=1}^{n}(v_{ij} - v_j^-)^2} \quad .$$

S_i^+ represents distance from ideal positive solution and S_i^- represents distance from ideal negative solution too.

STEP 6: Calculating ideal solutions of relative proximity

In this last step, relative proximity can be calculated using ideal negative and ideal distinction as follows:

$$C_i^* = \frac{S_i^-}{S_i^- + S_i^*} \quad .$$

C_i^* value must be between $0 \leq C_i^* \leq 1$, and $C_i^* = 1$, and

$$C_i^* = 0 .$$

When an alternative is close to positive ideal solution, value of C_i^* is close to 1, otherwise, the alternative is close to negative ideal solution and its value approaches to 0. This situation of each alternative is represented in Figure 1.
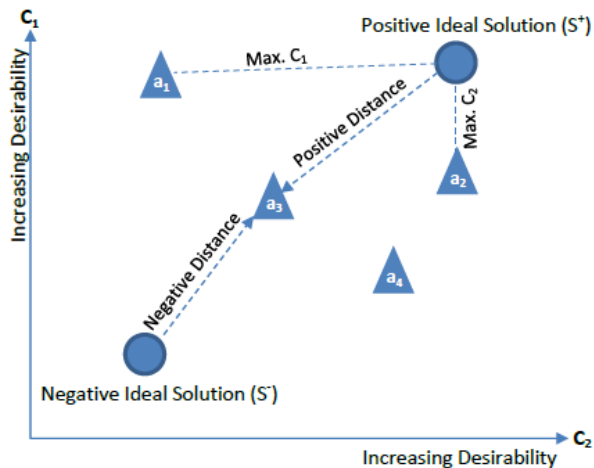

Fig. 1. Illustration of TOPSIS

When we calculate S_i^+, S_i^- and C_i^*, then, we need to rank all results according to the descending order of the value. If any C_i^* of each alternative has distance from 1, it can be understood that it is placed too close to positive ideal point. And it can be said that it has higher performance than the others.

After ranking order by result in terms of descending order, we may calculate coefficient of correlation between two ranked variables. Spearman's Rho

(Rank-Order) Correlation is the nonparametric version of the Pearson product-moment correlation and it is used in strength of all series.

$$r_s = 1 - \frac{6 \sum D^2}{N(N^2 - 1)} .$$

In this formula, D^2 refers to square of two variables that are different according to rank order, N refers to count of total elements in the analysis. In the result of the Spearman's Rho Correlation, we can test the relationships between ranked results.

3. Results and implications

According to TOPSIS method, we analyzed performance measurements of Oman's financial institutions given a set of variables for the period from 2011 to 2015.

Table 3. All ranked results for financial institutions from 2011 to 2015

Financial Institutions	2011	2012	2013	2014	2015
Dhofar Int. Development & Inv. Holding Co.	1	8	4	2	1
Oman Housing Bank (S.A.O.C.)	2	2	1	1	2
United Finance Company	9	4	3	5	3
Ahli Bank SAOG	3	1	2	3	4
Muscat Finance Company Limited SAOG	4	3	7	4	5
Oman Arab Bank SAOC	5	5	5	7	6
Bank Sohar SAOG	8	10	8	8	7
Oman International Development and Inv. Co.	6	7	6	6	8
Oman Development Bank SAOC	7	6	9	9	9
HSBC Bank Oman	10	11	10	11	10
Bank Nizwa SAOG	NA	NA	14	10	11
National Bank of Oman (SAOG)	12	12	12	13	12
Bank Dhofar SAOG	11	9	11	12	13
Alizz Islamic Bank S.A.O.G	NA	NA	13	15	14
Central Bank of Oman	13	13	15	14	15
Bank Muscat SAOG	14	14	16	16	16

Table 3 represents all results of TOPSIS method for each year. The order is based on the year 2015. Dhofar International Development and Investment Holding Co., according to all variables and the results of TOPSIS method, is placed at the first rank in 2015 and 2011. However, in 2014, the company is ranked second after Oman Housing Bank. Moreover, Dhofar International Development and Investment Holding Co. is ranked 4th and 8th for the years 2013 and 2012, respectively.

Hence, it can be concluded that Dhofar International Development and Investment Holding Co. was more efficient in terms of given variables using the TOPSIS method in 2015 and 2011. In addition, Oman Housing Bank was ranked first for two consecutive years, 2013 and 2014. In 2012, Ahli Bank was more efficient, as it was ranked number one using the TOPSIS method. In contrast, Bank Muscat and Central Bank of Oman, which were considered the two

biggest banks in terms of total assets, were placed at the bottom of the list. This indicates that both are less efficient based on the method used in this paper.

As for Spearman's Rho Correlation, we tested relationships between ranked series among all banks and for all years. We used IBM SPSS 21 version to calculate Spearman's Rho Correlation, and the results are shown in Table 4.

Table 4. Spearman's Rho Correlations

			2011	2012	2013	2014	2015
Spearman's Rho	2011	Correlation coefficient	1.000	.867**	.416	.528*	.502*
		Sig. (2-tailed)		.000	.109	.035	.048
		N	16	16	16	16	16
	2012	Correlation coefficient	.867**	1.000	.425	.467	.431
		Sig. (2-tailed)	.000		.101	.069	.095
		N	16	16	16	16	16
	2013	Correlation coefficient	.416	.425	1.000	.932**	.944**
		Sig. (2-tailed)	.109	.101		.000	.000
		N	16	16	16	16	16
	2014	Correlation coefficient	.528*	.467	.932**	1.000	.971**
		Sig. (2-tailed)	.035	.069	.000		.000
		N	16	16	16	16	16
	2015	Correlation coefficient	.502*	.431	.944**	.971**	1.000
		Sig. (2-tailed)	.048	.095	.000	.000	
		N	16	16	16	16	16

**. Correlation is significant at the 0.01 level (2-tailed).
*. Correlation is significant at the 0.05 level (2-tailed).

Conclusion

We analyzed the performance of selected Omani financial institutions between the period of 2011 and 2015. The analyses include nine commercial banks, three specialized banks, two investment companies and two finance companies. All variables and information are collected and obtained from Bankscope and financial institutions financial statements such as their balance sheets, income statements and their official web sites. In this study, we have used Technique for Order Preference by Similarity to Ideal Solution (TOPSIS). It is called multicriteria decision analysis. In the TOPSIS method, we used financial and non-financial variables related to financial institutions using TOPSIS method. As a result of our analysis, it can be concluded that Dhofar International Development and Investment Holding Co. was more efficient in terms of given variables in the TOPSIS method in 2015 and other years, respectively, 2nd, 4th, etc. Moreover, Oman Housing Bank has 2nd rank in 2015, 1st in 2014 and the same result for 2013. The first 5 levels of the results show more efficient financial institutions in Oman than others.

After ranking order by result in terms of descending order with the TOPSIS method. We calculated coefficient of Spearman's Rho (Rank-Order) Correlation between two ranked variables. Moreover, Spearman's Rho Correlation analysis shows strong relationships between years, especially 2011 and 2012, where it look as one group. In the second group, those can be determined relationships from 2013 to 2015. In the group, there are strong relationships between variables in each year. All strong relations are significant at the 0.01 level as a statistically base. As a result of this correlation, all ranked results are related to other years. If a financial institution is at placement in level, it can be affected by the year before or the year after. However, Oman's financial institutions correlations show that there are 2 diffe-rent periods as effecting one year to the other. A further research study might compare the results using different approaches to measure the performance other than the TOPSIS approach.

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