

Performance Evaluation of Iranian Private Banks Using Data Envelopment Analysis (DEA)

Kasra Sajadi

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Approval of the Institute of Graduate Studies and Research

Prof. Dr. Ali Hakan Ulusoy
Acting Director

I certify that this thesis satisfies all the requirements as a thesis for the degree of Master of Science in Industrial Engineering.

Assoc. Prof. Dr. Gökhan İzbrak
Chair, Department of Industrial
Engineering

We certify that we have read this thesis and that in our opinion it is fully adequate in scope and quality as a thesis for the degree of Master of Science in Industrial Engineering.

Asst. Prof. Dr. Sahand Daneshvar
Supervisor

Examining Committee

1. Assoc. Prof. Dr. Adham Makkieh

2. Asst. Prof. Dr. Sahand Daneshvar

3. Asst. Prof. Dr. Mohammad Ali Mosaberpanah

ABSTRACT

Assessing performance of any organizations is necessary in guiding their future strategic decisions. Banks also are no exception. The purpose of this study is to evaluate the performance of 15 Iranian private banks using data envelopment analysis in 5 years period using Data Envelopment Analysis (DEA). The result will help these banks to improve their performances by focusing on inadequate factors and evaluate their efficiency among the competitors.

In the process of evaluating the banks, all of the important indicators are verified and inputs and outputs have been selected in order to assess the efficiency of the banks based on previous studies and the views of experts in this field. Then, the efficient and inefficient banks and their operational rank were determined. Also, reasons for inefficiency of banks would have been detected and proposal for improvements would be discussed.

Keywords: Private Banks, Banking efficiency, DEA, Iranian Banks, Privatization

ÖZ

Herhangi bir kurumun yapacağı performans değerlendirmesi o kurumun gelecekteki stratejik kararlarını belirtemede büyük önem taşımaktadır. Bankalar bu çalışmanın amacı 15 özel İran bankasının 5 yıllık bir süre içindeki performansını DEA (Veri Zarflama Analizi) kullanarak değerlendirmek.

Sonuçlar bankaların yetersizliklerine odaklanarak performanslarını geliştirme fırsatı yaratacak. Ayrıca rakipleri ile kıyasla araştırma ve araştırmadaki bankaların ne kadar etkili olduklarını analiz etme fırsatı verecektir.

Bankaların analizi esnasında tüm önemli göstergeler belirlenmiş ve onaylanmıştır, Ayrıca girdi ve çıktılar da önceki araştırmalara ve alanında uzman kişilerin fikirlerine başvurularak gerçekleştirilmiştir. Bunun üzerine yeterli ve yetersiz bankalar belirlenmiş ve operasyonel sıralamaları oluşturulmuştur.

Buna ek olarak bankaların yetersizlik sebepleri saptanmış ve iyileştirici projeler için önerilerde bulunulmuştur.

DEDICATION

TO MY FAMILY

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Chapter 1

INTRODUCTION

Nowadays evaluating efficiency in any organization is a must in order to know how selected strategies are affected and how plans should be chosen to compete with other rivals. Banks, which are playing vital role in any economic system of a country, are not an exception. Therefore, evaluating banks efficiencies and detecting effective factors for them are necessary because of the mentioned reasons. In this chapter, after describing the problem, the problem identification will be discussed. In the last section of this chapter, the aim and purpose of this study will be explained.

1.1 Problem Description

A bank is a financial institution that accepts deposits from the public and creates credit. In any country, banks have the most critical role in its economy. After entering private banks in the financial market, public demands for different types of bank services grow. Therefore, banks try to use any method that can improve their efficiency to attract more customers. For that reason, evaluating the performance of banks is very important to detect the efficient and inefficient factors.

Following the Iranian revolution (1979), the government had no other choice but to make all banks run by the government, because of distrust of people and investors to the banks and financial institutes, the government had to create trust by guaranteeing the obligations of banks.

For this reason, the government-run banking system in Iran made by merging small private banks. However, the lack of competition among banks caused an inefficient banking system and dissatisfaction of customers and the government.

With worldwide changing economy, the growth of world trade, presence of international banks in neighbor countries, spread of internet and traveling out of borders caused banks and people to know more about the performance of the banks in other countries and their services to customers. The significant difference between Iranian banks and banks all over the world was revealed.

When the war between Iran and Iraq finished, Iran got more stability. In 1998, the government decided to begin to privatize bank sector. It was started by offering part of shares to the stock market. The process of privatizing banks done during years. Banks slowly offered their shares in the market.

After 18 years of starting privatization, currently there are 20 private banks, eight government banks, and 5 Interest-free banks. 16 out of 20 private banks founded by the private or public sector and the rest privatized by offering their shares in the stock market. As time passed, the number of private banks increase. The chart below shows the growth of private banks in Iran after starting the privatization program.

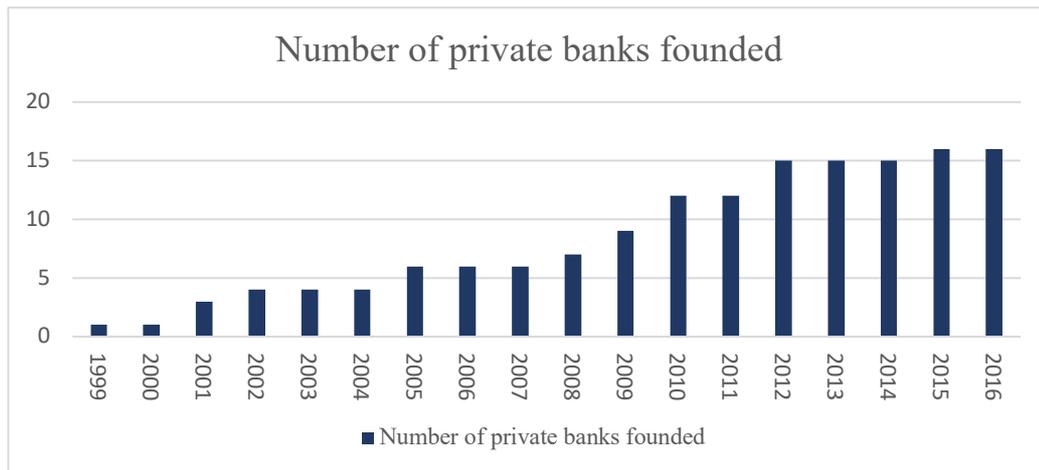


Figure 1.1: Number of Founded Private Banks per Year

One of the reasons that the number of private banks grows is because of different types of sanctions against Iran. After a wave of sanctions in 2006, the first banks that were sanctioned were government banks. In this era, private banks helped a lot in transferring money and trade with other countries.

1.2 Problem Identification

As the number of private banks increases, the competition between them getting harder. Therefore, evaluating the performance of these banks is critical to lead them to a better place.

Nowadays the owners of banks are looking for a trustable method to evaluate the performance of their banks to be sure that their strategies have appropriately executed, to know their place at present and make further plans.

The main problem in any bank or organization with many branches is what method should be used to evaluate correct performance. Owners using the proper evaluating method can put their resources into the essential targets of the organization. Proper performance evaluation helps owners, managers, and stuffs to know the weaknesses of their work and start to adapt necessary actions for improving performance.

In order to compare banks, because of the complexity of the banking system and variation of inputs and outputs, we need a method that can accept more than one input and output.

Data envelopment analysis (DEA) is a linear programming based non-parametric technique for evaluating the relative efficiency of homogeneous Decision-Making Units (DMU) based on multiple inputs and multiple outputs. There exist radial and non-radial models in DEA [1].

In this survey, we are trying to evaluate the efficiency of private banks in Iran. Evaluating the efficiency and assessing the performance will be done by DEA method over a five years period (2011-2015). The banks that reach better amounts of output giving a specific amount of input will be known as efficient banks and all other banks will be compared with each other.

One of the essential roles in any comparison is selecting criteria. many elements involved in the efficiency of any banks. Therefore, inputs and outputs should be selected carefully. In this case, inputs and outputs are selected by studying previous and similar research and using experts' opinions.

In any organization the amount of assets used is an essential criterion as an input. Number of branches and number of workers for each bank is the other inputs that are selected for this study. The main reason for existing any organization is to make a benefit. Therefore, the amount of benefit for each bank is a critical output. The other selected outputs are given facilities and total deposits. We will discuss more in the next chapter.

1.3 Aims and Scopes

In this study, we try to determine efficient and inefficient banks based on inputs and outputs and ranking them by efficiency. Every bank specifically will be studied, and the reasons for being an efficient or inefficient bank will be explained. The studied banks are compared, and analyzed by using the collected data and DEA method. Then, suggestions will be proposed to improve the performance of the inefficient banks. efficiency of programs, processes, and workers, is a sign of a pioneer organization.

In this study, after determining the general field, we started to study about different type of banks in Iran. Based on their importance and speed of growing in the market, we decided to study the private banks. In the next step, determining inputs and outputs based on previous studies and the view of experts was started. After That, the needed data extracted mostly by using annual reports of the banks under study.

In the next stage, data envelopment analysis (DEA) was used to analyze the given data (inputs and outputs) and rank banks by their efficiencies. At the end, we compare and analyze the results and suggest adjustments for improvements.

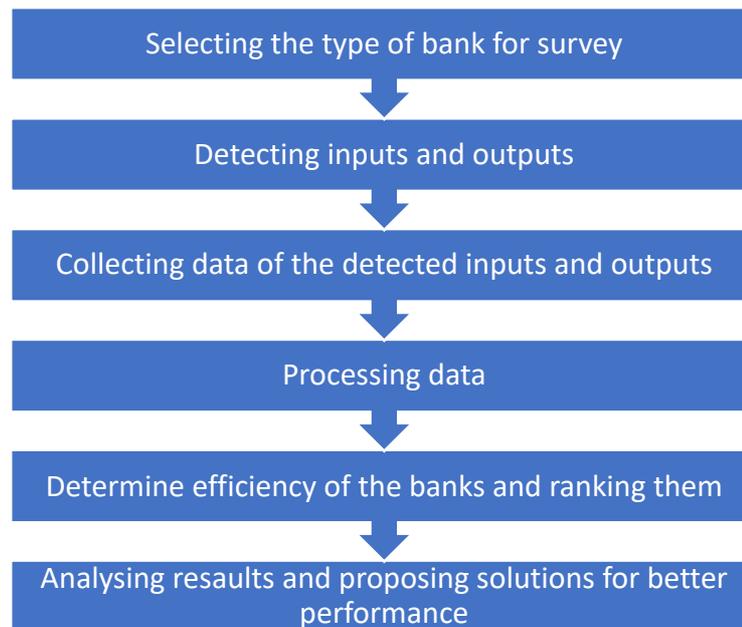


Figure 1.2: Stages of the Thesis

1.4 Structure of the thesis

So far in this report, a general view of the subject and the method that are used to solve this problem were described. In the next chapter, previous related studies will be reviewed. In chapter 3, the method that we used for this thesis is explained. In chapter 4, we will start to analyze and compare the results gained from chapter 3 to find efficient and inefficient banks and rank them. In the last chapter, we propose suggestions to improve the performance of the banks and make a conclusion.

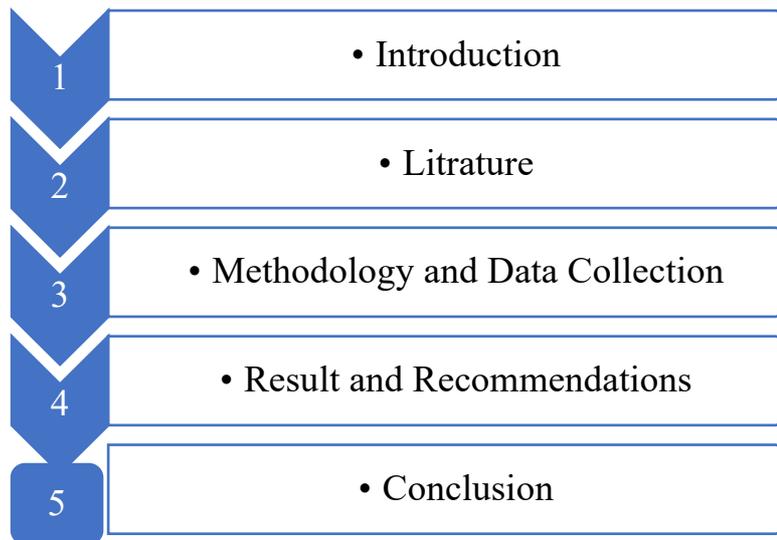


Figure 1.3: Structure of the Thesis

Chapter 2

LITERATURE

Evaluating performance issues have been considered for a long time by researchers. Without study and knowing the amount of progress and getting feedback about how applied strategies were effective, improving efficiency is not possible. All of these actions cannot be done without measuring and evaluating. For this reason, evaluate performance system is a must for any organization.

In the first section of this chapter, we will review different methods that previous researchers used in the banking industry. In the next section, we have a short history of DEA method. In the last section of this chapter, we study different types of DEA method used in previous researches.

2.1 Different types of methods used for evaluating efficiency

2.1.1 Stochastic Frontier Analyses (SFA)

In 2018, an article published by Mokhammad Anwar, about cost efficiency performance of Indonesian banks, after the economic crisis period in this country. The considered period is between 2002 and 2010, and the method has been used is stochastic frontier analysis (SFA) to estimate bank cost efficiency. In this paper, TOBIT regression is also used to reveal the determinants of Indonesian banks' cost efficiency [2].

There is an article written by Thiago Christiano Silva, Benjamin Miranda Tabak published in 2017 that compares two methods of evaluating efficiency. They

compare DEA method with stochastic Frontier model (SFA). The used data was collected from Chinese local banks. The primary purpose of this study is comparing a parametric and a non-parametric method. The result of this article is the similarity between DEA and SFA estimation for efficiency, but the difference in the bank-level estimations [3].

Yizhe Dong and Michael Firth et al., studied the performance of Chinese commercial banks in 2016. The selected period for this survey was between 2002 to 2013, and the used method was single frontier approach (SFA) as developed by Aigner, Lovell, and Schmidt (1977) to estimate cost and profit efficiencies. The result of their study was that the cost and profit efficiencies had improvements over time. Also, these banks are more profit-efficient than cost-efficient. However, foreign banks are the most cost-efficient and the least profit efficient. They conclude that the gap between domestic banks and foreign banks has gotten more after the World Trade Organization transition period (2007–2013) [4].

A study was done by Thi Lam Anh Nguyen in 2018 about Diversification and bank efficiency in six members of the Association of Southeast Asian Nations (ASEAN) countries in 2007-2014. The used method is Stochastic Frontier Approach (SFA). The criteria for this study are asset, funding, and income. The results show that more income-diversified banks have lower cost efficiency while more asset-diversified banks have only lower persistent cost efficiency. More funding-diversified banks caused higher profit efficiency, while more asset-diversified banks caused only higher persistent profit efficiency [5].

In 2017, Oscar Carvalho and Adnan Kasman, used SFA to determine the convergence of cost and profit efficiency in Latin American Banks. They surveyed banks in 19 Latin American and the Caribbean countries over the period 1999–2013[6].

In 2017, an article about comparing banking efficiency between EU members and candidate countries. They compared banks in ten South East European countries and find out how differences in efficiency are related to EU membership. the other issue that they concerned about is how the financial crisis in 2008 affect the banks' efficiency in these ten south east European countries. The used method in this article is SFA, and the conclusions are that the efficiency gap is closing because of adjustments of the less efficient banks. And during the financial crisis, the efficiency of south east European countries banks has been improved [7].

An article was published in April 2019, Mohamed Chaffai and Paolo Coccorese compared MENA (middle east and north African) banks with international banks. Their criteria are cost and revenue efficiency, and they compared banks in 52 countries in 2000-2012 period. They used stochastic meta frontier, made two models and identified two inefficiency components, managerial inefficiency, and technology inefficiency. The results show that banks costs could be reduced by 13%, while revenue could be increased by 17% if their banking systems [8].

2.1.2 Other Methods for Evaluating Efficiency

In February 2019, Christos Alexakis and Marwan Izzeldin worked on productivity in Islamic and conventional banks. They used Malmquist productivity index (MPI) to asses performance in this type of banks in Gulf Cooperation Council (GCC) countries. They mainly focus on the financial crisis period (2006-2012), and their criteria were financial ratios like cost, revenue, and profit. What they concluded is

that Islamic banks are not as good as conventional banks in term of cost performance, but share equal revenue performance with the conventional ones [9].

In March 2019, Xiang Chena and Tsu-Tan Fu studied profit inefficiency and productivity convergence between Taiwanese and Chinese banks. They believed that Malmquist productivity index has a problem because it cannot model the distance function with input contraction and output expansion. Therefore, they developed the difference based Luenberger productivity indicator (LPI) with the directional distance function. The duality between the directional distance function and the profit function provides the LPI to be a useful mean for performance assessment when profitability is the overall goal of firms [10].

In 2016, Abdul Rashid and Sana Jabeen worked on an article about comparing Conventional banks with Islamic banks in Pakistan. In this study, at first, they chose the determinants of performance by constructing the financial performance index (FPI) based on CAMELS' ratios. The used data was collected in 2006-2012 period. By using the GLS regression, the results show that conventional banks are better in operating efficiency, reserves, and overheads than Islamic banks, whereas deposits and market concentration are better in Islamic banks [11].

Sunil K.Mohanty and Hong-Jen Lin et al., surveyed comparing the efficiency of banks in Gulf Cooperation Council (GCC) countries. What they did was comparing cost and profit efficiencies of Islamic banks with that of conventional banks. They used heteroskedastic stochastic frontier (HSF) models to compare these banks in 6 countries membered in GCC. The results show that measures of cost and profit efficiencies of banks vary widely across the six Gulf countries over the same period.

Based on this survey, the essential factors in the efficiency of these banks are country-specific variables [12].

2.2 Different Types of DEA Method

2.2.1 Multi-Stage DEA

A study was done by Xiaoyang Zhou and Zhongwen Xu in June 2019 about evaluating efficiency of banks under uncertainty. They used a multi-period DEA model in three stages: capital organization, capital allocation, and profitability. The selected banks for this study are inefficient in China over 2014 to 2016 period. Study in three stages helps the authors to identify efficiencies and weaker stages. The results show that for all the evaluated banks, the capital allocation stage performed relatively better, while the capital organization and profitability stages were relatively weak across the whole banking system [13].

In 2016, Fadzlan Sufian and Fakarudin Kamarudin researched about determinants of efficiency in the Malaysian banks. They used a two-stage method. In the first stage, they calculated the efficiency of individual banks during 1999-2008 by using bootstrap data envelopment analysis. In the next stage, they used bootstrap regression to find the effect of origins in banks efficiency. The results indicated that the efficiency of these banks improved during the period. Also, banks from Asian countries are more efficient in comparison to foreign banks [14].

There is an article published in 2017 by Hirofumi Fukuyama and Roman Matousek. In this article, they developed a bank network revenue function to evaluate banks' network revenue performance. They researched Japanese banks operating from 2000 to 2013. Two-stage network DEA was used for this survey and what they concluded is that these banks did not reach the optimal levels in their production process and

they suggested expanding the banks' activities in securities and other earning assets [15].

Another article was written in 2014 by Ke Wang and Wei Huang et al., about the efficiency of the Chinese commercial banking system. The used method in this paper is additive two-stage DEA, and they studied 16 major Chinese banks in 2003-2011. In this survey, the authors utilized network DEA to research selected banks. They believed that in order to detect the inefficiency of the banking system, two-stage DEA is more effective than conventional DEA. The results show that the overall efficiency of selected banks to research improved during the study period [16].

2.2.2 Bootstrapped DEA

An article published in 2018 by Filipa Da Silva Fernandes and Charalampos Stasinaks, they surveyed the efficiency of peripheral European domestic banks over 2007-2014. The used method is DEA-Double Bootstrapped Truncated Regression. They utilized data envelopment analysis on a Malmquist Productivity Index to evaluate efficiencies. After that, a Double Bootstrapped Truncated Regression is applied to obtain bias-corrected scores and examine whether changes in the financial conditions affect banks' efficiency levels differently. They concluded that higher levels of liquidity and credit risk exert a negative effect on banks [17].

A research was done in 2019 by Hien Thu Phan, Sajid Anwar at al., about competition and stability in East Asian commercial banks. They studied banks in 4 East Asian countries (China, Hong Kong, Malaysia, and Vietnam) in ten years starting in 2004. They used SFA, DEA, and Sub-sampling bootstrapped DEA. The results show that increased competition may result in a decrease in stability. Also, it

shows bank size, credit risk, and market concentration have a positive effect on stability [18].

In 2016, Chris Stewart and Roman Matousek et al., did a survey on efficiency in the Vietnamese banking system. They considered 48 Vietnamese banks in 1999-2009 period. They applied DEA and bootstrapped DEA in two stages. In the first stage, relative efficiency scores were determined. In the next stage, they bootstrapped the DEA scores with a truncated bootstrapped regression. They concluded that large banks are more efficient than smaller banks. Also, non-state-owned commercial banks are more efficient than state-owned commercial banks [19].

2.2.3 Network DEA

A study assessed management efficiency in Japanese regional banks. This study was done by Satoshi Ohsato and Masako Takahashi in 2015, and they used Network DEA because they believed that because of diversification in management situations in Japanese banks, simple DEA has limitations. Their aim for this study is evaluating the overall and divisional efficiencies of Japanese regional banks [20].

Peter Wanke and Carlos Barros researched Brazilian banks in 2014, using two-stage Network DEA. In the first stage, they measured cost efficiency by considering the number of branches and employees. In the next stage, they assessed productivity efficiency by equity and permanent assets. They compared 40 Brazilian banks in 2012. What they concluded is that Brazilian banks are heterogeneous. This means some of the banks are cost efficiency, and the others are productive efficiency [21].

2.2.4 Other Methods of DEA

In 2015, He-Boong Kwon and Jooh Lee did research on production modeling of large U.S. banks. They used a new method by combining DEA and back propagation

neural network (BPNN). They believed that two-stage DEA model lacks predictive capacity. Therefore, they developed this method to add predictive power [22].

In 2016, Peter Wanke Md. Abul Kalam Azad et al., did research over 114 Islamic banks from 24 countries by using TOPSIS and neural networks in two stages. First, TOPSIS is used in a two-stage approach to assessing the relative efficiency of selected banks. Then, in the second stage, neural networks are combined with TOPSIS results as part of an attempt to produce a model for banking performance with effective predictive ability. The result of this study shows that variables that are related to country and cost structure are very useful in efficiency [23].

In March 2018, Meiqin Wu and Changhong Li et al., surveyed the global productive efficiency of Chinese banks. They used DEA cross-efficiency method to evaluate 16 Chinese commercial banks' performance between 2007-2014. Also, a model named VIKOR aggregates used to measure the productive efficiency of the selected banks. This model helps to compare the solution with the attitude index, which ranks all of the banks. The results indicate that the cross-efficiency interval can provide more information than the traditional DEA model, that the banking system efficiency of China has been improved during the study period [24].

2.3 DEA in Other Researches

Because of the ability of DEA method for accepting multiple input and output, it is a common way to evaluate bank efficiencies all over the world. Table 2.1 shows how authors chose their factors as inputs and outputs. Because of popularity of DEA method, many researchers use this method to evaluate banking system's efficiency.

Table 2.1: Recent researches using DEA method

Authors	Year Published	Country	Number of researched banks	Inputs	Outputs
-Simona Alfiero -Alfredo Esposito -Emmanuel Kwasi Mensah -Mehdi Toloo [25]	2019	Europe	250	-Employees -Assets -Equity -Personnel Expenses	-Deposits Banks -Loans -Net income revenue -Net fees commission
-Ning Zhu -Jens Leth Hougaard -Zhiqian Yu -Bing Wang [26]	2019	China	16	- Operational expenses	-Non-performing loans - Interest income -Non-interest income
-Iago Cotrim Henriques -Vinicius Amorim Sobreiro -Herbert Kimura [27]	2018	Brazil	37	-Fixed assets -Total deposits -Personnel expenses	Total loans
Rachita Gulati Sunil Kumarb [28]	2016	India	76	-Price of physical capital -Price of labor -Price of loanable funds	-Advances -Investments -Non-interest income
-Ovidiu Stocia -Seyed Mehdiian -Alina Sargu [29]	2015	Romania	24	-Deposits -Total cost -Employees -Owned equipment	-Net total revenues -Daily average rate
-Amir Moradi Motlagh -Alperhan Babacan [30]	2015	Australia	8	-Interest expense -Non-interest expense	-Interest income -Non-interest income
Efehan Ulas Burak Keskin [32]	2015	Turkey	47	-No. of branches -No. of employees -Total assets	-Total deposit -net profit -Net interest income

Table 2.1 (continued)

Authors	Year Published	Country	Number of researched banks	Inputs	Outputs
Iveta Řepková [33]	2014	Czech	11	-Labor -Deposits	-Loans -Net interest income
Emília Zimková [34]	2014	Slovakia	16	-Employees -Deposits -Fixed assets	-Earning assets
-Romzie Rosman -Norazlina Abd Wahab -ZairyZainol [35]	2014	Middle Eastern and Asian	79	-Deposits -Fixed assets -Personnel expenses	-Loans -Earning asset
Georgios E. Chortareas Claudia Girardone AlexiaVentouri [36]	2012	Europe	22	-Personnel expenses -Total fixed assets -Deposits	-Total loans -Total other earning assets -Fee-based income
-Hirofumi Fukuyama -Roman Matousek [37]	2011	Turkey	25	-Employees -Capital -Deposits	-Securities -Loans
-Qiang Deng -Wai Peng Wong -Hooy Chee Wooi -Cui Ming Xiong [38]	2011	Malaysia	12	-Branches -Staff -Deposits	-Loans -Profit
-Roberta B.Staub -Geraldoda Silva e Souza -Benjamin M.Tabak [39]	2010	Brazil	127	-Interest expenses -Operational expenses -Personnel expenses	-Investments -Total loans -Deposits
-Rajiv D.Banker -HsihuiChang -Seok-Young Lee [40]	2010	Korea	14	-Yearly deflated interest revenue -Yearly deflated other operating revenue	-Yearly deflated interest expense -Yearly deflated other operating expense

Table 2.1 (continued)

Authors	Year Published	Country	Number of researched banks	Inputs	Outputs
-Tyrone T.Lin -Chia-Chi Lee -Tsui-Fen Chiu [41]	2009	Taiwan	117	-Employees -Interest expense -Deposits -Current deposit	-Loans -Earning -Revenue -Interest revenue
-Göran Bergendahl -Ted Lindblom [42]	2008	Sweden	88	-Credit losses -Personnel expenses -Non-interest expenses	-Loan -Deposit -Other earning assets
Olena Havrylchuk [43]	2006	Poland	31-52	-Deposits -Fixed assets -Labor	-Loans -Treasury bonds -off-balance items
-Xiaogang CHEN -Michael SKULLY -Kym BROWN [44]	2005	China	43	-Interest expenses -Non-interest expenses -Price of deposits -Price of capital	-Loans -Deposits -Non-interest income
-George E.Halkos -Dimitrios S.Salamouris [45]	2004	Greek	20	-Interest expenditures -Total assets -Employees -Operating expenditures	-Interest Income -Net Profit
-Chiang Kao -Shiang-Tai Liu [46]	2004	Taiwan	24	-Total deposits -Interest expenses -Non-interest expenses	-Total Loans -Interest Income -Non-Interest Income
-Leigh Drake -Maximilian J.BHall [47]	2003	Japan	145	-General and administrative expenses -Fixed assets -Retail and wholesale deposits	-Total loans -Liquid assets and other investments -Other income
Milind Sathye [48]	2002	Australia	29	-Labor -Capital -Loanable funds	-Loans -Demand Deposits

Chapter 3

METHODOLOGY

This chapter contains three sections. In the first section, we have a review on DEA method, its advantages and CCR model. Then, the inputs and outputs characteristics for this thesis are explained. In the last section, the collected data in a table form is shown.

3.1 Data Envelopment Analysis

Measuring efficiency had been considered all the time by researchers because of its importance in evaluating the performance of any organization. In 1957, Farrell used a non-parametric method for assessing a small factory. Later on, Charnes et al[49] developed the Farrells' theory and presented a model named Data Envelopment Analysis. In recent years in many countries all over the world in order to evaluate performance and other common activities in different fields, different functions of DEA have been seen.

The reason for the popularity of this method in comparison to other methods is the possibility of studying complicated connections and mostly unknown between multiple inputs and outputs that usually exist in studied issues. DEA gives the possibility to reconsider the problems that have been solved before in other ways. For example, the possibility of assessment using this method to detect inefficient sources in high-profit companies.

DEA is based on a series of improvement using linear programming. In this method efficiency curve is formed by points that are produced by linear programming. To determinate these points constant return to scale or variable return to scale can be used. After some improvements, linear programming determines that if selected Decision-Making Unit (DMU) is placed on the efficiency curve or not. In this way, efficient and inefficient DMUs separated.

3.1.1 Advantages and Disadvantages

Some of the advantages of DEA, as listed by Bhat et al. [50], are as follows:

- It can handle multiple inputs and multiple outputs
- It is unnecessary to have an assumption of a functional form relating inputs to outputs
- DMUs are directly contrasted against a contemporary or combination of contemporaries
- Inputs and outputs do not need to have similar units

Meanwhile, once again according to Bhat et al[50]., the disadvantages are:

- Results are sample specific
- Since it is an extreme point technique, measurement error can cause noteworthy problems
- It can tell you how well you are doing compared to your peers but not compared to a theoretical maximum
- It is a non-parametric technique, thus, statistical hypothesis tests are difficult
- Large problems can be computationally intensive because a standard formulation creates a separate LP for each DMU

3.1.2 CCR Model

CCR model is the basic DEA model as introduced by Charnes et al. (1978)[49]. This model was modified by Banker et al. (1984) and became the BCC model which accommodates variable returns to scale. The CCR model presupposes that there is no significant relationship between the scale of operations and efficiency by assuming constant returns to scale (CRS) and it delivers the overall technical efficiency. The CRS assumption is only justifiable when all DMUs are operating at an optimal scale. However, firms or DMUs in practice might face either economies or diseconomies to scale. Thus, if one makes the CRS assumption when not all DMUs are operating at the optimal scale, the computed measures of technical efficiency will be contaminated with scale efficiencies.

Every multiple for inputs, produce the same multiple for outputs. If input X produces output Y , then Input λX Produces λY for $\lambda > 0$. In the other words, if point (X, Y) is feasible, then $(\lambda X, \lambda Y)$ is feasible. If input X produce output Y , then input λX produce output λY for $\lambda > 0$ and efficiency ratio for (X, Y) and $(\lambda X, \lambda Y)$ is identical. Therefore, any changes in inputs results in the same changes for outputs which means increasing in inputs does not cause saving or increasing costs. CCR model uses constant return to scale. For variable return to scale, any multiple for inputs can produce more or less in outputs. When increasing in outputs is more than increasing in inputs, the return to scale is increasing and when increasing in outputs less than increasing in inputs, return to scale is decreasing.

For the CCR model, it is of the assumption that there are n DMUs to be evaluated, of which each DMU consumes varying amounts of m different inputs to produce s different outputs. Particularly, DMU_j consumes x_{ij} amount of input i and produces y_{rj} amount of output r . Furthermore, it is of the assumption that $x_{ij} \geq 0$ and $y_{rj} \geq 0$,

and that each DMU has at least one positive input and one positive output value. In the ratio-form of DEA, the ratio of outputs to inputs is used to measure the relative efficiency of the $DMU_j = DMU_o$ to be evaluated relative to the ratios of all the $j = 1, 2, \dots, n$ DMU_j . The CCR construction can be interpreted as the reduction of the multiple-output/multiple-input situation (for each DMU) to that of a single ‘virtual’ output and ‘virtual’ input. In mathematical programming speech, this ratio, which is to be maximized, forms the objective function for the particular DMU being evaluated, so that symbolically:

$$\max h_o(u, v) = \sum_r u_r y_{ro} / \sum_i v_i x_{io} \quad (3.1)$$

in which the variables are the u_r ’s and the v_i ’s, while the y_{ro} ’s and x_{io} ’s are the observed output and input values of DMU_o , which is the DMU to be evaluated. Without further additional constraints, displayed below, (1.1) is unbounded. A set of normalizing constraints, of which there is one for each DMU, reflects the condition that the virtual output to virtual input ratio of every DMU, including $DMU_j = DMU_o$, should be less than or equal to unity. As such, the mathematical programming problem may be stated as:

$$\max h_o(u, v) = \sum_r u_r y_{ro} / \sum_i v_i x_{io} \quad (3.2)$$

subject to

$$\sum_r u_r y_{rj} / \sum_i v_i x_{ij} \leq 1 \text{ for } j = 1, \dots, n$$

$$u_r, v_i \geq 0 \text{ for all } i \text{ and } r$$

The transformation for linear fractional programming selects the solution (u, v) for which $\sum_{i=1}^m v_i x_{io} = 1$, and produces the equivalent LP problem in which the change of variables from (u, v) to (μ, ν) is a result of the Charnes-Cooper transformation[49]:

$$\begin{aligned}
\max \quad & z = \sum_{r=1}^s \mu_r y_{ro} \\
\text{S.t.} \quad & \sum_{i=1}^m v_i x_{io} = 1 \\
& \sum_{r=1}^s \mu_r y_{rj} - \sum_{i=1}^m v_i x_{ij} \leq 0 \quad j = 1, \dots, n \\
& \mu_r \geq 0 \quad r = 1, \dots, s \\
& v_i \geq 0 \quad i = 1, \dots, m
\end{aligned}$$

$$\max z = \sum_{r=1}^s \mu_r y_{ro}$$

subject to

$$\sum_{r=1}^s \mu_r y_{ro} - \sum_{i=1}^m v_i x_{ij} \leq 0 \quad (3.3)$$

$$\sum_{i=1}^m v_i x_{io} = 1$$

$$\mu_r, v_i \geq 0$$

for which the LP dual problem is:

$$\theta^* = \min \theta$$

subject to

$$\sum_{j=1}^n x_{ij} \lambda_j \leq \theta x_{io} \quad i = 1, 2, \dots, m; \quad (3.4)$$

$$\sum_{j=1}^n y_{rj} \lambda_j \geq y_{ro} \quad r = 1, 2, \dots, s;$$

$$\lambda_j \geq 0 \quad j = 1, 2, \dots, n;$$

3.2 Inputs and Outputs Characteristics

3.2.1 Inputs

3.2.1.1 Number of Employees

Cost of human resources in banks is a considerable part of bank costs. Determination exact amount of these costs in each bank included personnel's salaries is difficult and complicated. For this reason, in this thesis, instead of using cost of employee, number of employees is used. This factor determined as an input.

3.2.1.2 Number of Branches

One of the most important factors in cost and efficiency for banks is the number of branches. Increasing number of branches cause attracting more customers and transactions but on the other hand, add a new branch has a lot of costs for banks. Although nowadays electronic banking has been extended, lots of banking transactions needed to be done in bank branches and existing branches is necessary for banks and customers.

3.2.1.3 Total Assets

Assets, based on economic science, is every properties and rights that have a monetary value. Some of the assets are objectives, like lands, buildings, facilities, equipment, vehicles, assets in a storehouse, or cash. Alternatively, it can be financial rights.

Also, assets are divided into two general groups. Current assets are the items that are expected in a year on any time to turn into cash, consumed or sold. Usually, current assets categorized by the period needed to turn to cash. For example, funds, office equipment, debtors, inventory are current assets. The other type of assets is fixed assets that usually have a long life. Like lands, buildings, facilities, machinery, and equipment, vehicles. For this type of assets always depreciation cost is considered (except land).

3.2.2 Outputs

3.2.2.1 Total Deposit

Deposit is money that is paid to banks by a natural or juridical person under specific circumstances. Banks use this collected money to pay as a loan to intrapreneurs, industries, business owners, and generally any applicant. One of the main goals of

any banks is attracting more deposits. In this thesis, we consider total deposits as an output.

3.2.2.2 Net Annual Profit

Net annual profit is gained profit in an accounting period time, which is calculated by subtracting costs and taxes from total income. Being positive or negative, this amount depends on the performance of the considered system. This factor is an output in this study.

3.2.2.3 Bank Facilities

Banks are specialized in providing financing to their customers, and the various options offered in the market, for example overdraft services, deferred payment plans, lines of credit, revolving credit, term loans, letters of credit, and swingline loans. This factor is considered as an output in this thesis.

3.3 Collected Data

The needed data for inputs and outputs, were extracted mostly from banks annual reports. In table a in appendix 1 (page 63) the data for 15 banks in 5 years have been shown.

Chapter 4

RESULTS AND ANALYSES

After reviewing previous studies in this field and acquainting with expert's opinions, we select inputs and outputs for this study. Then, we collect the needed data to evaluate 15 private bank efficiencies during Iran in a five years period between 2011-2015. All the data extracted from annual reports from each bank. The collected data is shown in appendix 1.

4.1 Bank efficiencies

In this chapter, we start analyzing the collected data using DEA method. The data have been analyzed by PIM-DEA software. The efficiencies of the banks are shown in table 4.1.

Table 4.1: Efficiencies of the researched banks

No	Name	2011	2012	2013	2014	2015
1	Mellat	67.21	73.75	64.22	100	61.18
2	Tejarat	61.24	64.95	58.65	53.4	55.04
3	Saderat	45.4	50.84	45.44	42.93	48.37
4	Parsian	100	100	86.88	84.09	90.88
5	EghtesadNovin	100	99.62	93.39	88.99	82.65
6	Pasargad	91.4	100	92.63	85.34	100
7	Karafarin	78.4	79.13	72.36	68	75.41
8	Sarmaye	58.37	76.91	65.66	52.5	68.46
9	Shahr	80.66	89.51	81.53	76.29	94.03
10	Saman	82.46	90.05	88.16	82.54	86.39
11	Sina	72.88	75.93	65.79	69.02	67.68
12	Ghavamin	100	100	79	77.32	84.27
13	Ansar	74.49	80.93	74.21	65.15	63.64
14	Iran Zamin	77.56	26.94	13.19	9.02	9.05
15	Khavar miane	100	100	63.42	97.45	71.66

In order to rank the banks by their efficiencies, we calculate the average of efficiencies for each bank for 5 years and then we ranked them by average efficiency. Pasargad Bank is the most efficient bank among the researched banks by 93.87% efficiency in average. Eghtesad Novin Bank and Parsian bank are second and third bank in ranking with similar results, 92.93% and 92.37% respectively. The results are shown in figure 4.1:

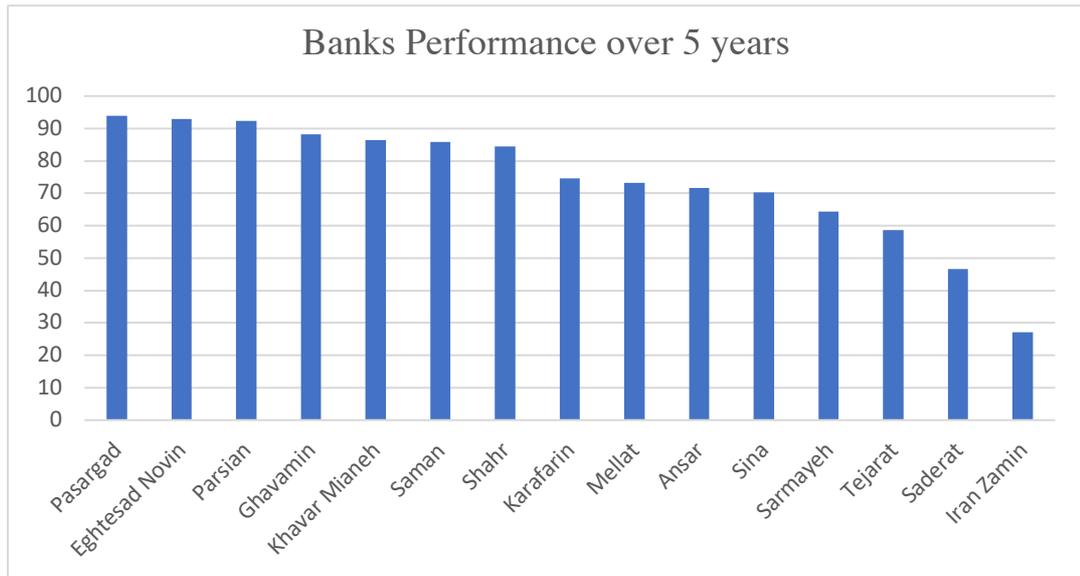


Figure 4.1: Ranking banks by their efficiencies

By using table 4.1 we can find the number of efficient and inefficient banks, and categorize them by the amount of their efficiencies. Most of the banks have more than 90% efficiency, and only 4 out of 75 DMUs have efficiency less than 50%. The results are shown in figure 4.2.

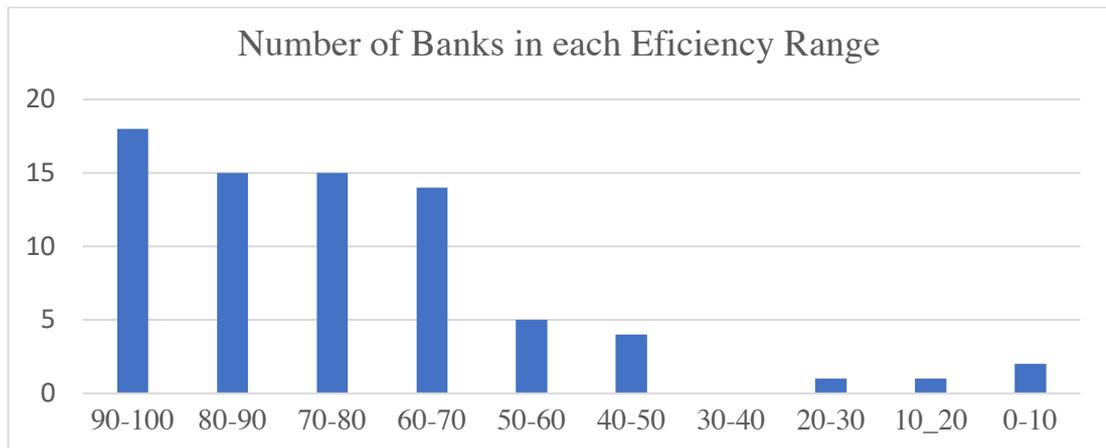


Figure 4.2: Number of Banks Based on Their Efficiencies

The charts below show the efficiency of each bank during the five years. These banks selected as samples. As you can see, most of the banks improved their

efficiencies during the study period, Except for Iran Zamin bank. 10 out of 75 DMUs are 100% which means 10 banks in different years were efficient banks.

- Bank Mellat

Mellat Bank was founded from merging 10 commercial banks after Iran revolution, started privatizing in 2009. This bank is the biggest private bank in Iran based on assets. As it is shown in figure 4.3, two years after starting privatization this bank started to improve its performance. In the fourth year, this bank was an efficient bank but in the last year of the period, efficiency decreased. We will find the reason for this problem in the next chapter.

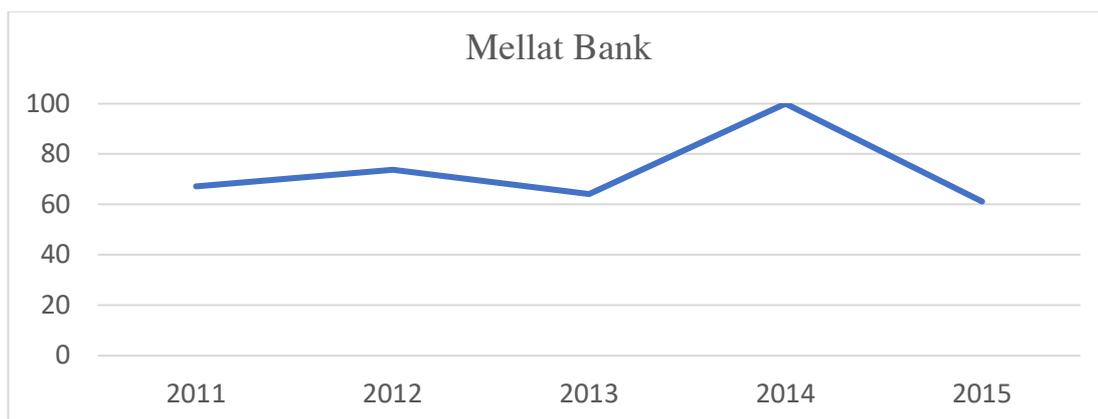


Figure 4.3: Bank Mellat Efficiency

- Tejarat Bank

Tejarat bank is the first Iranian bank founded in 1887. In the beginning, this bank is named Shahi bank and after Iran revolution, it changed to Tejarat bank. This bank started to privatization from 2009. Figure 4.4 demonstrated the performance of this bank during a period, you can see that the efficiency of this bank slightly changed each year, but it still far away from an efficient bank. The average efficiency is 58.65%.

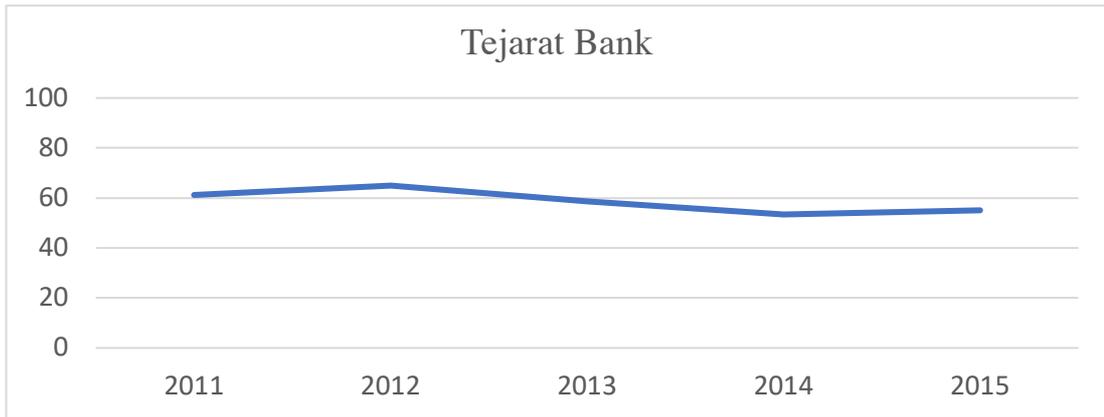


Figure 4.4: Bank Tejarat Efficiency

- Bank Saderat

Bank Saderat is the first in number of branches, founded in 1952. This bank has been privatized starting 2008. in figure 4.5 you can see that efficiency did not change a lot, and it is far away from an efficient bank.

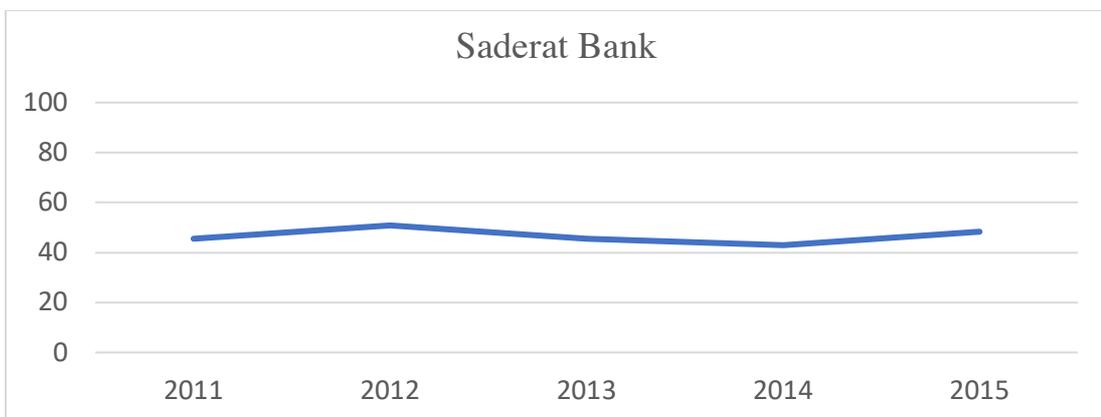


Figure 4.5: Bank Saderat Efficiency

- Parsian Bank

Parsian Bank founded in 2001 and started working 2002, is the second private bank in Iran. This bank is one of the three banks that can turn to an efficient bank in the last year of the period. The efficiency was 100% in 2011 and 2012, but it decreased in 2013. The efficient amount for inputs will be proposed in the next chapter.

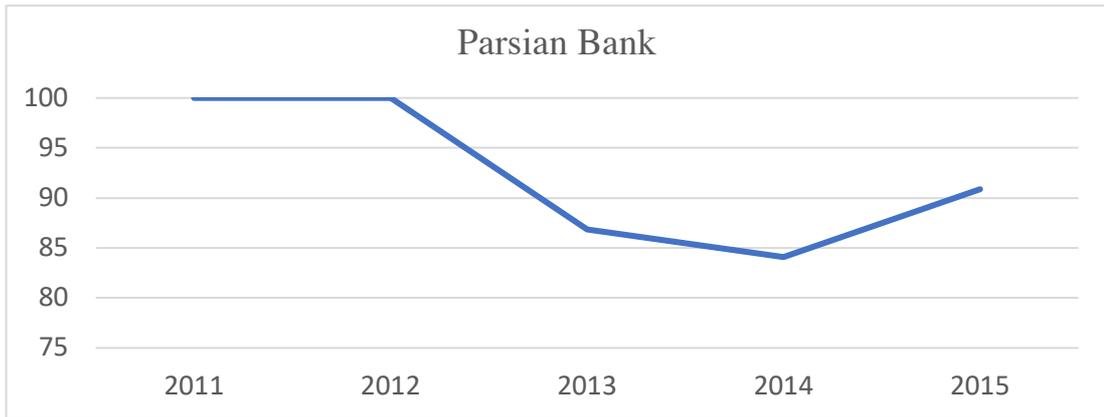


Figure 4.6: Parsian Bank Efficiency

- Bank Iran Zamin

Bank Iran Zamin got the last place in the bank ranking in figure 4.7, with 27.15% average efficiency in the period. This bank founded by giving permission to a financial institute, named Mola Al Movahedin Financial institutions, to promote as a bank in 2011. As you can see in table 1 in appendix 1 (page 63), this bank had minus value in annual net profit in the last three years of the period. This is one of the main reasons for the low efficiency of this bank.



Figure 4.7: Bank Iran Zamin Efficiency

- Bank Ghavamin

Bank Ghavamin founded in 2000 as a financial and credit institution. Ghavamin Bank is the first bank in the country with the aim of helping to enhance the living quality and financial support of working staff of Iran police forces. Figure 4.8 demonstrates that this bank was efficient in 2011 and 2012, but it decreased significantly in 2013.

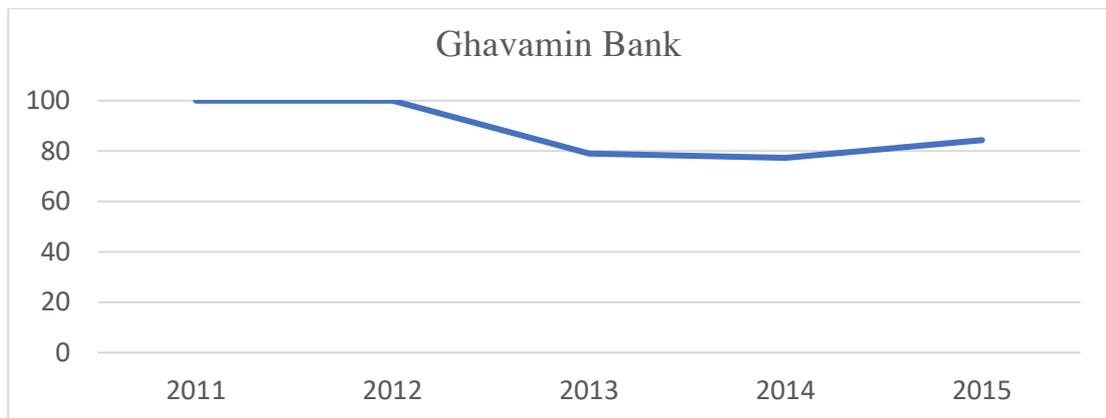


Figure 4.8: Bank Ghavamin Efficiency

- Eghtesad Novin Bank

Eghtesad Novin Bank is the first private bank in Iran Founded in 2001. This bank had an overall improvement in performance and it is very close to being an efficient bank. This banks efficiency slightly decreased during 5 years.

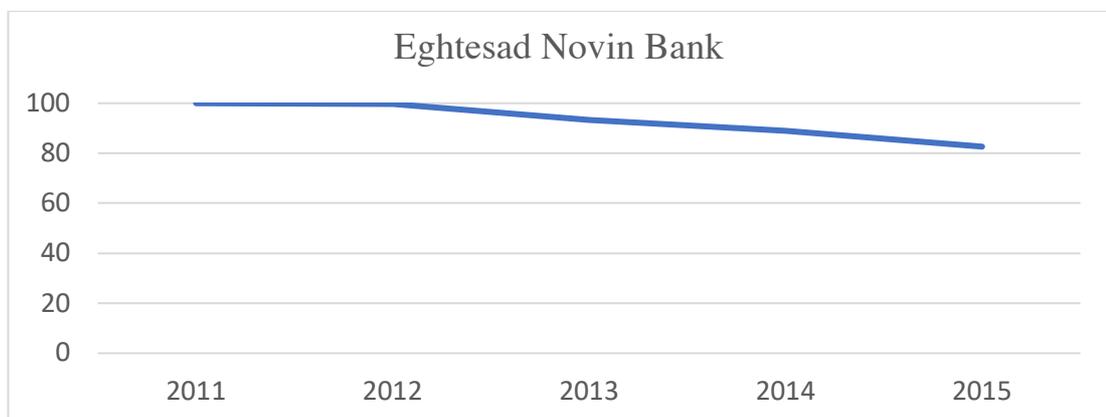


Figure 4.9: Eghtesad Novin Bank Efficiency

- Bank Pasargad

Bank Pasargad unlike the previous banks, started working as a complete private bank in 2006. Figure 4.10 demonstrates this bank was an efficient bank in 2012 and 2015, but efficiency decreased in 2013 and 2014.

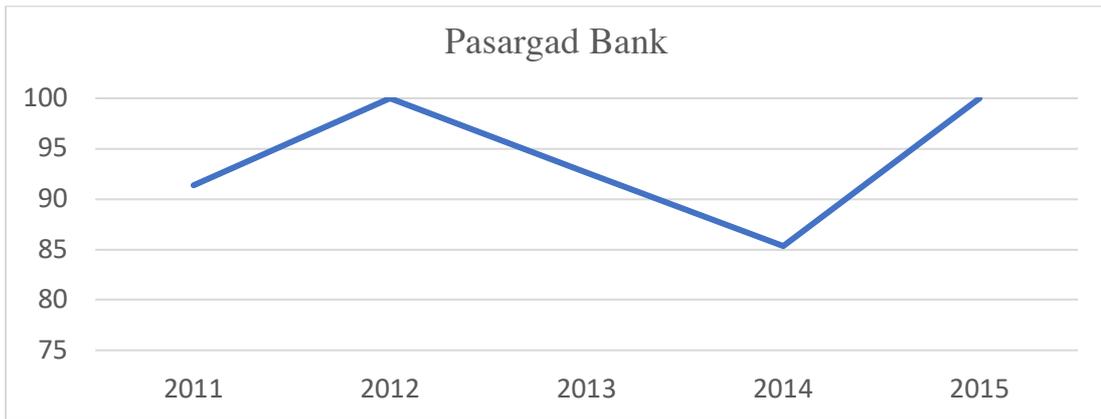


Figure 4.10: Bank Pasargad Efficiency

- Karafarin Bank

Karafarin bank, which is translated to Entrepreneurship Bank established in 2001 is one of the leading banks in this section. As it is shown in figure 4.11, this bank still far away from an efficient bank and during these years it has not been improved.

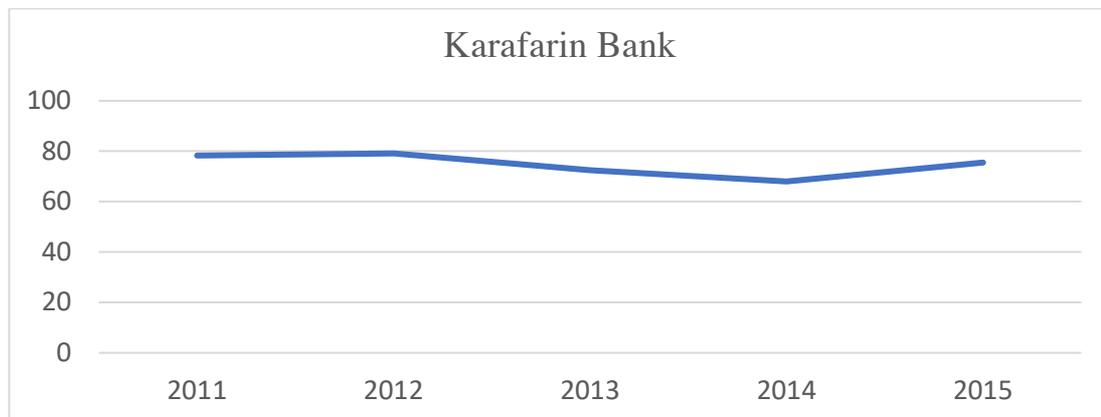


Figure 4.11: Karafarin Bank Efficiency

- Sarmayeh Bank

Sarmayeh Bank is the sixth private bank in Iran founded in 2005. Efficiency in the first year was 58.37%, and it have been changed during the period. Efficient amount of inputs to be an efficient bank will be proposed in the next chapter.

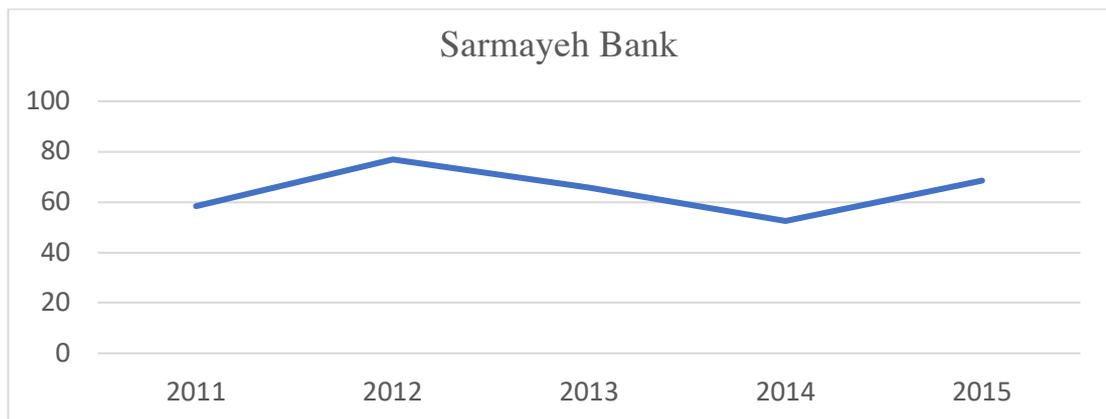


Figure 4.12: Sarmayeh Bank Efficiency

- Shahr Bank

Shahr Bank founded in 2009 by Tehran municipality. Most of the shareholders are the municipality of big cities in Iran. Figure 4.13 demonstrates that this bank had significantly improved during five years. Shahr bank became an efficient bank in the last year of the period. The efficiency in 2011 was 80.66% and they could reach to 94.03% in the last year.



Figure 4.13: Shahr Bank Efficiency

- Saman Bank

This bank started its activities as a Credit Institution in September 1999. Subsequently, on August 2002, it received a full banking license and changed its name to Saman Bank. This bank's efficiency was 82.46% in the first year, and it did not change much during the period, they could manage to keep the efficiency more than 80%. The needed changes will be proposed in the next chapter.

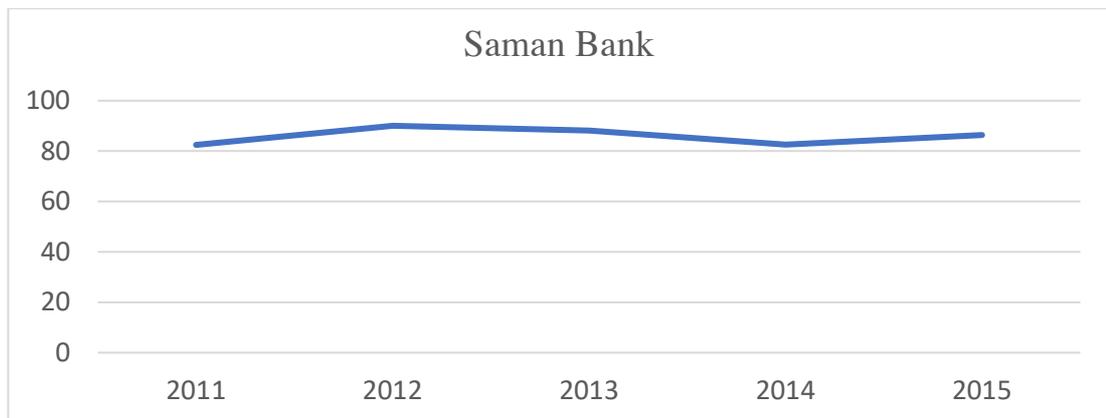


Figure 4.14: Saman Bank Efficiency

- Sina Bank

This bank did not have a lot of changes in its efficiency during the period. all the efficiencies for this bank were between 60% to 80%. Efficient amount of each input will be suggested in the next chapter.

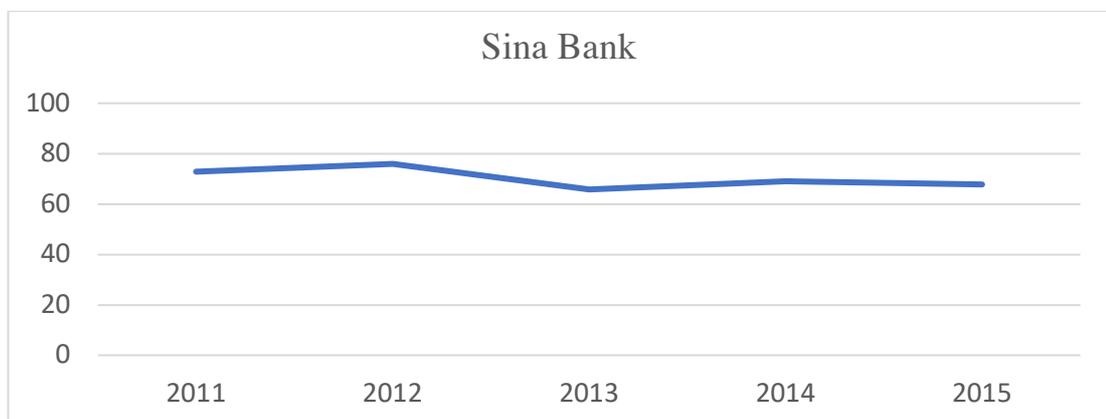


Figure 4.15: Sina Bank Efficiency

- Ansar Bank

Ansar bank had improvements during the study period. This bank was 74.49% efficient in the first year, it improved next year but the efficiency decreased in the last three years.

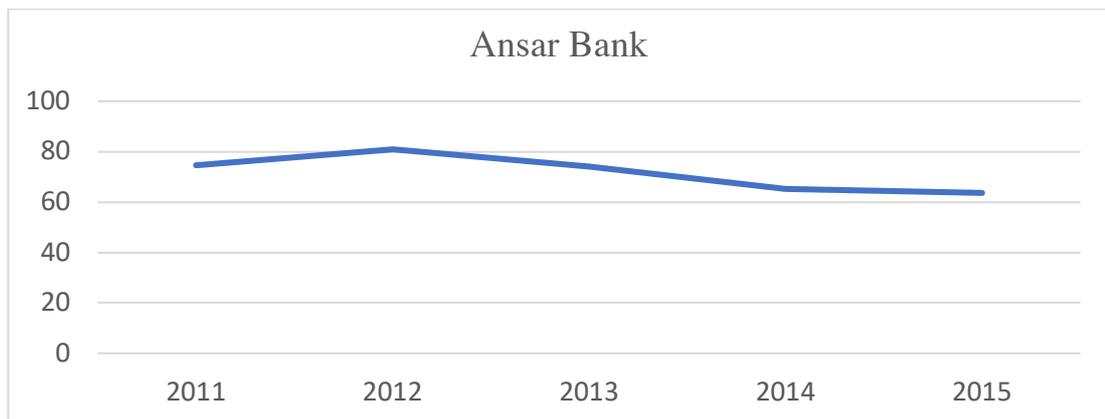


Figure 4.16: Ansar Bank Efficiency

4.2 Detect Significant Input Indexes

After an overall analysis on the results, we try to find out which input (number of employees, number of branches and total assets) has more effect on the efficiency of each bank. There are two ways to solve this problem. The first method is using weights of each input indexes and their averages. The other method is eliminating an input and get the results with the remaining two input. And repeat the test for all three inputs. Then we compare the result to see which input has more effect on the efficiencies of the banks. First, we use weights to detect the most effective input.

4.2.1 Weights Method

in this method, we collect weights of each DMU for a specific input, calculate average and compare the results for all three inputs. Based on table 2 shown in appendix 2, the results are shown in table 4.2.

Table 4.2: Comparing weights of the inputs

Number	Input 1	Input 2	Input 3
Name	No. of Employees	No. of Branches	Total Assets
Total weights	367.93	264.89	950.36
Average	4.90	3.53	12.67

By looking at the results, we can see that the average weights for input 3 is much higher than the others. The averages for input 1 and 2 are nearly close. Which means the effect of input 3 (total asset) is more than input 1 and input 2. To verify the result, we test by another method mentioned before.

4.2.2 Eliminate Inputs Method

In this method, we eliminate one input and get results with the remaining two inputs. We repeat the test three times and each time, we eliminate an input to see the effect of that input on the results. Table 4.3 is a sample of how the calculation is done.

Table 4.3: Sample of calculating for detecting significant indexes

Name	Efficiency	Efficiency without input 1	Difference
DMU01	40.21	42.31	2.1
DMU02	30.71	36.25	5.54
DMU03	21.33	25.31	3.98
DMU04	76.95	77.54	0.59
DMU05	67.05	74.34	7.29

As it is shown in table 4.3, we eliminate input 1 and do the calculations by input 2, input 3 and the same three output. After calculating without input 1, we calculate the difference between the efficiency with all three inputs and the efficiency with just input 2 and input 3 without input 1. Then, we calculate total differences and average difference for all DMUs. As it is shown in table 4.4, when we eliminate input 1 and do the test with two remaining inputs, each DMU has an increase in efficiency 7.73% on average. We recalculate efficiencies for each DMU but this time, we eliminate

input 2 and use input 1 and input 3. This time, the efficiency of each DMU increase by 2.39% averagely.

Table 4.4: Comparing the effect of each input on efficiency

	without Input 1	without Input 2	without input 3
Total Difference	575.55	160.17	1850.88
Average	7.67	2.13	24.67

We do the same calculations without input 3 and with input 1 and input 2. As you can see in table 4.4, the amount of change in each DMU is 22.58% averagely, which is much higher than the results without input 1 and results without input 2. This means that the effect of input 3 on the efficiency is much higher than the two other inputs. In other words, if we increase the amount of input 3, we get more improvement in efficiency than the two other inputs.

This method approved the previous method's result in section 2.1. after detecting the most significant input index, we will find the most significant output index in the next section.

4.3 Detect Significant Output Indexes

It is one of the main factors for any organization to know which one of their outputs is more important and should be more considered because it can affect decisions about plans and strategies. Like section 2, we will detect significant output indexes by two methods. First, we use DMU weights and in the other method, one output will be eliminated and calculations will be done again. First, we start by weights method.

4.3.1 Weights Method

in this method, the weight of each DMU is calculated and it is shown in table 2 in appendix. We calculate total weights for each output, average and then by comparing the average weights for each output, we will find out which output is more important.

Table 4.5: Comparing weights of the outputs

Number	Output 1	Output 2	Output 3
Name	Total Deposits	Annual Net Profit	Given Facilities
Total weights	684.92	128.07	18.18
average	9.13	1.70	0.24

As it is shown in table 4.5, output 1 has more weights on average than the other two outputs. In other words, Total deposit should be more considered by banks because by increasing output 1, efficiency will be more improved more than by increasing output 2 or output 3.

4.3.2 Eliminate Output Method

In order to find the most significant output indexes, by using this method, we eliminate one of the outputs, recalculate the process by all three inputs and the remaining two outputs and then, comparing the created difference in all DMU efficiencies.

First, we eliminate output 1 and do the calculations to find efficiencies by all three input, and output 2 and output 3. As it is demonstrated in table 4.6, when we eliminate output 1 and calculate the efficiency of each DMUs using three inputs and the remaining two outputs, each DMU changes 41.51% averagely. Now, we do the same process with output 2 and output 3. We eliminate output 2 and using output 1 and output 3 and with all three inputs, the efficiency of each DMU changes 2.03% averagely. Similarly, by eliminating output 3 and using output 1 and output 2 the efficiency of each DMU changes by 0.28% on the average.

Table 4.6: Comparing the effect of each output on efficiency

	Without output 1	Without output 2	Without output 3
Total Difference	3113.93	152.74	21.25
Average	41.51	2.03	0.28

The results show that changes in output 1 which is total deposit, has more effect on the efficiency than the other two outputs. In other words, changes in output 1 cause more change in efficiency. The result of this method approved the previous method's result.

4.4 Comparison Criteria

After evaluating efficiency of each bank and finding significant input and output indexes, now we need to find the reasons for inefficient banks. For this reason, we need to know how we should compare banks with each other.

Efficient banks can be known as a benchmark for inefficient banks. Each efficient bank has a weight and we name it λ . As you can see in chapter 3, the amount of λ is calculated by solving the dual model of linear programming DEA for efficient banks. These weights show us how efficient performance of a bank should be, and efficient amount of inputs and outputs can be calculated as a linear combination. Therefore, by using this Knowledge we can present a better evaluation of inefficient banks.

For example, based on table 4.7, DMU068 which is an inefficient bank should be compared to DMU5, DMU12, and DMU15 which have the following weights 0.13, 0.09 and 0.14 respectively. When we have more than one bank as a benchmark, the benchmark with more weight should be chosen.

Table 4.7: Benchmark banks and their weights for inefficient banks

Name	DMU4	DMU5	DMU12	DMU15	DMU19	DMU21	DMU27	DMU30	DMU45
DMU01	0.19	2.51	0	0	0	0	0	0	0
DMU18	0	1.12	0	0	0	0	0.39	0	0
DMU32	0	0.94	0	0	0	0	0.47	0	0
DMU54	0	0.08	0.3	0.02	0	0	0	0	0
DMU61	0.45	1.34	0	0	0	0	0	0	0
DMU68	0	0.13	0.09	0.14	0	0	0	0	0
DMU73	0	0	0	0.15	0	0	0.2	0	0

Chapter 5

CONCLUSION AND SUGGESTIONS

In this chapter, first each inefficient bank is compared with the benchmark bank with the greatest λ . After that, the main reasons for inefficiency in Iranian Private banks will be discussed. At the end, we have some suggestions for further studies.

After calculating efficiencies for each bank and determination of efficient and inefficient banks, now we must find solutions to turn inefficient banks into efficient banks.

5.1 Suggest Improvements and Comparison

When a bank is detected as an inefficient bank, that means this bank uses unbalanced sources (inputs). in the other word, by adjusting inputs in the right way we can change efficiency and make it more efficient.

Assume that (X_0, Y_0) is an inefficient DMU by θ efficiency then for $(\theta X_0, Y_0)$, That DMU will be efficient. In other words, if we multiply inputs times θ , then the efficiency will be 1. The amount of efficiency (θ) is always between 0 to 1. Which means if we multiply inputs to efficiency value, it is always less than the initial inputs ($\theta X_0 < X_0$). This means to turn an inefficient bank into an efficient bank, we have to decrease inputs as we keep the outputs the same.

For example, DMU61, which is bank Mellat in 2015, was 61.18% efficient. The inputs for this DMU were 21342 employees, 1590 Branches and 60,950,842,453 U.S

Dollar total assets. If we use just 61.18% of each input, the efficiency will be 100% as table 5.1 shows.

Table 5.1 Suggestions for Mellat Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Mellat	DMU61	21,342	1,590	60,950,842,453	61.18%
Suggested changes		$0.6118 \times (21,342)$ =13,053	$0.6118 \times (1,590)$ =972	$0.6118 \times (60,950,842,453)$ =37,277,535,244	100%

If an inefficient DMU needs a benchmark to know how it should be to be an efficient bank, we have to find a benchmark so that DMU can compare itself with it. To do this, we use λ to find the best benchmark. An inefficient bank can compare itself with any benchmark with λ more than 0. But DMUs with greater λ are better benchmarks for that inefficient DMU.

Table 5.2 λ values for each bank in last year

Name	DMU4	DMU5	DMU12	DMU15	DMU19	DMU21	DMU27	DMU30	DMU45	DMU46	DMU66
DMU61	0.45	1.34	0	0	0	0	0	0	0	0.33	0
DMU62	0	0.61	0	0	0	0	0.37	0	0	0.32	0
DMU63	0	0.77	0	0	0	0	0.32	0	0	0.52	0
DMU64	0.64	0.08	0	0	0	0	0	0	0	0.04	0
DMU65	0.09	0.56	0	0	0	0	0	0	0	0.03	0
DMU66	0	0	0	0	0	0	0	0	0	0	1
DMU67	0.08	0	0	0	0	0	0	0.28	0.04	0.03	0
DMU68	0	0.13	0.09	0.14	0	0	0	0	0	0.01	0
DMU69	0	0.44	0	0	0	0	0.19	0	0	0	0
DMU70	0.25	0	0	0	0	0	0	0	0.19	0.03	0
DMU71	0	0.01	0.22	0.13	0	0	0	0	0	0.04	0
DMU72	0	0.43	0	0	0	0	0.4	0	0	0.12	0.01
DMU73	0	0	0	0.15	0	0	0.2	0	0	0.11	0
DMU74	0	0	0	0.36	0	0	0.01	0	0	0.01	0
DMU75	0.01	0	0	0	0	0	0	0.83	0	0	0.01

To compare DMU61, the best benchmark should be selected. Benchmarks for this bank can be DMU4 and DMU5. The amount of λ for these DMUs are 0.45 and 1.34 respectively. We can compare DMU61 to any of these benchmarks, but for better result we suggest that the inefficient DMU be compared with the benchmark with greater λ .

Table 5.3 Comparison Table for Mellat Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Mellat	DMU61	21,342	1,590	60950842453	41815841489	42493676844	27888739904
Parsian	DMU5	2,662	233	16582218885	14211995387	16063885991	11522489456

By looking at table 5.2, we see that the DMU61 is much bigger bank than DMU5. This difference in the size of these two banks makes them incomparable. To solve this problem, by comparing number of branches, it can be seen that number of branches for DMU61 is nearly 7 times greater than number of branches in DMU5. Thus, if we multiply DMU64 data to seven, then the comparison is more reasonable because of being data in the same range. Noting that, the number of branches is selected because it is a proper criterion to estimate the size of a bank.

Table 5.4 Comparison Table for Mellat bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Mellat	DMU61	21,342	1,590	60950842453	41815841489	42493676844	27888739904
Eghtesad Novin	DMU5	2,662	233	16582218885	14211995387	16063885991	11522489456
DMU5×7		18,634	1631	116075532195	99483967709	112447201937	80657426192

Bank Tejarat did not change much during 5 year, started 61.24% efficiency and in the last year the efficiency was 55.04%. Suggested inputs for this bank is proposed in table 5.5.

Table 5.5 Comparison for Tejarat Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Tejarat	DMU62	18,913	1715	36074214680	59.62%
Suggested Inputs		11348	1029	628124226	100%

We compare the performance of Tejarat Bank in the last year of the period with Bank Eghtesad Novin in the first year, according to the amount of λ in a similar manner explained before (Table 5.6).

Table 5.6 Comparison Table for Bank Tejarat

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Tejarat	DMU 62	18,913	1715	36074214680	27766470159	28444305513	15712049345
Eghtesad Novin	DMU 5	2,662	233	16582218885	14211995387	16063885991	11522489456
DMU5×7		18,634	1631	116075532195	99483967709	112447201937	80657426192

Table 5.7 shows the comparison between Saderat Bank and the best benchmark for it determined by λ .

Table 5.7 Comparison for Saderat Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Saderat	DMU63	31,138	2,600	46525571020	36105013542	36782848897	22261607719
Eghtesad Novin	DMU5	2,662	233	16582218885	14211995387	16063885991	11522489456
DMU5×11		29282	2563	182404407735	156331949257	176702745901	126747384016

Eghtesad Novin Bank has had a satisfying performance, this bank was efficient in first and second year, after that the efficiency is slightly decreased to 82.65%.

Table 5.8 Suggestions for Eghtesad Novin Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Eghtesad Novin	DMU65	3,162	251	14546518160	82.65%
Suggested inputs		2613	207	12022697259	100%

Based on table 5.2 and amount of λ , Eghtesad Novin bank should compare with itself in first year. The table below shows the amount of inputs and outputs for each bank.

The collected data are shown in table 1 in appendix.

Table 5.9 Comparison for Eghtesad Novin Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Eghtesad Novin	DMU65	3,162	251	422,139,957	325,471,535	2,399,545	268,402,751
Parsian	DMU5	2,662	233	16582218885	14211995387	16063885991	11522489456

Karafarin Bank could not improve during the period, started from 78.4% to 75.41% in the last year. Suggested inputs for the last year are proposed in table 5.10.

Table 5.10 Suggestions for Karafarin Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Karafarin	DMU67	1,673	107	4249043591	75.41%
Suggested inputs		1262	81	3204203772	100%

Table5.11 Comparison for Karafarin Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Karafarin	DMU67	1,673	107	123,307,245	104,602,671	2,667,863	81,260,142
Khavar miane	DMU30	216	8	1744462990	1167348433	2781559987	1042850894
DMU30×13		2808	104	22678018874	15175529624	36160279829	13557061628

Suggested inputs for Sarmaye bank are shown in table 5.12.

Table 5.12 Suggestions for Sarmayeh Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Sarmaye	DMU68	1,361	140	161,308,204	74.26%
Suggested inputs		1,010	104	119,787,472	100%

Based on λ values, it is better to compare this bank with DMU15. Table 5.13 compares these two banks.

Table 5.13 Comparison for Sarmayeh bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Sarmaye	DMU68	1,361	140	5558518401	3710560613	4388395968	3613531185
Khavar miane	DMU15	599	101	526321879.1	80673225.38	1932563830	242442383

Saman Bank is famous for its internet banking services. This bank was the first bank to provide internet banking to its customers. The efficiency of Saman Bank is more than 80% in every year of the period. therefore, by number of changes it can be an efficient bank.

Table 5.14 Suggestions for Saman Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Saman	DMU70	2,475	143	8915124604	86.39%
Suggested inputs		2138	124	7701776145	100%

Table 5.15 Comparison for Saman bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Saman	DMU70	2,475	143	8915124604	7282805376	7960640731	4393517195
Parsian	DMU4	4,531	288	29434510921	24126128601	25978019205	19659156844

The efficient amounts of inputs needed to be an efficient bank for Sina bank are proposed in table 5.16.

Table 5.16 Suggestions for Sina Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Sina	DMU71	2,424	257	169,134,175	67.78%
Suggested inputs		1643	174	114,639,144	100%

Table 5.17 Comparison for Sina Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Sina	DMU71	2,424	257	5828193487	4886054721	5563890076	3508212750
Ghavamin	DMU12	3,498	507	16303006778	15621705140	17473595745	14231050744
DMU12×0.5		1749	254	8151503389	7810852570	8736797873	7115525372

Ghavamin Bank was an efficient bank in the first year and second year, but after that the efficiency decreased down to 80%. In table 5.18 efficient amount of inputs are suggested.

Table 5.18 Suggestions for Ghavamin Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Ghavamin	DMU72	6,783	734	576,865,042	84.27%
Suggested inputs		5,716	618	486,124,171	100%

Table 5.19 Comparison for Ghavamin Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Ghavamin	DMU72	6,783	734	19878188904	19034555858	19712391213	16585507891
Eghtesad Novin	DMU5	2,662	233	16582218885	14211995387	16063885991	11522489456
DMU5×3		7986	699	49746656656	42635986161	48191657974	34567468368

Ansar Bank had improvements every year, but it is still far away from an efficient bank (63.64% efficiency in the last year).

Table 5.20 Suggestions for Ansar Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Ansar	DMU73	5,232	642	7,746,412,543	63.64%
Suggested inputs		3329	409	4,929,816,942	100%

Table 5.21 Comparison for Ansar Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Ansar	DMU73	5,232	642	7746412543	8012480979	8690316334	6469554549
Khavarmiane	DMU27	4,356	784	21293590350	20428235598	22042447152	16984632201

Bank Iran Zamin is the only bank with always decreasing efficiency. The efficiency of this bank in the last year of the period was only 9.05%. The collected data shown in table 1 in appendix demonstrated that this bank had minus value in annual profit output. The main problem of this bank is so much debts to the Central Bank of Iran.

Table 5.22 Suggestions for Iran Zamin Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	efficiency
Iran Zamin	DMU74	2,219	327	126,463,472	9.05%
Suggested inputs		200	30	392,202,360	100%

Table 5.23 Comparison for Iran Zamin Bank

Bank	DMU	No. of employees	No. of branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
Iran Zamin	DMU74	2,219	327	126,463,472	12,850,012	-813,093	17,721,494
Khavarmiane	DMU15	11	4	526321879	80673225	1932563830	242442383.7
DMU15×82		902	328	43158394088	6615204481	158470234043	19880275466

5.2 Conclusion

- Among the benchmark banks, Eghtesad Novin Bank is selected more than the others. is not only an efficient bank, but also is a good benchmark for inefficient banks.
- By calculating total deposits for all of the banks each year and the amount of changes each year, we will see a significant increase in the fourth year of the period. the cause of this sudden increase in deposits is the government policy that allowed banks to increase their interest rates up to 22%. Because of intense competition among banks, most of them presented the highest interest rates to their customers.

Another important reason that explains a sudden increase in total deposits is increasing foreign currency rates. By reason of increasing foreign currency rates, inflation is increased as well, and this set of factors led people to put their money in banks because investment under that situation had a lot of risks.

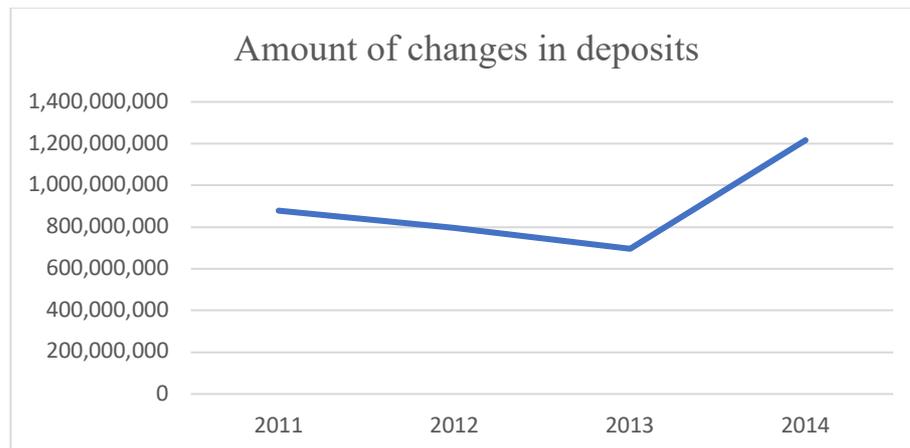


Figure 5.1 Amount of changes in deposits

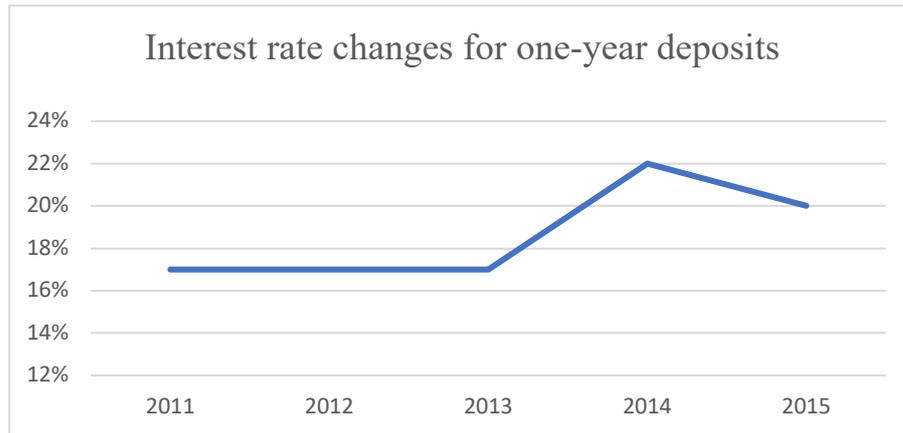


Figure 5.2 Interest rates during period

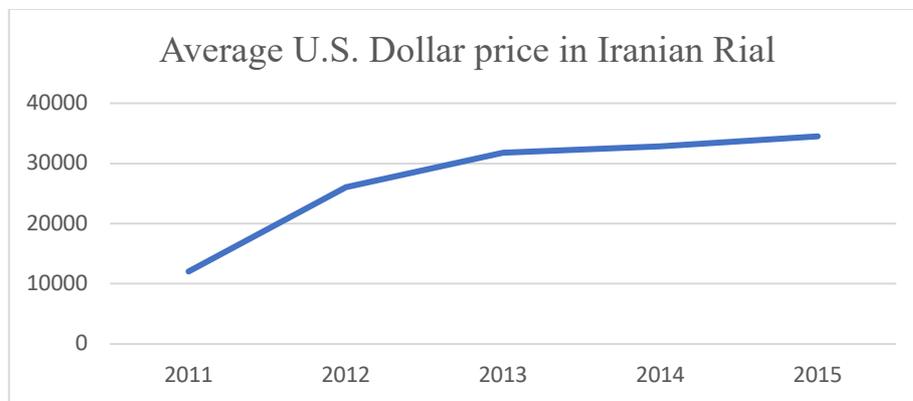


Figure 5.3 U.S Dollar price in Iranian Rial during period

- By extracting figures below from collected data, we will see that Given facilities are always significantly increased during these five years although the average interest rate for given facilities nearly still the same. One of the main causes of this issue is increasing demands. By increasing foreign currency rates and following that, increasing inflation, purchasing power of the community is significantly decreased. Therefore, to cover costs and increased prices, people intended to request for loan. This is one of the main reasons for increasing in total facilities.

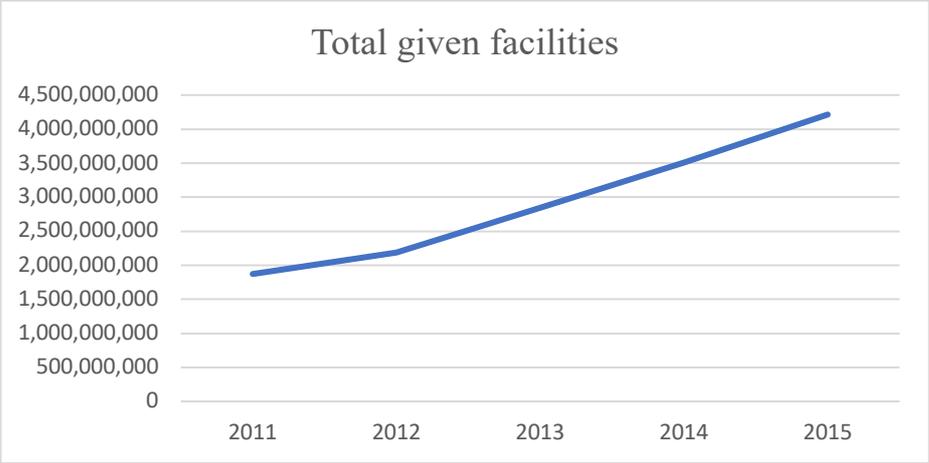


Figure 5.3 Total Private Banks Facilities

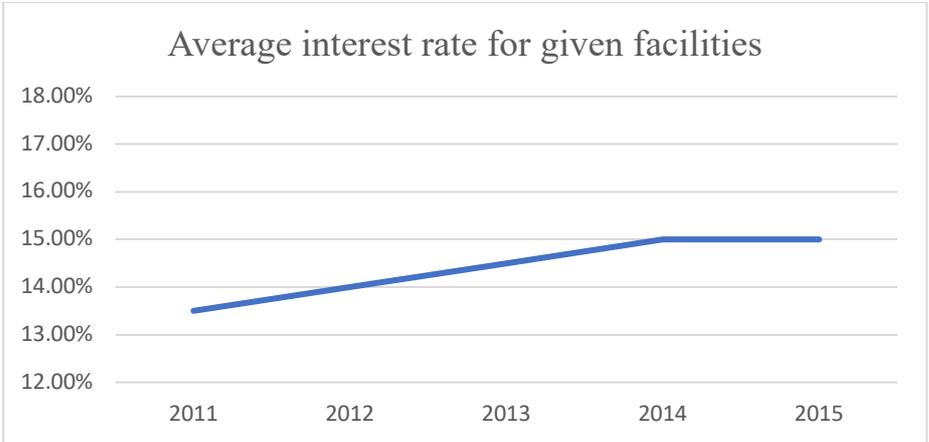


Figure5.4 Average Interest Rate

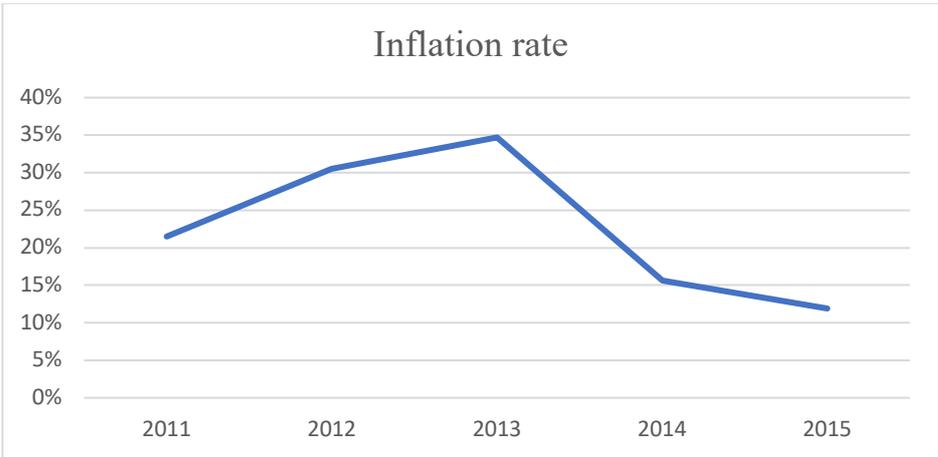


Figure 5.5 Inflation Rate During the period

- When foreign currency rates increased, private banks helped a lot to Iran economic system by attracting people's saving and money. As foreign currency rates increase, a lot of people want to invest in this field. and a great amount of Iranian Rial will turn to foreign currencies, which means the value of national currency will decrease in this situation.

What government and banks did was that the government allowed banks to increase their interest rates, and because of that, people tended to put their saving and money in banks with high-interest rate instead of buying foreign currencies with a lot of risk and tolerance in market. Therefore, banks could attract a lot of money and helped the government to manage the market.

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APPENDICES

Appendix 1: Collected data

Year	Name of The Bank	DMU Number	No. of Employees	No. of Branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
			I1	I2	I3	O1	O2	O3
2011	Mellat	DMU1	23,014	1,792	73221457729	52963105818	54814996423	34816844380
	Tejarat	DMU2	20,724	1,866	48328747693	37175887686	39027778290	26956009885
	Saderat	DMU3	33,465	2,972	57578695914	38706965449	40558856053	29957024572
	Parsian	DMU4	4,531	288	29434510921	24126128601	25978019205	19659156844
	Eghtesad Novin	DMU5	2,662	233	16582218885	14211995387	16063885991	11522489456
	Pasargad	DMU6	3,251	296	20881985125	16109252118	17961142723	12928533139
	Karafarin	DMU7	1,416	78	5668387686	3922011109	5773901713	2880201186
	Sarmaye	DMU8	1,466	145	5508249106	3038310676	4890201280	3182483431
	Shahr	DMU9	1,102	218	3604042553	2955134250	4807024854	1372027208
	Saman	DMU10	1,953	153	10337279703	7475129072	9327019676	6306236302
	Sina	DMU11	2,374	268	6286809264	5271688383	7123578987	4091360949
	Ghavamin	DMU12	3,498	507	16303006778	15621705140	17473595745	14231050744
	Ansar	DMU13	4,953	688	7776260403	8905488797	10757379401	8026967709
	Iran Zamin	DMU14	599	101	548221709.7	40003106.76	1891893711	4952833.741
	Khavar miane	DMU15	11	4	526321879.1	80673225.38	1932563830	242442383.7
2012	Mellat	DMU16	22,495	1,714	96975145495	67938817824	69553029378	37669971279
	Tejarat	DMU17	19,879	1,861	50446813146	39558610208	41172821763	26344608649
	Saderat	DMU18	32,730	2,793	58675591006	43027923601	44642135155	28883605531
	Parsian	DMU19	4,492	288	30322451584	24562884950	26177096504	19829606187
	Eghtesad Novin	DMU20	3,029	233	18567355654	15689577794	17303789348	12890010586
	Pasargad	DMU21	3,396	297	24377451666	19527125718	21141337272	14343445429
	Karafarin	DMU22	1,487	88	6323230182	4451739455	6065951009	3700548498
	Sarmaye	DMU23	1,444	144	6852272280	4778772854	6392984408	3898734367
	Shahr	DMU24	1,467	260	6710314541	5701591991	7315803545	2518526834
	Saman	DMU25	2,231	149	10855554161	8673651075	10287862629	6245730182
	Sina	DMU26	2,420	275	7641485721	6251410963	7865622518	3988371984
	Ghavamin	DMU27	4,356	784	21293590350	20428235598	22042447152	16984632201
	Ansar	DMU28	5,124	688	9990924586	11185309207	12799520762	836584687.3
	Iran Zamin	DMU29	2,075	352	1333708764	105305350.4	1719516905	209085097.7
	Khavar miane	DMU30	216	8	1744462990	1167348433	2781559987	1042850894
2013	Mellat	DMU31	22,157	1,633	74005758814	49656921603	50719001512	30632420334
	Tejarat	DMU32	19,365	1,826	46497698828	33849449706	34911529615	20757464392
	Saderat	DMU33	32,401	2,712	45503137250	34482731278	35544811187	22419105772
	Parsian	DMU34	4,477	293	25274416392	18756346742	19818426651	15450463960

	Name of The Bank	DMU Number	No. of Employees	No. of Branches	Total Assets	Total Deposits	Annual Net Profit	Given Facilities
	Eghtesad Novin	DMU35	3,119	251	15818127045	13360334107	14422414017	10511946817
	Pasargad	DMU36	3,486	326	20369862858	16504993575	17567073484	11834743372
	Karafarin	DMU37	1,547	98	5413200853	3828010529	4890090438	3146775228
	Sarmaye	DMU38	1,419	143	6619246909	3975794989	5037874899	4577283246
	Shahr	DMU39	2,151	261	7073213109	6206897144	7268977053	2802725933
	Saman	DMU40	2,425	148	10882411695	8586350683	9648430592	5556550726
	Sina	DMU41	2,446	277	6640028778	5071626370	6133706279	3412537390
	Ghavamin	DMU42	6,512	774	18587304195	17012349711	18074429620	14108479888
	Ansar	DMU43	5,115	688	9333686356	9982310728	11044390638	7832250094
	Iran Zamin	DMU44	2,349	351	2141162356	195205388.5	1257285298	433550672.2
	Khavar miane	DMU45	282	11	1657113547	1208448734	2270528643	148534096.4
2014	Mellat	DMU46	22,478	1,602	5592564585	36002275813	36760974544	25431025880
	Tejarat	DMU47	19,065	1,809	36482946195	27279340842	28038039573	16113873722
	Saderat	DMU48	31,944	2,608	42367836155	31398834998	32157533729	20593393952
	Parsian	DMU49	4,241	293	21299175801	15952493308	16711192039	12564801558
	Eghtesad Novin	DMU50	3,126	251	13129539939	11276091179	12034789910	9531198480
	Pasargad	DMU51	3,685	327	17134405832	13719444440	14478143171	10611307942
	Karafarin	DMU52	1,640	101	4322638909	3185580437	3944279168	2666375092
	Sarmaye	DMU53	1,379	143	5376023682	2739721603	3498420334	3689199637
	Shahr	DMU54	2,180	261	8307640529	6472073321	7230772052	3167557951
	Saman	DMU55	2,509	145	8798485941	6926957882	7685656613	4706162070
	Sina	DMU56	2,363	255	5746892352	4878721063	5637419794	3266947005
	Ghavamin	DMU57	6,766	763	17459481544	16327423805	17086122536	14492546727
	Ansar	DMU58	5,230	677	8039929109	8319807189	9078505920	6743660855
	Iran Zamin	DMU59	2,338	353	3042766228	248987233.4	1007685964	623269988.8
Khavar miane	DMU60	326	13	1596085856	1186057430	1944756162	1083639256	
2015	Mellat	DMU61	21,342	1,590	60950842453	41815841489	42493676844	27888739904
	Tejarat	DMU62	18,913	1715	36074214680	27766470159	28444305513	15712049345
	Saderat	DMU63	31,138	2,600	46525571020	36105013542	36782848897	22261607719
	Parsian	DMU64	4,403	293	22571729669	18077129118	18754964473	13242326430
	Eghtesad Novin	DMU65	3,162	251	14546518160	11215421606	11893256961	9248888732
	Pasargad	DMU66	3,815	327	18426085045	14577851551	15255686906	1.18E+11
	Karafarin	DMU67	1,673	107	4249043591	3604502791	4282338146	2800142729
	Sarmaye	DMU68	1,361	140	5558518401	3710560613	4388395968	3613531185
	Shahr	DMU69	2,195	270	12003438387	10189931427	10867766782	5196648070
	Saman	DMU70	2,475	143	8915124604	7282805376	7960640731	4393517195
	Sina	DMU71	2,424	257	5828193487	4886054721	5563890076	3508212750
	Ghavamin	DMU72	6,783	734	19878188904	19034555858	19712391213	16585507891
	Ansar	DMU73	5,232	642	7746412543	8012480979	8690316334	6469554549
	Iran Zamin	DMU74	2,219	327	4357803997	442798483.8	1120633839	610664851.8
Khavar miane	DMU75	349	16	2568719814	1286956168	1964791523	1847874431	

Appendix 2: Weights of each DMU of CCR model

Name	Index1	Index2	Index3	Index4	Index5	Index6	Name	Index1	Index2	Index3	Index4	Index5	Index6
DMU01	0.39	0.28	0.74	0.86	0	0	DMU38	5.73	0.59	7.95	7.4	1.33	0
DMU02	0.72	0.1	0.99	1.12	0	0	DMU39	0	6.16	6.18	6.62	0.33	0
DMU03	0.51	0.07	0.71	0.78	0	0.03	DMU40	5.63	0.58	7.81	7.27	1.31	0
DMU04	1.29	0.9	2.43	2.82	0	0	DMU41	2.02	0.28	2.79	3.09	0	0.13
DMU05	3.08	0.41	4.23	4.78	0	0	DMU42	3.52	0	4.8	4.19	0.8	0
DMU06	5.2	0	2.3	3.75	0	0.22	DMU43	2.55	0	37.19	0	7.3	0
DMU07	0	11.78	11.82	12.51	0.7	0.15	DMU44	2.98	43.87	47.55	50.43	3.15	0
DMU08	7.66	1.02	10.82	9.92	1.79	0.52	DMU45	1.22	0.16	1.67	1.89	0	0
DMU09	11.96	0	16.31	14.22	2.72	0	DMU46	0.86	0.11	1.18	1.33	0	0
DMU10	3.04	2.72	6.4	6.5	0.69	0.32	DMU47	0.6	0.08	0.82	0.93	0	0
DMU11	5.88	0.6	8.15	7.59	1.37	0	DMU48	1.64	1.15	3.09	3.58	0	0
DMU12	2.78	0.37	3.85	4.08	0.16	0.18	DMU49	2.44	1.71	4.63	5.25	0	0.22
DMU13	3.88	0	5.32	4.5	0.87	0.3	DMU50	2.72	0.36	3.74	4.23	0	0
DMU14	9.96	0	145.37	0	28.51	0	DMU51	0	12.67	12.77	13.21	0.74	0.82
DMU15	125.04	12.86	173.49	161.44	29.09	0	DMU52	7.94	1.06	11.22	10.28	1.86	0.54
DMU16	0.34	0.24	0.64	0.74	0	0	DMU53	5.13	0.6	7.15	6.7	1.2	0
DMU17	0.72	0.1	0.99	1.12	0	0	DMU54	0	7.16	7.17	7.69	0.38	0
DMU18	0.52	0.07	0.71	0.8	0	0	DMU55	6.19	0.64	8.58	7.99	1.44	0
DMU19	1.8	0.81	2.17	2.77	0	0	DMU56	2.06	0.28	2.84	3.15	0	0.13
DMU20	1.97	1.38	3.73	4.23	0	0.18	DMU57	3.7	0	5.08	4.3	0.83	0.28
DMU21	4.52	0	2.15	3.48	0	0	DMU58	2.84	0	25.54	0	4.72	4.14
DMU22	0	10.49	10.57	10.94	0.61	0.68	DMU59	0	47.72	48.08	49.73	2.79	3.1
DMU23	6.78	0.8	9.46	8.86	1.59	0	DMU60	0.45	0.32	0.86	0.99	0	0

Name	Index1	Index2	Index3	Index4	Index5	Index6	Name	Index1	Index2	Index3	Index4	Index5	Index6
DMU24	7.24	0	9.87	8.6	1.65	0	DMU61	0.87	0.12	1.19	1.35	0	0
DMU25	2.88	2.59	6.06	6.28	0.66	0	DMU62	0.59	0.08	0.8	0.91	0	0
DMU26	5.18	0.61	7.22	6.76	1.21	0	DMU63	1.56	1.1	2.95	3.42	0	0
DMU27	2.14	0.29	2.94	3.33	0	0	DMU64	2.28	1.6	4.33	4.91	0	0.21
DMU28	3.41	0	4.64	4.05	0.77	0	DMU65	2.55	0.35	3.53	3.91	0	0.16
DMU29	3.8	0	55.56	0	10.9	0	DMU66	0	12.51	12.55	13.28	0.75	0.16
DMU30	31.13	4.1	43.81	41.05	7.37	0	DMU67	7.81	1.04	11.05	10.13	1.83	0.53
DMU31	0.4	0.28	0.76	0.88	0	0	DMU68	4.04	0.54	5.54	6.27	0	0
DMU32	0.76	0.1	1.04	1.18	0	0	DMU69	0	7.13	7.14	7.66	0.38	0
DMU33	0.58	0.08	0.79	0.9	0	0	DMU70	6.01	0.71	8.38	7.85	1.41	0
DMU34	1.43	1.01	2.72	3.09	0	0.13	DMU71	1.93	0.26	2.66	2.95	0	0.12
DMU35	2.17	1.52	4.1	4.75	0	0	DMU72	3.76	0	5.16	4.37	0.84	0.29
DMU36	2.46	0.33	3.37	3.81	0	0	DMU73	7.81	0	10.72	9.08	1.76	0.6
DMU37	0	11.21	11.29	11.68	0.65	0.73	DMU74	0	45.91	28.42	35.49	0	2.84
DMU38	6.88	0.92	9.73	8.92	1.61	0.47	DMU75	5.73	0.59	7.95	7.4	1.33	0

Appendix 3: Values of λ for each DMU

Name	DMU4	DMU5	DMU12	DMU15	DMU19	DMU21	DMU27	DMU30	DMU45	DMU46	DMU66
DMU1	0.19	2.51	0	0	0	0	0	0	0	0.35	0
DMU2	0	1.16	0	0	0	0	0.4	0	0	0.35	0
DMU3	0	0.78	0	0	0	0	0.49	0	0	0.49	0
DMU4	1	0	0	0	0	0	0	0	0	0	0
DMU5	0	1	0	0	0	0	0	0	0	0	0
DMU6	0	0.88	0	0	0	0.18	0	0	0	0	0
DMU7	0.08	0	0	0	0	0	0	0.8	0.38	0.02	0
DMU8	0	0.09	0.06	0.83	0	0	0	0	0	0.02	0
DMU9	0	0	0	0.9	0	0	0.11	0	0	0.02	0
DMU10	0.03	0.39	0	0	0	0	0	0.66	0	0.01	0
DMU11	0	0	0.18	0.76	0	0	0.05	0	0	0.04	0
DMU12	0	0	1	0	0	0	0	0	0	0	0
DMU13	0	0	0	0.76	0	0	0.21	0	0	0.12	0.01
DMU14	0	0	0	0.59	0	0	0	0	0	0.02	0
DMU15	0	0	0	1	0	0	0	0	0	0	0
DMU16	0.91	2.62	0	0	0	0	0	0	0	0.24	0
DMU17	0	1.23	0	0	0	0	0.5	0	0	0.33	0
DMU18	0	1.12	0	0	0	0	0.39	0	0	0.53	0
DMU19	0	0	0	0	1	0	0	0	0	0	0
DMU20	0.28	0.62	0	0	0	0	0	0	0	0	0
DMU21	0	0	0	0	0	1	0	0	0	0	0
DMU22	0.12	0	0	0	0	0	0	0.86	0	0.02	0
DMU23	0	0.22	0.08	0.57	0	0	0	0	0	0.01	0
DMU24	0	0	0	0.64	0	0	0.26	0	0	0.01	0
DMU25	0.24	0.09	0	0	0	0	0	0.61	0	0.02	0
DMU26	0	0.04	0.29	0.53	0	0	0	0	0	0.03	0
DMU27	0	0	0	0	0	0	1	0	0	0	0
DMU28	0	0	0	0.53	0	0	0.34	0	0	0.12	0
DMU29	0	0	0	0.42	0	0	0	0	0	0.02	0
DMU30	0	0	0	0	0	0	0	1	0	0	0
DMU31	0.76	1.41	0	0	0	0	0	0	0	0.31	0
DMU32	0	0.94	0	0	0	0	0.47	0	0	0.3	0
DMU33	0	0.65	0	0	0	0	0.33	0	0	0.51	0
DMU34	0.71	0.05	0	0	0	0	0	0	0	0.02	0
DMU35	0.1	0.7	0	0	0	0	0	0	0	0.03	0
DMU36	0	1.07	0	0	0	0	0.05	0	0	0.01	0
DMU37	0.1	0	0	0	0	0	0	0.54	0	0.02	0
DMU38	0	0.16	0.07	0.33	0	0	0	0	0	0.01	0.01
DMU39	0	0	0.34	0.23	0	0	0	0	0	0.03	0

Name	DMU4	DMU5	DMU12	DMU15	DMU19	DMU21	DMU27	DMU30	DMU45	DMU46	DMU66
DMU40	0.29	0	0	0	0	0	0	0	0.47	0.03	0
DMU41	0	0	0.22	0.32	0	0	0.02	0	0	0.03	0
DMU42	0	0.2	0	0	0	0	0.5	0	0	0.11	0
DMU43	0	0	0	0.28	0	0	0.29	0	0	0.11	0
DMU44	0	0	0	0.39	0	0	0	0	0	0.01	0
DMU45	0	0	0	0	0	0	0	0	1	0	0
DMU46	0	0	0	0	0	0	0	0	0	1	0
DMU47	0	0.48	0	0	0	0	0.46	0	0	0.31	0
DMU48	0	0.62	0	0	0	0	0.24	0	0	0.49	0
DMU49	0.48	0.21	0	0	0	0	0	0	0	0.04	0
DMU50	0.04	0.61	0	0	0	0	0	0	0	0.04	0.01
DMU51	0	0.82	0	0	0	0	0.04	0	0	0.04	0
DMU52	0.07	0	0	0	0	0	0	0.37	0	0.03	0
DMU53	0	0.07	0.07	0.26	0	0	0	0	0	0.01	0.01
DMU54	0	0.08	0.3	0.02	0	0	0	0	0	0.02	0
DMU55	0.22	0	0	0	0	0	0	0	0.3	0.03	0
DMU56	0	0	0.21	0.17	0	0	0.01	0	0	0.04	0
DMU57	0	0.19	0	0	0	0	0.44	0	0	0.12	0.01
DMU58	0	0	0	0.18	0	0	0.21	0	0	0.11	0
DMU59	0	0	0	0.33	0	0	0	0	0	0.01	0
DMU60	0.03	0	0	0	0	0	0	0.44	0	0	0
DMU61	0.45	1.34	0	0	0	0	0	0	0	0.33	0
DMU62	0	0.61	0	0	0	0	0.37	0	0	0.32	0
DMU63	0	0.77	0	0	0	0	0.32	0	0	0.52	0
DMU64	0.64	0.08	0	0	0	0	0	0	0	0.04	0
DMU65	0.09	0.56	0	0	0	0	0	0	0	0.03	0
DMU66	0	0	0	0	0	0	0	0	0	0	1
DMU67	0.08	0	0	0	0	0	0	0.28	0.04	0.03	0
DMU68	0	0.13	0.09	0.14	0	0	0	0	0	0.01	0
DMU69	0	0.44	0	0	0	0	0.19	0	0	0	0
DMU70	0.25	0	0	0	0	0	0	0	0.19	0.03	0
DMU71	0	0.01	0.22	0.13	0	0	0	0	0	0.04	0
DMU72	0	0.43	0	0	0	0	0.4	0	0	0.12	0.01
DMU73	0	0	0	0.15	0	0	0.2	0	0	0.11	0
DMU74	0	0	0	0.36	0	0	0.01	0	0	0.01	0
DMU75	0.01	0	0	0	0	0	0	0.83	0	0	0.01