

## Measuring gas demand security using Principal Component Analysis: A case study

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### Abstract

Safeguarding the energy security is an important energy policy goal of every country. Assuring sufficient and reliable resources of energy at affordable prices is the main objective of energy security. Due to such reasons as special geopolitical position, terrorist attacks and other unrest in the Middle East, securing Iran's energy demand and increasing her natural gas exports have turned into a critical issue. The aim of this paper is to develop a composite index for evaluating the gas demand security. To this purpose, six individual indices (i.e. gas exports dependency, transportation cost, economic dependency, political stability of the exporting country, political stability of the importing country, and the purchasing power of the importing country) are identified and the Principle Component Analysis (PCA) method is employed to weigh and combine the indices into GDSI (Gas Demand Security Index). The results show an interesting counter-intuitive phenomenon that the political stability of the importing and exporting countries have respectively the most and the least effects on the obtained composite index.

**Keywords:** Gas demand security index, natural gas export, Principal Component Analysis (PCA)

### 1-Introduction

Natural gas has become the third most consumed energy in the world, and its share in the global energy consumption has reached the point 23.7% (BP statistical review of world energy. 2002-2014, 2014). The level of natural gas consumption has increased from 1960.14 bcm (billion cubic meters) in 1990 to 3468.6 bcm in 2015 and raised to around 3630 bcm in 2016, with expected increase rate of 1.6% per year (Gas. Analysis and Forecasts to 2022, 2017). Due to its flexibility, versatility, reliability, high efficiency, easy storage and transportation, capability of being highly compressed, and having less environmentally harmful impacts, compared to other fuels, natural gas is gaining much share in the global energy mix in the recent years (International Energy Agency (IEA), 2014).

It is anticipated that the natural gas consumption will have the fastest growing rate among other energy sources and is estimated to triple by 2025 in the Asian emerging markets which are going to be the center of this growth (Energy Information Administration (EIA), 2005).

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China is taking EU's place to become the second largest market in terms of natural gas consumption after US (Zhi-Guo, Cheng and Dong-Ming, 2018).

Considering the rich reserves of natural gas, high success rate in the new gas fields exploration, and increasing growth of fossil fuels demand in highly import dependent countries such as China, European Union, India, and Turkey which are potential markets for Iran's natural gas export (through pipelines or in the LNG form), There is a promising outlook for Iran's natural gas exports.

However, recent political instabilities, natural disasters, and terrorist attacks in some exporting/ importing countries of the Middle East have caused the energy security to be considered in different countries in its various supply chain layers including production, distribution, conversion, transportation, supply, and demand. Out of these layers, supply and demand are more important and have been addressed more both politically and strategically because factors affecting them are mostly out of a government's control; they highly depend on the energy exporting/importing country. On the other hand, energy experts believe that use of oil or gas weapon by a major oil or gas exporting country is a threat to world's energy security (Alhajji, 2007). Hence, securing energy demand and increasing natural gas exports of Iran has become an interesting topic because of her specific geopolitical position and other reasons mentioned earlier.

The 2017 and 2018 World Economic Forum reports (The Global Risks, 2017), and (The Global Risks, 2018) clearly suggest various environmental, geopolitical, and social risks could bring economics insecurities. Therefore, for all gas exporting countries, including Iran, demand security is important. This paper aims to quantify and evaluate this security based on six individual indices: dependency on gas export, transportation cost, economic dependency, exporting country's political stability, importing country's political stability, and the purchasing power of the importing country. The gas demand security depends on various factors ought to be considered collectively not individually. In other words, due to the correlation between the individual indices, determining the impact of each one on the gas demand security without considering other indices is impossible and we need to address the issue through a holistic approach. To determine the GDSI by weighting and combining the individual indices, the PCA method is employed to calculate the weights and avoid assigning ad hoc and arbitrary weights to the indices. The 1993-2015 time series data of each index on an annual basis are used as the input variables for the PCA method. The GDSI measures the gas demand security of Iran quantitatively with higher values meaning higher demand security against disruptions and shocks in the demand. We also calculate the contribution of each individual index in determining the proposed model. Through the proposed GDSI, policy-makers can analyze Iran's gas demand security status in the future under different scenarios and select the potential energy markets to maximize the value of GDSI.

The scientific originality of this paper lays in considering demand side of a country's energy security performance for the first time using composite index.

The rest of the paper has been structured as follows: Section 2 presents the literature on the energy supply and demand security. Section 3 discusses Iran's status in terms of production, consumption, and major gas importers and the methodology used for the selection of individual indices. Section 4 constructs the GDSI using PCA method and analyzes the results. Section 5 discusses the numerical results and finally Section 6 presents conclusions and provides ideas for further researches.

## **2-Literature review**

Energy security means ensuring the adequate amount of energy for people, economic and social activities, and defensive targets at a reasonable price (Koyama and Kutani, 2012). Although there are a large number of energy security definitions provided by researchers and policy-makers, there is no precise definition mainly due to its multidimensional and esoteric nature. Majority of studies address the energy supply security or energy security in its general form. Accordingly, it is possible to define the energy demand security as the "availability of a sustainable and continuous demand for energy exports at a market price that covers the transaction and production costs" (Dike, 2013).

One can classify the related studies at the micro level (rural areas, villages, islands, etc.) or at the macro level (a country or a set of countries) utilizing individual or composite indicators, type of energy examined, and the focus on the demand or on the supply security (Narula and Reddy, 2015).

To date, almost all of the studies on energy security have addressed the supply aspect, while works on the demand side are few. Further, most studies that have examined the supply aspect, concentrate more on the primary energy sources, particularly crude oil and natural gas. These studies adopt an aggregated approach and construct a composite index.

Gupta (2008) measures 26 oil-importing countries' vulnerability to supply disruptions and constructs by using the PCA method, a composite index named Oil Vulnerability Index (OVI). Biresselioglu, Yelkenci and Oz (2015) evaluate the supply security of the natural gas in 23 importing countries using six individual indices which have been aggregated to supply security index (SSI) through PCA method. M. Radovanović, Filipović and Pavlović (2017) develop a composite energy security index (ESI) for 28 European Union countries with a sustainable approach, proposing six individual indicators which have been combined to ESI through PCA method. Proposing 23 individual indicators, Wang and Zhou (2017) employ the Subjective and Objective Weight Allocation method to develop a framework for evaluation of the energy security in different countries. They apply the Z-standardization and PCA methods to normalize and weigh the indicators respectively. Erahman *et al.* (2016) assess the performance of Indonesia's energy security through constructing an energy security composite index based on fourteen individual indicators which have been weighted through PCA and equal weights methods. Narula, Sudhakara Reddy and Pachauri (2015) develop a Sustainable Energy Security (SES) index for India's energy demand sub-system by using a hierarchical structure. Shaikh, Ji and Fan (2016) construct a holistic indicator which measures the network system stability based on Ecological Network Analysis (ENA) approach to evaluate the LNG supply stability for top five gas importing countries in Asia Pacific region.

Scope of some studies has been confined to such micro scales as rural regions and manufacturing companies.

Zhang and Su (2016) propose ten individual indicators for China's sustainable development of rural household energy and have aggregated them to establish the "Rural Energy Sustainable Development Index" using a novel method called grouped PCA. Mainali *et al.* (2014) measure the energy sustainability of rural regions of five developing countries by constructing a composite index called "Energy Sustainability Index" and have used the PCA method to aggregate thirteen individual indicators into that composite index. Zhang *et al.*

Not many studies have used the aggregated approach; they have proposed several individual indicators without aggregating them into a composite index.

Dastan and Selcuk (2016) review the energy supply security of the electricity and natural gas in Turkey in 2012 and suggest several indicators to measure the underlying risks. Tongsopt *et al.* (2016) study the energy security of the ASEAN (Association of Southeast Asian Nations) countries based on a 4-A individual indicator framework: Applicability, Availability, Affordability, and Acceptability. Radovanović, Filipović and Pavlović (2017a) determine the impact level of five variables on energy security of 28 European states between 1990 and 2012 using panel data analysis.

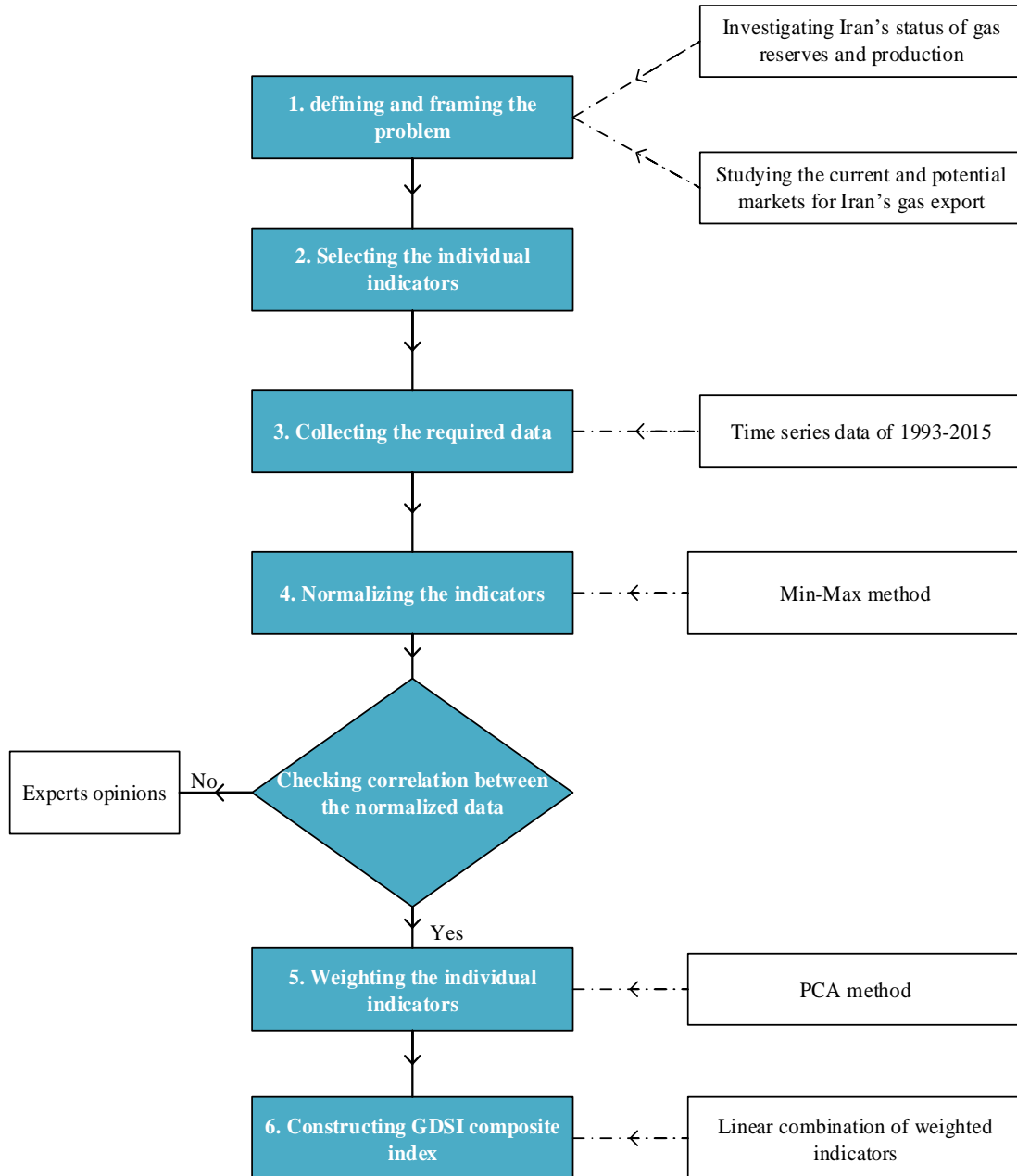
As mentioned earlier, studies that have investigated the energy demand security are very rare, however, as an example, Dike (2013) addresses the exports demand security of oil and gas in OPEC members with a disaggregated approach and develop two main indicators that measure the level of energy demand security risks.

With respect to the existing literature in the energy security, from the international perspective, there are few contributions in the demand side since the most countries/ regions are net importers, and as a result the supply side of those countries/ regions' energy security ought to be considered. The only study addressing energy demand security employs a disaggregated approach which develops several indicators to investigate the exports demand security of OPEC members. On the other hand, From the national perspective, all of research work with qualitative or quantitative indicators, investigate the demand or supply security with a disaggregated approach. Therefore, to fill the research gaps, we seek to

quantitatively measure Iran’s gas demand security performance in recent years with a composite index based on six individual indicators which are weighted and then aggregated to that composite index through PCA method.

### 3-Methodology

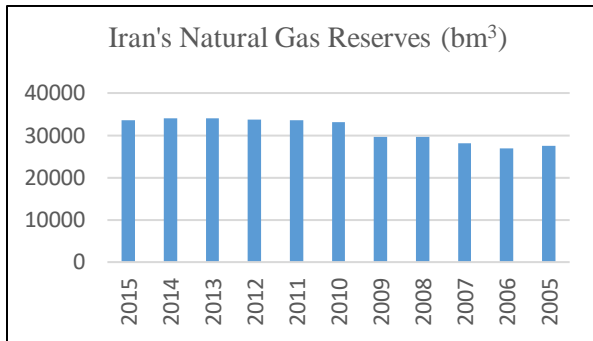
This section details the methodology employed. Fig 1 presents an outline of the steps for constructing the GDSI.



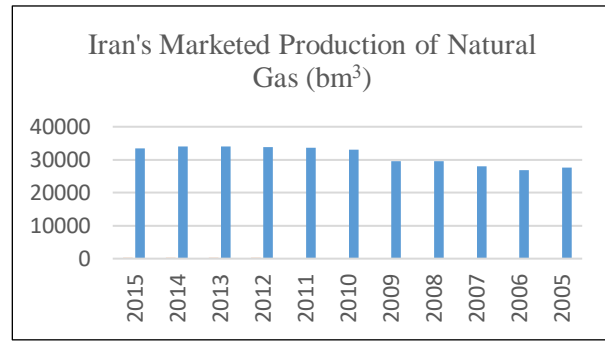
**Fig 1.** Steps for constructing GDSI (Source: Authors)

### 3-1-Iran’s situation of natural gas reserves and production

After Russia, Iran has the second-largest proved reserves of natural gas in the world. 17% of the world’s total natural gas reserves and more than 33% of OPEC’s belong to Iran. Further, Iran’s rate of natural gas exploration is estimated to be 79% which is approximately 44- 49% higher than the world’s average (U.S. Energy Information Administration, 2015). However, most of these rich reserves of natural gas are assigned to meet the domestic consumption causing Iran’s share in the global gas trade to be very low. This is mainly due to the lack of modern infrastructures for importing or exporting natural gas in the LNG form. As shown in Fig 2, natural gas reserves of Iran reached their maximum in 2014. Fig 3 illustrates Iran’s marketed production of natural gas between 2005 and 2015.



**Fig 2.** Iran’s proved natural gas reserves by the year (source: OPEC ASB from 2005 to 2016)



**Fig 3.** Iran’s marketed production of natural gas by the year (source: OPEC ASB from 2005 to 2016)

### 3-2-Current gas exports markets of Iran

In spite of having rich natural gas reserves and incremental exploration rate of new gas fields, Iran accounts for less than 1% of the international gas trade (U.S. Energy Information Administration, 2015). The country has failed to increase the export despite various efforts. to date, only Turkey, Azerbaijan, and Armenia import natural gas from Iran. With 7.8 bcm import in 2015 Turkey is the main export market for Iran’s natural gas (*BP Statistical Review of World Energy*, 2016). Based on a swap deal, Iran exports gas to Armenia and receives part of the electricity produced in Hrazdan in return. Similarly, Iran-Azerbaijan gas trade is actually a swap contract by which Iran imports gas from Azerbaijan and exports it to Nakhchivan autonomous republic in Azerbaijan (Jalilvand, 2013). In this research, the amount of gas exported to Nakhchivan has been regarded as that exported to Azerbaijan.

### 3-3-Indicators for assessing Iran’s gas demand security

In this paper, six individual indicators are selected and classified in two categories: first, indicators that address the exporting country (Iran) and second, indicators that address the importing country/countries. The former consists of three indicators: gas exports dependency, economic dependency, and political stability of the exporting country. Similarly, the latter contains three indicators: political stability of the importing country, purchasing power of the importing country, and transportation costs. We have collected the time series data of each indicator between 1993 and 2015. The six indicators are scrutinized as follows:

- Gas exports dependency ( $X_1$ )

This indicator stands for the ratio of Iran’s gas exports value to the total exports value in a given year.

$$\frac{\text{Gas exports value}}{\text{Total exports value}} \tag{1}$$

- Economic dependency ( $X_2$ )

This indicator stands for the ratio of Iran's gas exports value to the GDP.

$$\frac{\text{Gas exports value}}{\text{GDP}} \quad (2)$$

- Political stability of the exporting country ( $X_3$ )

To quantify this index, we use ICRG (International Country Risk Guide) index developed by PRSG (Political Risk Services Group). ICRG index is ranged between 0 to 100 in which values between 80 and 100 refer to the countries with a very low political risk and values between 0 and 50 refer to the countries with a very high political risk (Howell, 2011).

- Political stability of the importing country ( $X_4$ )

Similar to the previous index, the political stability of the three importing markets of Iran's natural gas including Turkey, Azerbaijan, and Armenia has been assessed according to their ICRG ratings. Ratio of the natural gas volume exported to each of these countries to the total natural gas exports of Iran in a year is considered as the weight to construct the weighted average of the political stability index in the importing countries. Therefore, we use this weighted average in the calculations.

$$\sum_{i=1} \frac{\text{Import}_i}{\text{TotalExports}} \times \text{ICRG}_i \quad (3)$$

Where  $\frac{\text{Import}_i}{\text{TotalExports}}$  denotes the share of country  $i$  in Iran's total gas exports in a specific year.

- Transportation costs ( $X_5$ )

This index is stemmed from infrastructural and transportation disruptions. Similar to what Dike (2013) and Le Coq and Paltseva (2009) have done, the distance between the capitals of the importing and exporting countries is employed as a proxy for the gas transportation's potential risks. Regarding distance, this index can take three different values as shown in equation (4). The longer is the distance between the two capitals, the more will be the vulnerability of the exporting country to disruptions in the gas imports of the importing countries because of the technical failures in the pipelines and other gas production facilities. Since this index has only three values in different years, we use ICRG rating of each importing country as the weight to develop the weighted average of the transportation costs as shown in equation 0.

$$D_i = \begin{cases} 1 & \text{distance} < 1500 \text{ km} \\ 2 & 1501 \text{ km} < \text{distance} < 4000 \text{ km} \\ 3 & \text{distance} > 4001 \text{ km} \end{cases} \quad (4)$$

$$\sum_{i=1} \frac{ICRG_i}{TotalICRG} \times D_i \quad (5)$$

- Purchasing power of the importing country ( $X_6$ )

The GDP of each importing country in a given year represents the purchasing power of that country. Since in this paper there are three importing countries with different GDP values, we consider the average of each country's GDP in a given year for this index.

### 3-4- Collecting data

Time series data for six individual indicators have been derived from the World Bank annual reports of such international energy associations as the British Petroleum (BP), International Energy Agency (IEA), Energy Information Administration (EIA), and Oil Producing Exporting Countries (OPEC), and direct contacts with the Plan and Budget Organization (PBO) gas experts (See Table 7 and Table 8 in Appendix).

### 3-5-Normalizing indicators

Contrary to other researches that have addressed the energy security in multiple countries in a specific year, this paper has focused only on one single country (Iran). Hence, the required data of each index had to be collected for several years. Accordingly, we have used the time series data of each index between 1993 and 2015. Then, the Min-Max method has been used to normalize the indicators.

### 3-6-Weighting indicators

PCA is a multivariate statistical method used for studying a set of correlated variables in such various fields as economy, social issues, environmental concerns, job satisfaction, political situation, security of energy demand or supply and to some extent supply chain network design (Moradi *et al.*, 2018). For more information about supply chain and its applications, interested readers can refer to (Sabbaghnia *et al.*, 2018), (Razmi and Sabbaghnia, 2015), and (Rabbani *et al.*, 2018).

The logic behind the PCA is to transform a set of correlated to uncorrelated variables called principal components or PCs, so that they are a linear combination of the initial variables. These PCs have two major attributes: first, they are uncorrelated, i.e. the covariance among them is zero and second, their variance from the first PC to the last declines in a descending order. In this research, the PCA approach is employed mainly, due to its capability in index construction and data dimensionality reduction. In fact, in lieu of considering six individual indicators, we have based our analysis on a PC, which is the linear combination of these indicators. In other words, the latent or unobserved variable in this study is Iran's gas demand security index called GDSI which is the linear combination of six individual indicators.

## 4-Constructing GDSI using PCA

PCA consists of the following steps: forming the correlation or covariance matrix of indicators, calculating the Eigenvalues, and finally determining the coefficients or weights. Using time series data of the individual indicators during 1993-2015 as the input variables for the Minitab statistical software requires that two prerequisites be met: first, there is no correlation among the time series data of each individual indicator, and second, at least two of the individual indicators must be correlated. For the first constraint, we check the interdependence among time series data of each index. If there is no correlation among them, they will be interdependent and using them for the PCA method will be possible. However, when they are correlated, their being stationary or non-stationary must be investigated. If they are stationary, we can apply them as inputs for the PCA method, but if the time series is non-stationary, a first (or higher) order difference of that time series must be calculated to make that time series stationary. therefore, we adjust the indicators' data to tailor them for the PCA method. Using the Pearson correlation test in Minitab, the correlation among the indicators is calculated and the results are shown in Table 1. As

shown, the results clearly indicate that the individual indicators are correlated enough to implement the PCA method.

To obtain the Eigenvalues ( $\lambda_i$ ), the following equation must be solved for  $\lambda$  where I is the identical matrix and R is the covariance matrix of the normalized indicators according to Table 2, calculated eigenvalues are shown in Table 3.

$$|R - \lambda I| = 0 \quad (6)$$

**Table 1.** results of the correlation test among the normalized indicators

indicators	X <sub>1</sub>	X <sub>2</sub>	X <sub>3</sub>	X <sub>4</sub>	X <sub>5</sub>	X <sub>6</sub>
X <sub>1</sub>	1					
X <sub>2</sub>	0.439	1				
X <sub>3</sub>	0.155	-0.419	1			
X <sub>4</sub>	0.004	0.158	-0.197	1		
X <sub>5</sub>	-0.295	-0.314	-0.020	0.017	1	
X <sub>6</sub>	-0.068	-0.512	0.368	-0.525	-0.076	1

**Table 2.** Covariance matrix

indicators	X <sub>1</sub>	X <sub>2</sub>	X <sub>3</sub>	X <sub>4</sub>	X <sub>5</sub>	X <sub>6</sub>
X <sub>1</sub>	0.01413					
X <sub>2</sub>	0.01050	0.04046				
X <sub>3</sub>	0.00170	-0.00780	0.00818			
X <sub>4</sub>	0.00012	0.00816	-0.00446	0.06263		
X <sub>5</sub>	-0.00118	-0.00594	0.00083	-0.00961	0.01148	
X <sub>6</sub>	-0.00074	-0.00949	0.00299	-0.01185	0.00221	0.00812

As shown in Table 3,  $\lambda_1 > \lambda_2 > \lambda_3 > \lambda_4 > \lambda_5 > \lambda_6$  because  $\lambda_j = \text{Var}(PC_j)$  and variance of the PC<sub>j</sub> changes in a descending order for j=1, 2, ..., 6. The scree plot of the individual indicators has been illustrated in Fig 4 and implies that the first three PCs are most important in establishing the GDSI. As mentioned earlier, we want to provide only one composite indicator called GDSI, hence, the coefficient of each individual indicator belonging to the first PC (not the first three PCs) with utmost importance in determining the GDSI, has been considered as the weight of that indicator as shown in table 4.

Due to using min-max in lieu of z-standardization method for data normalization, as shown in table 3, cumulative amount of eigenvalues is not equal to the number of the indicators, i.e. 6.

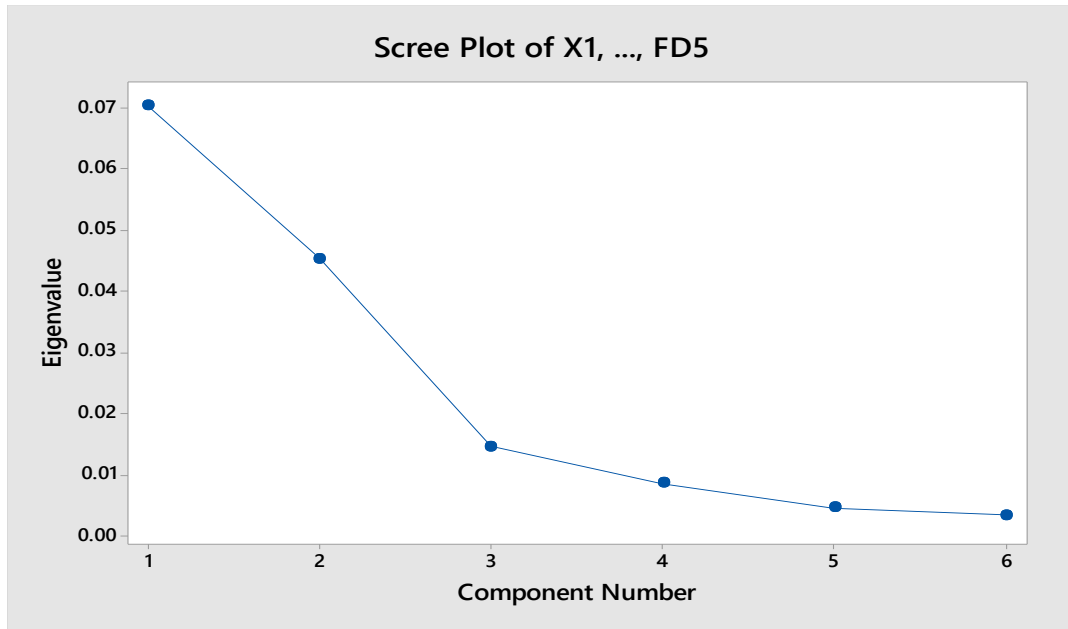
**Table 3.** Eigenvalues

	$\lambda_1$	$\lambda_2$	$\lambda_3$	$\lambda_4$	$\lambda_5$	$\lambda_6$
<b>Eigenvalues</b>	0.070322	0.045326	0.014438	0.008490	0.004459	0.003284
<b>Variability</b>	0.481	0.310	0.099	0.058	0.030	0.022
<b>Cumulative</b>	0.481	0.790	0.889	0.947	0.978	1.000



**Table 4.** Weights of the indicators

Indicator	PC <sub>1</sub>	PC <sub>2</sub>	PC <sub>3</sub>	PC <sub>4</sub>	PC <sub>5</sub>	PC <sub>6</sub>
X <sub>1</sub>	0.085	-0.307	0.731	0.423	-0.391	0.181
X <sub>2</sub>	0.410	-0.806	-0.202	0.010	0.211	-0.312
X <sub>3</sub>	-0.123	0.106	0.474	0.130	0.838	-0.171
X <sub>4</sub>	0.868	0.454	0.136	-0.046	-0.028	-0.140
X <sub>5</sub>	-0.040	0.072	0.228	-0.216	-0.315	-0.861
X <sub>6</sub>	-0.235	0.185	-0.360	0.869	-0.017	-0.281



**Fig 4.** Scree plot of X<sub>1</sub>, ..., X<sub>6</sub>

As a result, according to table 4, the GDSI is calculated through the following equation where  $t$  refers to a specific year between 1993 and 2015 and  $X_1, \dots, X_6$  represent the normalized values of the gas exports dependency, economic dependency, political stability of the exporting country, political stability of the importing country, transportation costs, and the purchasing power of the importing country respectively.

$$GDSI_t = 0.085X_1 + 0.410X_2 - 0.123X_3 + 0.868X_4 - 0.040X_5 - 0.235X_6 \quad (7)$$

## 5-Discussing the results

According to the proposed PCA model, the GDSI value is calculated for each year between 1993 and 2015. As shown in Table 5, the GDSI value is negative only in 1997 while it has reached its highest value in 2013. In other words, GDSI can assume any real number whether positive or negative. Hence, the sign

of the GDSI must be regarded in comparing its values through the time period. Results indicate that the higher is the value of GDSI in a year, the higher will be the gas exports demand security in the same year.

**Table 5.** GDSI value between 1993 and 2015

Year	GDSI value	Year	GDSI value
1993	0/566086389	2005	0/833327283
1994	0/522207506	2006	1/099977471
1995	0/654801101	2007	0/991071845
1996	1/017082473	2008	0/987196679
1997	-0/24163718	2009	0/842293447
1998	0/717419121	2010	1/209840801
1999	0/638884041	2011	1/261115811
2000	0/776024517	2012	1/365236111
2001	0/680307657	2013	1/914609932
2002	1/322840969	2014	1/780760975
2003	0/746905841	2015	1/531611089
2004	0/508429472		

It is important to note that the calculated weights should not be interpreted as the partial regression coefficients, because contrary to the typical linear regression, the latent or dependent variable of the model, i.e. the GDSI index, has not been considered in the model. Hence, more weight or coefficient of a variable does not necessarily mean more contribution of that variable in establishing the composite index. Similarly, a variable sign (positive or negative) does not mean its direct or indirect relation with the GDSI. For example, the purchasing power of the importing country ( $X_6$ ) is directly related to the GDSI, however, its coefficient in the PCA model is negative.

The average share of the indicators in the model can be estimated based on the weights obtained for each indicator. Accordingly, following Gupta (2008), the normalized value of every indicator in a specific year has been multiplied by its weight in the model and then divided by the value of the GDSI in the same year to calculate the average share of every indicator in the GDSI (Table 6).

**Table 6.** average share of each index in the PCA model

	$X_1$	$X_2$	$X_3$	$X_4$	$X_5$	$X_6$
<b>Average share (%)</b>	6.72	30.75	7.16	37.42	2.21	15.72

As shown, there is a considerable difference among the shares of the indicators in determining the GDSI. The political stability of the importing countries ( $X_4$ ) has the most share. Obviously, it can be concluded that the effect of the political stability of the importing country on the GDSI is significantly higher than that of the exporting country.

Considering the significance of the indicators in the proposed GDSI, to reduce the vulnerability to possible disruptions in the gas demand of the importing countries, Iran's energy policymakers should: (a) accord high priority to the political stability of the importing countries, (b) lessen the share of the natural gas in Iran's total exports value by adopting appropriate economic policies, and (c) select the potential exports markets with high GDP values, i.e. high purchasing power.

## 6-Conclusions

The main contribution of this research was the development of a composite index (GDSI) to assess Iran's gas demand security performance based on six individual indicators which have been weighted and

then aggregated to that composite index through PCA method. Accordingly, the main objective of this study is to establish a composite index to quantify and evaluate Iran's gas exports demand security for which there are many influential individual indicators. Analyzing Iran's situation of the natural gas reserves and production, and investigating the current and potential gas exports markets in this research led to proposing six individual indicators. These indices were then aggregated to GDSI using the PCA approach. Results have indicated that the GDSI reached its maximum in 2013 and its minimum in 1997. Further, the GDSI model has provided us with some key points in terms of the impact of each individual indicator on Iran's gas demand security status. Considering the contribution of each indicator in determining the GDSI model, we suggested some policies that the energy policymakers may adopt to promote Iran's gas demand security level.

Our study, being of an exploratory nature, raises a number of opportunities for future research. First, new research may address more indicators such as: stage of economic development, energy expenditures, price & market volatility, climate or environment factors, technology innovation and adoption, and develop similar composite index for other energy sources, specifically the crude oil. Further, from the international perspective, it may be an interesting topic to develop a composite index for measurement of the energy demand security in energy exporting countries/ regions.

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## Appendix

**Table 7. Research Data (Source: OPEC and World Bank)**

Year	Iran's Total Export Value (million USD)	Iran's Gas Export Value (million USD)	Iran's Gas Export Volume (mcm)	Iran's GDP (million USD)	Azerbaijan's GDP (million USD)	Armenia's GDP (million USD)	Turkey's GDP (million USD)
1993	18080	321/33	500	73766	3973	1201	180170
1994	19434	344/41	420	105878	3314	1315	130690
1995	18360	367/48	100	104763	3052	1468	169486
1996	22391	390/56	100	107455	3177	1597	181476
1997	18381	413/63	0/00	100877	3963	1639	189835
1998	13118	436/70	0/00	117861	4446	1894	269287
1999	21030	459/78	275/00	104656	4581	1845	249751
2000	28345	482/85	350/00	96440	5273	1912	266568
2001	23904	505/93	358	115435	5708	2118	196005
2002	28186	529/00	670	115234	6236	2376	232535
2003	33991	552/07	3520	132251	7276	2807	303005
2004	44403	633/95	3560	158828	8681	3577	392166
2005	64366	1145/32	4320	187438	13245	4900	482980
2006	76190	1456/95	5690	222128	20983	6384	530900
2007	97668	1514/74	6160	307355	33050	9206	647155
2008	101289	1462/30	5800	350588	48852	11662	730337
2009	88326	1061/11	5670	360625	44291	8648	614554
2010	112788	1733/35	8420	463971	52903	9260	731168
2011	144874	2231/20	9100	564459	65952	10142	774754
2012	107409	2111/05	8400	583164	68731	10619	788863
2013	91793	2362/84	9400	380348	73560	11121	823243
2014	86235	1940/49	9600	416490	75198	11644	798797
2015	77974	1375/36	8400	387611	53047	10561	718221

mcm: million cubic meters

**Table 8. Research Data. Continued (Source: PRS and, BP)**

Year	Iran's ICRG	Azerbaijan's ICRG	Armenia's ICRG	Turkey's ICRG	Volume of gas exported to Turkey (mcm)	Volume of gas exported to Azerbaijan (mcm)	Volume of gas exported to Armenia (mcm)
1993	66/56	53/42	50/10	65/22	500/00	0/00	0/00
1994	65/71	53/12	49/98	64/54	100	0/00	0/00
1995	64/85	52/82	49/85	63/87	100/00	0/00	0/00
1996	70/58	52/52	49/73	56/44	100/00	0/00	0/00
1997	61/77	52/23	49/61	64/02	0/00	0/00	0/00
1998	64/84	52/53	53/35	48/36	0/00	0/00	0/00
1999	59/4	52	48/53	65/99	244/00	0/00	0/00
2000	62/31	52/4	44/76	61/24	301/00	0/00	0/00
2001	57/14	52	49/37	64/74	358/00	0/00	0/00
2002	54/77	51/83	48/9	52/53	670/00	0/00	0/00
2003	54/77	50/19	48/14	60/42	3520/00	0/00	0/00
2004	54/64	51/33	47/57	63/54	3560/00	0/00	0/00
2005	57/67	50/6	48/45	62/59	4320/00	0/00	0/00
2006	57/67	50/79	49/78	59/56	5690/00	0/00	0/00
2007	56/53	50/41	50/47	57/99	6160/00	0/00	0/00
2008	55/02	49/91	49/08	56/66	5800/00	0/00	0/00
2009	52/81	50/47	49/27	56/5	5250/00	420/00	0/00
2010	50/09	50/28	48/71	53/03	7770/00	250/00	400/00
2011	47/63	50/09	47/76	52/34	8400/00	400/00	300/00
2012	46/94	47/82	47/19	52/53	7500/00	540/00	360/00
2013	48/58	47/13	47/19	49/18	8700/00	420/00	280/00
2014	50/47	47/13	47/06	47/92	8900/00	450/00	250/00
2015	54/64	42/77	47/25	47/73	7800/00	360/00	240/00

mcm: million cubic meters