

Dynamic Sustainable Success Prediction Model for Infrastructure Projects: A Rough Set Based Fuzzy Inference System

Abstract

Purpose

Successful implementation of infrastructure projects has been a controversial issue in recent years, particularly in developing countries. In this study, an attempt is made to propose a Decision Support System (DSS) for the evaluation and prediction of project success while considering sustainability criteria.

Design/methodology/approach

To predict sustainable success factor, the study first developed its Sustainable Success Factors (SSF) and Sustainable Success Criteria (SSC). These then formed a decision table. A Rough Set Theory (RST) was then implemented for rules generation. The decision table was used as the input for the rough set, which returned a set of rules as the output. The generated rule sets were then filtered in Fuzzy Inference System (FIS), before serving as the basis for the DSS. The developed prediction tool was tested and validated by applying data from a real infrastructure project.

Findings

The results show that the developed rough set -fuzzy method has strong ability in evaluation and prediction of the project success. Hence, the efficacy of the DSS is greatly related to the rule-based system, which applies RST to generate the rules and the result of the FIS was found to be valid via running a case study.

Originality/value

Use of DSS for predicting the sustainable success of the construction projects is gaining progressive interest. Integration of RST and FIS has also been advocated by the seminal literature in terms of developing robust rulesets for impeccable prediction. However, there is no preceding study adopting this integration for predicting project success from the sustainability perspective. The developed system in this study can serve as a tool to assist the

decision-makers to dynamically evaluate and predict the success of their own projects based on different sustainability criteria throughout the project life cycle.

Key Words: Infrastructure projects, rough set theory, sustainability, project success, the fuzzy inference system

1. Introduction

Over the last decade, there has been a growing emphasis on the successful implementation of construction projects ([Pour Rahimian et al., 2019](#); [Pour Rahimian et al., 2020](#); [Zarghami et al., 2018](#)). The primary challenge has always been the ambiguities associated with assessing and anticipating success on such projects, since predicting several probable issues from a vast set of data is more problematic ([Cheng et al., 2013](#)). However, there is still no consensus between project management researchers as to how project success is measured or predicted ([Elbarkouky, 2012a](#); [Pinto and Slevin, 1988](#)) since the critical characteristic of construction projects is their unpredictability in comparison with rather static production industries ([Loosemore et al., 2003](#)).

Traditional approaches addressing project success are restricted to deliver projects on cost, time and quality. Since construction investment mainly focuses on economic benefits (and these kinds of projects have mostly detrimental impacts on social and environmental dimensions) considering sustainable development would be a necessity in presenting a comprehensive success prediction model ([Kibert, 2016](#); [Kolo et al., 2014](#)), despite the shortage of relevant studies within the extant literature. On the other hand, accomplishing construction projects successfully requires continuous monitoring and control by construction managers. The project manager and project team have to make judgment during project execution about whether the project will be successful or not ([Ding and Banihashemi, 2017](#)).

Therefore, this study aimed at designing a dynamic Decision Support System (DSS) that is capable of prediction of project success from the sustainability perspective throughout the project lifecycle. The

developed DSS model employs Rough Set Theory (RST) ([Pawlak, 1982](#)) in conjunction with the Fuzzy Inference System (FIS) to help the project develop robust rulesets for impeccable prediction.

The remainder of the paper is organised as bellow. The following section, “Literature Review”, investigates the concepts of success in construction projects. Research Methodology section explains and presents data collection and demonstrates the review of RST and fuzzy logic. The Results and Analysis section outlines how the rough-fuzzy based prediction model was developed, tested and validated through an illustrative example. The article concludes with practical recommendations to use of this model in the future works.

2. Literature review

2.1. Project Success

The definition of project success has undergone some transformations over the years. Traditionally, a construction project was deemed successful when it met criteria related to time, cost, and quality ([Atkinson, 1999](#)). As pointed out by [Belassi and Tukel \(1996\)](#), research on project success needs to distinguish success factors and success criteria. [Cooke-Davies \(2002\)](#) highlighted the difference between the success criteria and success factors. Project success factors or Critical Success Factors (CSFs) are independent variables of a project, which contribute to achieving success in a project ([Müller and Turner, 2007](#); [Rockart, 1982](#)).

Project success criteria are dependent variables by which the success or failure of a project will be judged and measured by its stakeholders ([Belassi and Tukel, 1996](#); [Pinto and Slevin, 1988](#); [Rockart, 1982](#)). Factors constituting the success criteria are commonly referred to as the key performance indicators (KPIs). The difference between KPIs and CSFs needs to be taken into consideration. [Cox et al. \(2003\)](#) argued that CSFs are the efforts made – or strategies adopted – to achieve the success of a project.

Whereas, KPIs are the compilations of data measures used to assess the performance of the construction project. In other words, the KPIs are essential for comparing effectiveness, efficiency and quality of the actual and estimated performances in of both workmanship and product.

Literature regarding the area of project success reveals that various authors have identified many success determinants, either from experience or research. [Westerveld \(2003\)](#) revealed that along with the conventional measures of cost, time, quality, and scope, there are other five KPIs that are used most frequently: 1) client's appreciation, 2) project personnel appreciation, 3) users' appreciation, 4) contracting partners' appreciation and 5) stakeholders' appreciation. [Belout and Gauvreau \(2004\)](#) emphasised on the project team's ability to manage project risks and to resolve the problems encountered on the project to evaluate the project success. In a study by [Cserhati and Szabo \(2014\)](#), analysis of correlations revealed that relationship-oriented success factors, such as communication, co-operation and project leadership, play a vital role in the successful implementation of projects.

Infrastructure projects require large budgets and a prolonged schedule, and they involve many complicated procedures. Several attempts have been made to address project success in these projects. [Brundtland et al. \(1987\)](#) identified success factors for large projects using factor analysis method. These factors were grouped into four major categories: 1) incompetent designers and contractors, 2) poor estimation and change management, 3) social and technological issues, and 4) improper techniques and tools. In another research, [Ogunlana \(2010\)](#) argued that these traditional criteria for success were not sufficient to determine whether or not the project was successful, and quantitative as well as qualitative criteria such as environmental regulations, building performance, and client satisfaction should also be considered.

2.2. Sustainable success indicators

[Al-Tmeemy et al. \(2011\)](#), measured the success of building projects for sustainable social housing in Nigeria. They identified several critical success factors influencing sustainable housing. In a more recent study, [Krajangsri and Pongpeng \(2016\)](#) addressed the impacts of sustainable infrastructure assessments on construction project success using Structural Equation Modelling (SEM). In their research, construction project success was measured using six criteria: time, cost, quality, client satisfaction, safety, and environment. [Banihashemi et al. \(2017\)](#) looked at the CSFs affecting the integration of sustainability into project management practices of construction projects in developing countries. Having innovation diffusion theory as the theoretical point of departure, they identified 59 CSFs pertaining to the triple bottom line of sustainability (environmental, social and economic). A number of most related studies with different perspectives toward success criteria and success factors are summarised in Table1 and Table 2.

Table 1. Summary of available previous studies on sustainable success factors

Table 2. Summary of available previous studies on sustainable success criteria

2.3. Project success prediction

In recent years, there have been many studies proposing different prediction models for project success. [Kim et al. \(2009\)](#) developed an SEM to predict the project success of uncertain international construction projects. In their study, through a comparative analysis of SEM with multiple regression analysis and artificial neural network, SEM showed a more accurate prediction of performance. Moreover, [Cheng et al. \(2010\)](#) proposed an Evolutionary Support Vector Machine Inference Model (ESIM) to the dynamic prediction of project success. Their model integrated the process of continuous assessment of project performance (CAPP) to select factors that influence project success dynamically. Furthermore, [Cheng et al. \(2013\)](#) proposed an evolutionary Gaussian process inference model for the dynamic success

prediction. They built an Evolutionary Gaussian Process Inference Model (EGPIM), using a Gaussian process, along with Bayesian inference and particle swarm optimisation. The model was trained using the EGPIM and proved quite precise at predicting the success of a project and had exceptional performance in time-series applications. Besides, one of the outstanding models presented to predict construction cost and schedule success using artificial neural networks and support vector machines classification proposed by [Wang et al. \(2012\)](#). Their findings showed that early planning status can be efficiently adapted to predict project success and the proposed Artificial Intelligence (AI) models can deliver acceptable prediction results.

From a slightly different point of view, [Khosravi and Afshari \(2011\)](#) developed a success measurement model for construction projects to establish a benchmark for measuring future construction projects and comparing the success of finished projects. The implemented methodology was a two-round Delphi study complemented with a questionnaire survey. [Khang and Moe \(2008\)](#) also presented a new conceptual framework for international development project success. They delineated a link between success factors and criteria in order to base their model for assessing the project status and forecasting the results of project throughout the life cycle. [Abd-Hamid et al. \(2015\)](#) also identified success predictors, and presented a conceptual framework for formulating the success of entrepreneurs in the construction industry. [Kim et al. \(2009\)](#) developed an analytic network process model to predict the Performance of International Construction Joint Ventures (ICJV). Within their study, the relationship between the important determinants of ICJV success was found. Eventually, they tested the model based on the eight real construction projects and obtained satisfactory results.

3. Research methodology

3.1. Sustainable success measurement

In developing sustainable success measurement criteria, this study followed [Akbari et al. \(2018\)](#) who established a list of sustainable success indicators, namely Sustainable Success Factors (SSF) and Sustainable Success Criteria (SSC). They designed a questionnaire survey using 20 SSFs and 15 SSCs, being classified into the triple bottom line of economic, social and environmental groups. Finally, they selected 11 factors and 9 criteria from these three groups, as mentioned in the following paragraphs.

The identified SSFs: Time management (F1), Cost management (F2), Quality management (F3), Risk management (F5), Leadership (F12), Competent project team (F13), Motivation (F14), Teamwork (F10), Attempt to preserve environment /environmental protection (F15), Waste management (F16) and Environment protection measures in project design (F18).

The selected SSCs: Project completion within time (C1), Project completion within budget (C2), Project quality (C3), Employer satisfaction/Client satisfaction/owner satisfaction (C6), Provision of employment opportunities (C8), Overall health and safety measures (C9), Environmental degradation (C11), Energy consumption (C15) and Effect on air and land pollution (C13).

Next, to identifying sustainable success indicators, the AHP approach was utilised to quantify and calculate success based on the data collected from 20 experts. With respect to the weight factors, the success indices of the three categories of sustainability were calculated using the following equations:

$$ECSI = 0.31 C3 + 0.38C1 + 0.31C2 \quad (1)$$

$$SOSI = 0.43C11 + 0.34C15 + 0.23C13 \quad (2)$$

$$ENSI = 0.48C6 + 0.37C9 + 0.15C8 \quad (3)$$

Where:

ECSI=Economical Success Index, SOSI=Social Success Index, ENSI=Environmental Success Index. It is noteworthy that the economic category is the most important one, with 62% of the weighting. It is two to four times greater than that of the social category (24%) and environmental category (14%), respectively.

Therefore, it was concluded that most attention should be paid to the economic point of view, whilst the social and environmental aspects should be taken into account as well. Moreover, sustainable success index is calculated as follows:

$$SSI=0.62 ECEI+0.24 SOSI+0.14 ENSI \quad (4)$$

Where: SSI=Sustainable Success Index

Subsequently, respondents were asked to give a score based on the status of the selected success factors and criteria in the aforementioned projects which they were responsible to them. These numbers were between 1 and 5 where: 1 represents “very bad”, 2 – “Slightly bad”, 3 – “Moderate”, 4 – “good”, and 5 – “excellent”. Finally, based on Equ.4, the sustainable success index of these projects had been calculated and a decision table used in RST was created.

3.2. Methodological Review

Recently, AI has been deployed as a powerful tool to support decision making in construction projects ([Martínez Magaña and Fernández-Rodríguez, 2015](#)). The machine learning techniques such as have been used for modelling building energy-related neural networks, support vector machine and decision trees ([Seyedzadeh et al., 2019](#)). These accurate and fast predictors enable exploration of a vast combination of building characteristics both in the design stage and retrofit planning ([Seyedzadeh et al., 2018](#)). To deal with the uncertainty in building and construction data (and weather) rough set was coupled with machine

learning techniques ([Shi and Li, 2008](#)). Fuzzy logic another AI method has also received significant attention in this area due to handling subjective arising during the construction process ([Lam et al., 2007](#)).

Fuzzy logic provides an accurate approach for dealing with vagueness. Their rule-based nature allows the use of information expressed in the form of natural language statements. Fuzzy inference is an expert system that interprets the values in the input vector, and based on some set of if-then rules, assigns values to the output vector.

RST does not need a membership function and compared to other statistical methods, and it does not require a large number of data points to find a rule ([Pheng and Hongbin, 2006](#)). The most important advantage of this theory is its capability of estimating the significance of specific attributes ([Liu and Yu, 2009](#)). The fundamental concept of the rough set algorithm for the proposed application is described in the following paragraphs.

Definition 1: Information Systems

Information systems are the set of objects described by their attributes and attribute values. The information system is defined as follows:

$$IS = (U, A) \tag{5}$$

Where U is the universe, a finite non-empty set of objects, $U = \{x_1, x_2, \dots, x_m\}$, and A is the set of attributes. Each attribute $a \in A$ (attribute a , belonging to the considered set of attributes A) defines an information function:

$$f_a : U \rightarrow V_a \tag{6}$$

Where V_a is the set of values of a , called the domain of attribute a . In all attributes, there are decision attributes and condition attributes.

Definition 2: Core and reduct of attributes

The concepts of core and reduct are two fundamental concepts of the rough sets theory. A reduct is the minimal subset of attributes that enables the same classification of elements of the universe as the entire set of attributes. In other words, properties that do not belong to a reduct are superfluous with regard to classification of elements of the universe. The core is the necessary element for rules and is the common portion of all reducts. Let B be a subset of A. The core of B is the set of all indispensable attributes of B. The following is an important property, linking the concept of the core and reducts:

$$Core(B) = \bigcap Red(B) \tag{7}$$

Where Red (B) is the set of all reducts of B.

The significance of an attribute can be measured by comparing the degree of partial dependency (γ) of a set, which includes the attribute with the degree of a set without the attribute. This idea can be formally described as follows:

$$\sigma_{(C,D)}(a) = \frac{(\gamma(C,D) - \gamma(C - \{a\}, D))}{\gamma(C,D)} \tag{8}$$

Where $a \in C$, and $\sigma(a)$ is the significance of attribute a ($0 \leq \sigma(a) \leq 1$). The significance of a set of attributes can be calculated in the same way as follows:

$$\sigma(a) = \frac{(\gamma(C,D) - \gamma(C - B, D))}{\gamma(C,D)} = 1 - \frac{\gamma(C - B, D)}{\gamma(C,D)} \tag{9}$$

Where B is a subset of C. The significance of a set B, i.e., $\sigma(B)$, represents the effect of eliminating the set. Thus, the set of decision attributes D will not be properly classified to the same extent as the degree of $\sigma(B)$ when taking out set B from C, the set of condition attributes. Thus, we can determine an approximate reduct, the best subset for explaining a decision, by determining the significance of all possible sets.

3.3. Attribute reduction and Rule generation using RST

As described earlier, a unique feature of the RST method is its generation of rules, which has great importance in the prediction of the outputs. For this purpose, the Rosetta toolkit was employed to induce rough based models. Rosetta tool lists the rules and provides some statistics for filtering the rules. Basic concepts of the RST and its supremacy for rule generation has been thoroughly discussed in the methodology review.

As described earlier, RST is used to identify the most significant features by computing subsets and cores. In order to generate reducts, Genetic Algorithm (GA) is applied as it provides a more exhaustive exploration of the search space ([Wroblewski, 1995](#)). Reducts generation has two options: full object reduction and object-related reduction. This reduction produces a set of decision rules through minimal attributes subset that distinguishes on a per-object basis, whilst reduct with full object reduction creates set of minimal attributes subset that designates functional dependencies ([Sulaiman et al., 2008](#)).

In this study, a full object reduction approach is adopted. Therefore, the reducts which are used to generate rules in Economic, social and Environmental categories are [F15, F16, F18], [F12, F10, F13, F14], [F15, F16, F18] respectively. A unique feature of the RS method is its generation of rules, which is of great importance in the prediction of the outputs. For this purpose, the Rosetta system for inducing rough based models was applied. Rosetta tool lists the rules and provides some statistics for the rules which are support, accuracy, coverage, stability and length. This study adopted its definition of the rule statistics adopted from [Sulaiman et al. \(2008\)](#).

The number of primary decision rules generated based on produced reducts in Economic, Social and Environmental categories are 17,18,16 respectively. To raise effectiveness, this study filters the decision rules according to the principle LHS support ≥ 2 . Using Rosetta, rules with the highest LHS Support in all sustainable groups are extracted. These rules are sorted based on LHS Support in Table 3. It is worth

mentioning that the LHS Support indicates the number of projects satisfying the condition of the rule while the RHS Support indicates the number of projects satisfying the decision of the rule. In this study, based on the set of generated rules, the projects were classified into four categories.

Table 3. Filtered rules extracted from Rosetta software

4. Result and Analysis

4.1. Explanation of proposed rough-fuzzy based model:

This section demonstrates the development of a success prediction model using Mamdani and Takagi–Sugeno fuzzy inference expert systems. A set of the rules obtained from RST was utilised to implement the proposed DSS. The description of the proposed model is presented through two stages as shown in Fig. 1. The first step was to generate the ‘if-then’ rules deriving from Table 3 and define membership functions, then build the fuzzy inference engine. In order to implement this model, in the first stage, decision-makers need to give a score to inputs in each sustainability categories of economic, social and environmental. The output of this stage is success scores defuzzified to obtain crisp values. There are three output variables in the first stage, which are considered as input variables in the second stage. In the second stage, these three input variables represent the status of economic, environmental, and social success. The relative importance weightage of each category is used to obtain the sustainable success index in total. For this purpose, to design this DSS, the basic concepts of the FIS are discussed.

Fig. 1. The description of the proposed model through two stages

4.1.1. Fuzzy membership functions in the proposed model:

In this study, membership functions are applied in the trapezoidal and triangular forms. A trapezoidal membership function is defined as Eq. (10). According to Eq. (10), if $b = c$ then the number is called a triangular fuzzy number.

$$\mu_{\tilde{A}}(x) = \begin{cases} 0 & x < a \\ \frac{1}{b-a}(x-a) & a \leq x \leq b \\ \frac{1}{c-d}(x-d) & c \leq x \leq d \\ 0 & x > d \end{cases} \quad (10)$$

In the first stage of the model, five fuzzy sets of membership functions are applied for both inputs and outputs of the FIS system. The fuzzy sets in the form of linguistic variables for inputs of stage 1 include “Very Bad”, “Slightly Bad”, “Moderate”, “Good” and “Excellent”. These variables are equivalent to fuzzy numbers on a numeric range of 1–5 as shown in Table 4. As for outputs of this stage, linguistic variables are considered as “very unsuccessful”, “unsuccessful”, “moderately successful”, “very successful” and “Extremely success”. These variables are equivalent to fuzzy numbers on the numeric range of 0–100 (see Table 4).

Table 4. The linguistic variables for inputs and outputs at the first stage

The corresponding triangular membership functions for the inputs and outputs of stage 1 are shown in Fig. 2. The DSS in this stage is developed using Mamdani’s FIS. Whereas, the FIS in the second stage is conducted based on the Takagi- Sugeno method. For the sake of simplicity, in this stage, this study only applied three fuzzy sets of membership functions for inputs of the FIS. However, a higher number of qualifiers can be used in the conceptual model to provide a better assessment. The fuzzy input sets in this

stage are in the form of linguistic variables including “Unsuccessful”, “Moderate”, “Successful” which are used to evaluate the sustainable success in the second stage. The corresponding fuzzy numbers of these fuzzy sets are presented in Table 4. As shown in Table 4, linguistic variables are equivalent to fuzzy numbers on a scale of 0-100. The corresponding trapezoidal membership function for the inputs in this stage is shown in Fig. 2. Yet, for output in this stage, the linear membership function of Sugeno-type is applied. The related fuzzy number for output of this stage is obtained using Equation (10). In the current study, the coefficients of the output membership functions of the designed FIS are obtained using AHP method. As there are three inputs in this stage, the output would be:

$$y=a_1x_1+a_2x_2+a_3x_3 \tag{11}$$

where a_1, a_2, a_3 denote weights resulted from the pairwise comparison matrices of the sustainability categories detailed in Section 3.1. Finally, Fuzzy rule based-matrix in the second stage is designed as shown in Table 5.

Fig. 2. The membership functions for inputs and outputs in the first stage and outputs in the second stage

Table 5. The fuzzy rule-based matrix in the second stage

4.1.2. Defuzzification and Calculation of the output in each stage:

Defuzzification refers to the way in which the fuzzy number is converted into a crisp value. In this study, in the first stage, in order to calculate the project success in each sustainability categories centre of area method (COM) is used, as shown in Eq. (12)

$$X_{COM} = \frac{\sum_{i=1}^n x_i \cdot \mu_i(x_i)}{\sum_{i=1}^n \mu_i(x_i)} \quad (12)$$

In Sugeno FIS, the conclusion of a fuzzy rule is constituted by a weighted linear combination of the crisp inputs rather than a fuzzy set. Besides, in this study, in the second stage, weighted average defuzzification method is used to calculate the output of DSS, which is the sustainability success index.

5. Case Study and Model Validation

A case study (Fig. 3) was chosen to test the validity of the model is an urban tunnel project in Iran-Tehran, including two lines over 3993 m of length. The budget for the project is roughly USD\$64 million. The project commenced in 2014, and it is projected to be completed by the end of 2019. The illustrative example has been utilised for this large-scaled project (This project hereafter will be mentioned as B).

A three-member assessment team was constituted by project managers who were also the head of their departments in this project. Decision makers were asked for their perception towards status of success factors in this project as shown in Table 6. Besides, two virtual projects were defined as the extremely successful (A) and unsuccessful (C). Finally, the sustainable success index of these projects are computed applying this model.

Fig. 3. Case study details

Table 6. Decision makers' perception toward status of success factors in project B

During the implementation and result extraction, inputs' values from the data collection process are transferred to the FIS system. For visualisation of the structure of the model, the rule viewer of the second stage of the project B is shown in Fig. 4. In the rule viewer, input values can be altered by moving

the red line, and the output can be observed along the column indicating the output variable. In the second stage of the model, there are three input variables (Economical, Social and Environmental) and three membership functions. Therefore, the rule base consisted of 27 (3^3) if-then rules. To verify these rules, the inputs were increased and output was examined. Inputs and outputs were in the range of [0–100]. The output surface of the second stage of FIS for the sustainable success index on the basis of Economical and Environmental success is shown in Fig. 5. It is seen that the more the input values increase, the more output value (sustainable success index) surges. Besides, as linear Takagi-Sugeno is applied in this stage, it is seen that inputs vary linearly with the output.

The validity of the proposed model was proven by obtaining the sustainable success index of the project B, which always varies between sustainable success index of projects A and C. This model was also tested by applying different Defuzzification methods to show the validity of the model. Different approaches such as the centre of area method (COM), the bisector of area method (BOM), mean of the maximum method (MOM), smallest of the maximum method (SOM), and largest of the maximum method (LOM) were applied. As can be seen in Table 7, assessment results for A, B and C have the same trend in all the Defuzzification methods, and this proves the validity of this model as well. Besides, obtained final value of stage 2 (70.5) gives exactly a sustainable success index calculated using Equ.4.

Table 7. Validation of the proposed model at the first stage

Fig.4. Rule viewer of the FIS for project B

Fig. 5. The output surface of the FIS for project B

5.1 Prediction of Sustainable success Index based on scenario analysis

This section addresses different scenarios for evaluating the prediction of the sustainable success index. It is assumed that each of the 27 rules in stage 2 can be considered as a potential scenario in case three numbers in each membership function range are considered as inputs. These numbers were determined 20 for unsuccessful range, 50 for moderate range and 80 for successful range. Hence the sustainable success status of projects can be predicted in these developing scenarios. For example, in the case of scenario 27, all parameter values are excellent, resulting in the highest score of 80. By reducing the economic values to the lowest value of itself (20), the SSI score is reduced to 42.8 (scenario 9), notwithstanding other parameters are maximum. Table 8 shows these scenarios.

Table 8. Prediction of SSI based on Scenario analysis

5.2 Sensitivity Analysis Using Fuzzy Inference System

To evaluate the impact of each of the economic, social and environmental areas on the sustainable success index, sensitivity analysis has been performed.

For the sensitivity analysis, two areas were chosen under the identical condition and the impact of the third area on the SSI was evaluated. The results of the scenarios are presented in figure 6.

Fig.6. sensitivity analysis based on different scenarios in sustainability areas.

In this figure, the X-axis shows the sustainability areas and the Y-axis represents the sustainable success index. According to the figure, the equation related to each graph with a different slope shows the link between sustainability areas and the sustainable success index. There is a striking difference in the coefficients of the equations. As can be expected among the Sustainability areas, the economic area has the most impact on the sustainable success index compared to environmental and social areas.

Based on the abovementioned scenarios we compared the sustainability areas. The Bar chart (fig.7) proves the abovementioned result "special attention should be paid to the economic point of view, whilst the social and environmental aspects should be taken into account as well".

Fig.7. Sustainability areas score

6. Conclusion

Infrastructure projects make a significant contribution to economic growth, social development and environmental activities, especially in developing countries, like Iran. With respect to great resource consumption and heavy investment of this kind of projects, the success of them on the basis of sustainability principles should be properly predicted before implementation as well as dynamically evaluated during the project life cycle. These measures will secure the achievement of project success. Therefore, this study was undertaken with the aim to design a DSS applying RST and FIS to predict the success of projects from the sustainability perspective.

In the present study, initially, RST was implemented for rules generation as explained in Section 3.4. The input of the rough set is a decision table, including success factors and criteria, and the output is a set of rules. Filtered rules served as the basis for the DSS. Drawing on the results and discussions, it can be inferred that rough set -fuzzy method demonstrated strong capability for evaluation and prediction of the project success. Hence, the developed model puts forward an opportunity for further improvements in achieving more sustainable success through its application.

The analysis results derived from the rough-fuzzy model suggests that economic success is the triggering fact to strengthen the social and environmental dimensions. Similarly, the social aspect needs to be taken into account as much as the environmental aspect to enhance the overall sustainable success. The result of this model can give an image of the present situation to the project managers and help them to maintain a balance among economic, social, and environmental performance for successful implementation of construction projects throughout the project life cycle.

This research was limited to infrastructure projects with a minimum budget of around \$30 million. Hence, this model should be treated with caution if being applied for smaller size projects. Further researches should be conducted in different contexts and other countries by focusing on the particular

infrastructure project types for comparative purposes. Moreover, as large projects have many stakeholders who have different objectives, it is recommended that the prediction of success from different stakeholders' view is studied separately. The proposed technique calls for further studies applying different approaches such as ANFIS to achieve better results in comparison with the proposed system. In this paper, the sufficiency of the rules has been somehow validated. However, it is highly recommended that potential authors repeat such prediction by implementing more rules.

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Figures

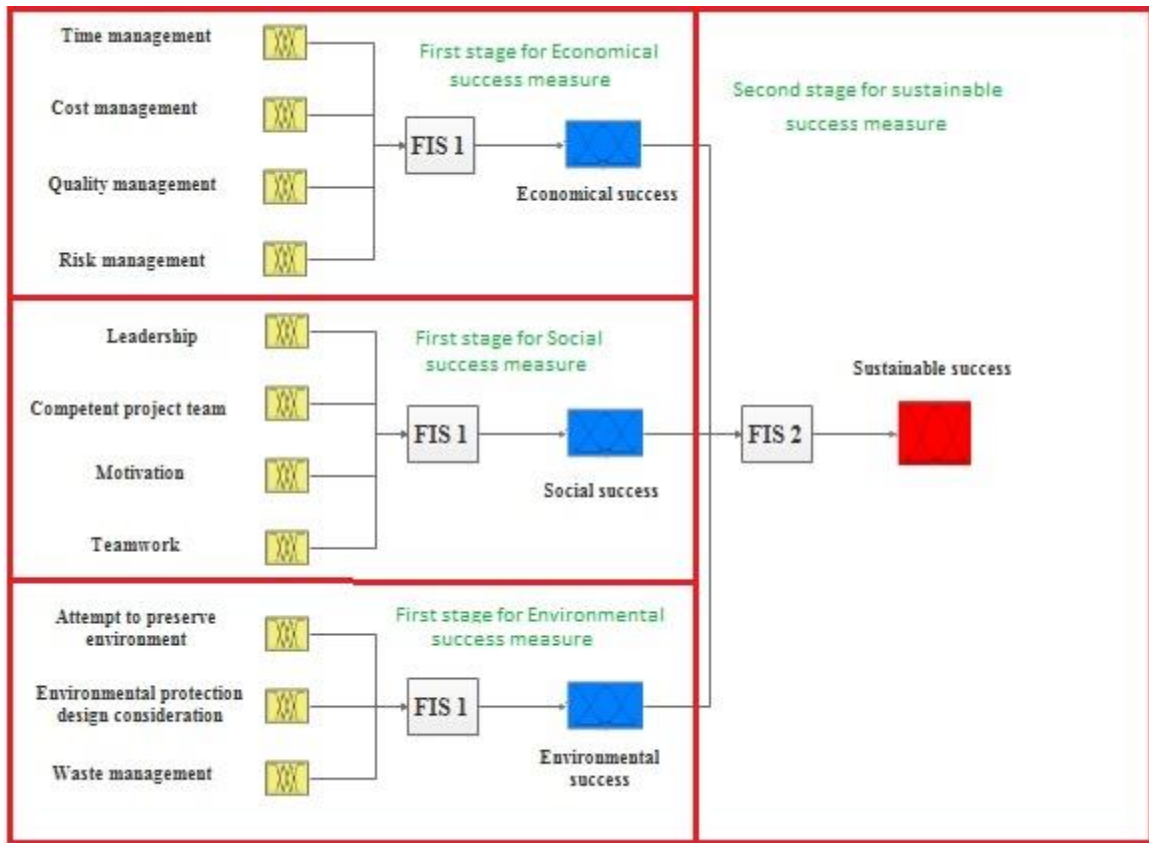
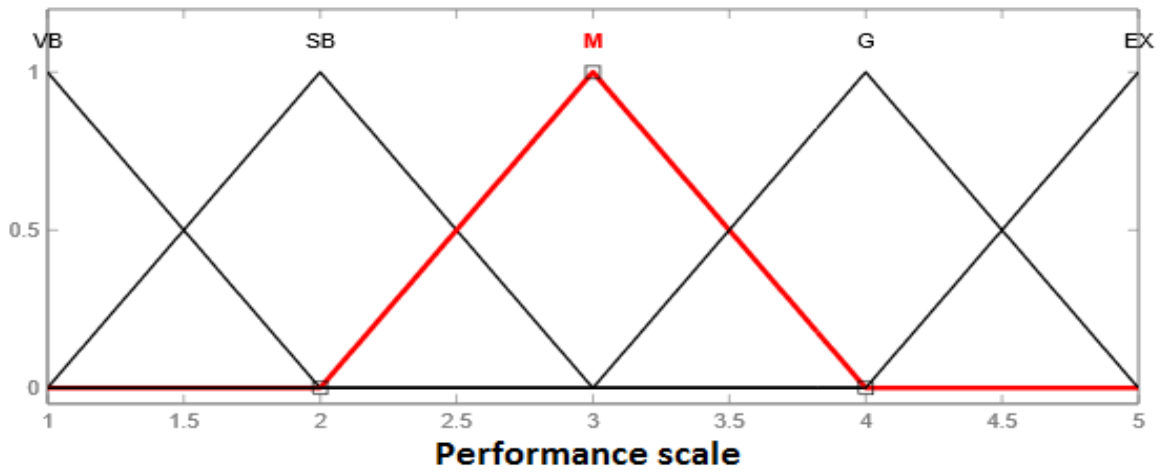
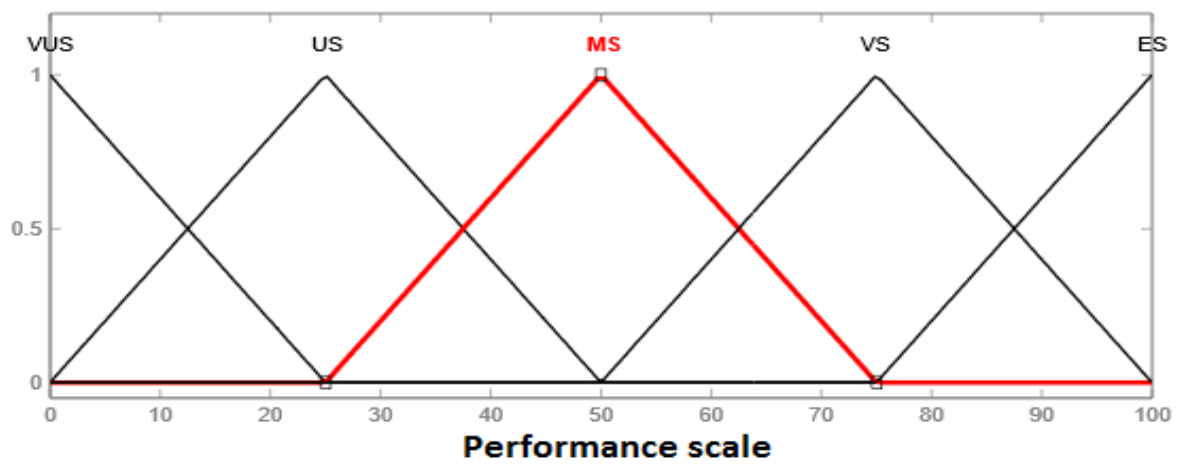


Fig.1. The description of proposed model through two stages

Degree of Membership, Inputs Stage 1



Degree of Membership, Outputs Stage 1



Degree of Membership, Outputs Stage 2

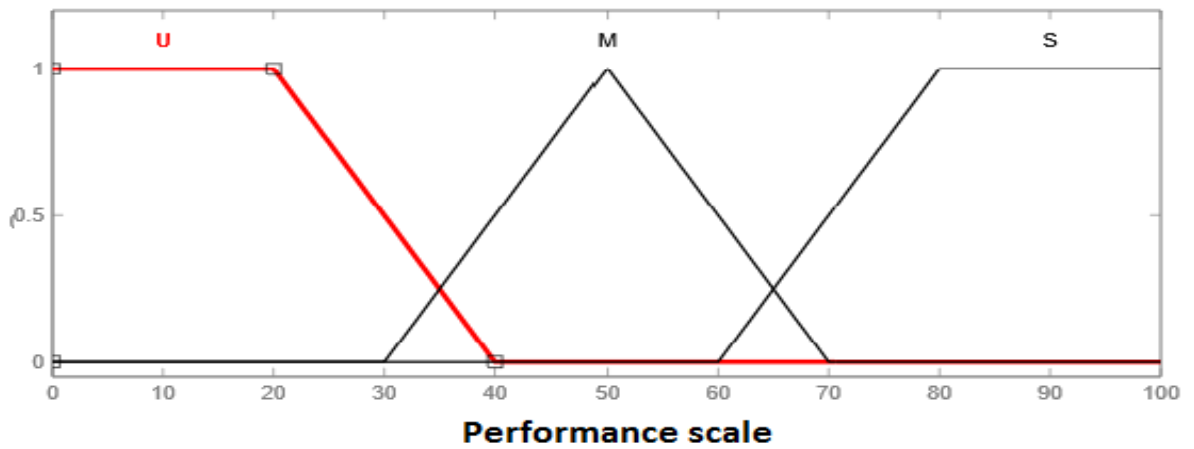
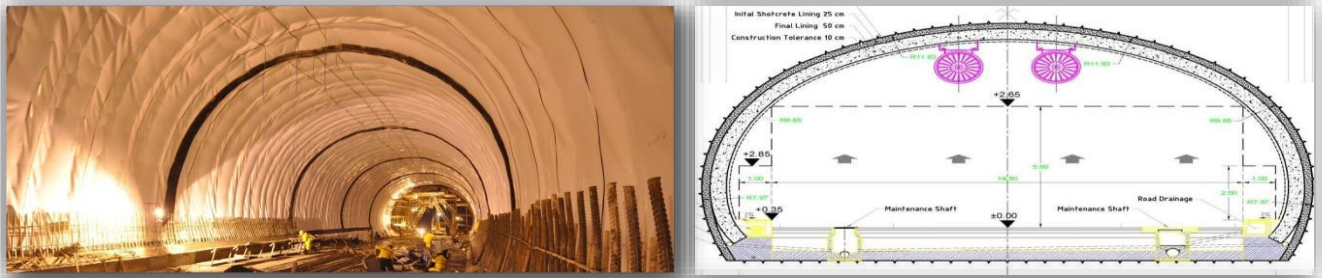


Fig. 2. The membership functions for inputs and outputs in the first stage and outputs in the second stage



Longitudinal profile of the Tunnel

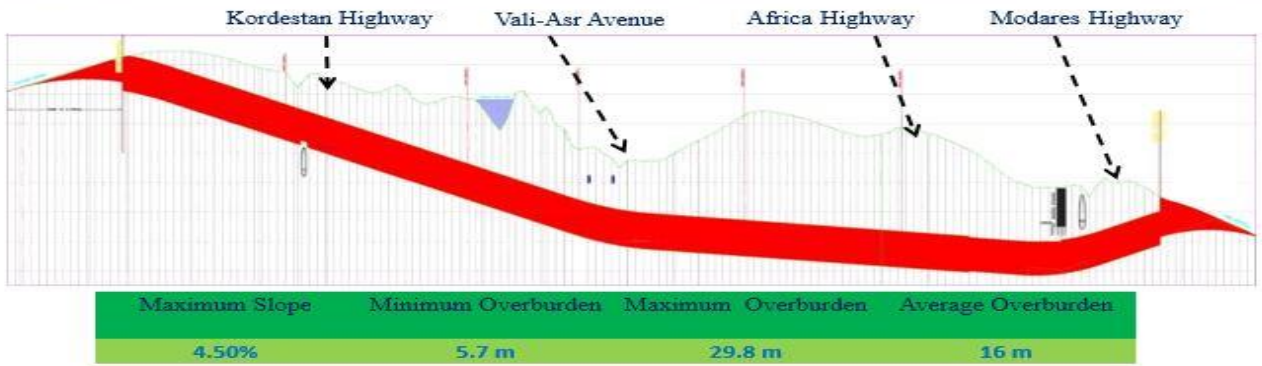


Fig. 3. Case study details

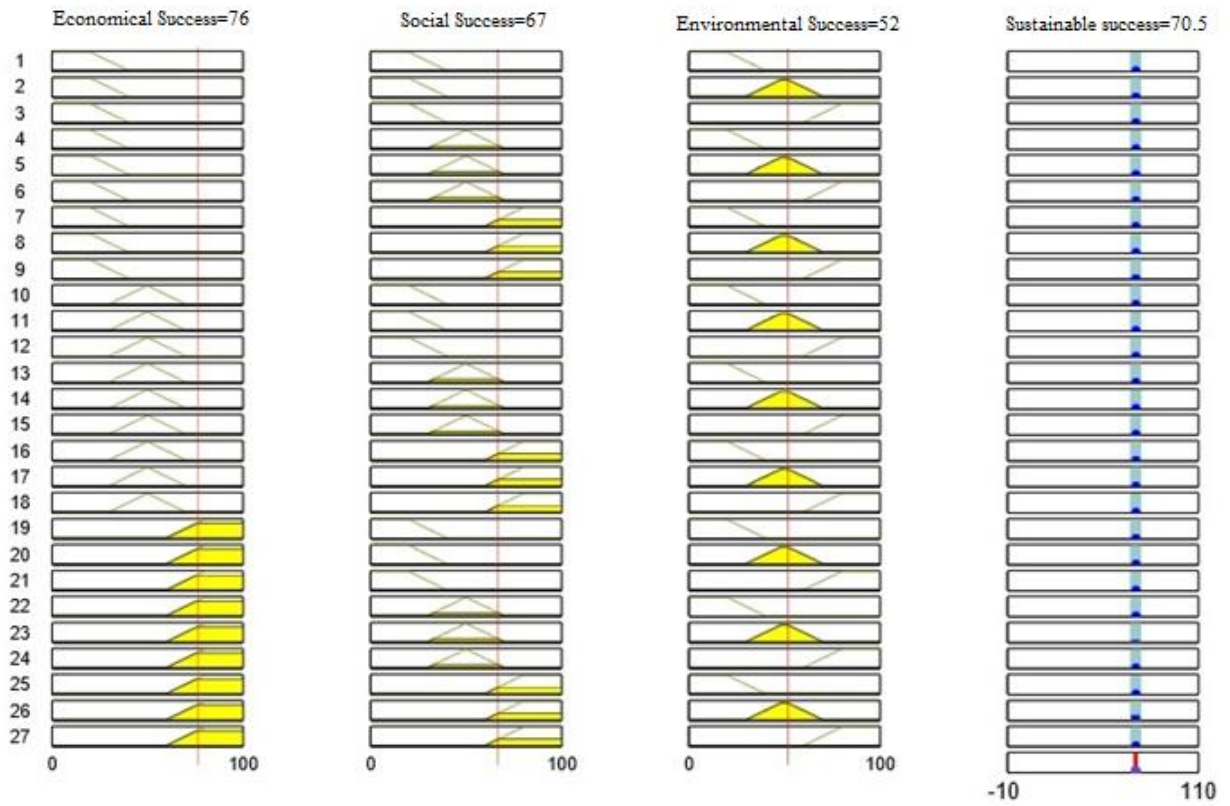


Fig. 4. Rule viewer of the FIS for the project B

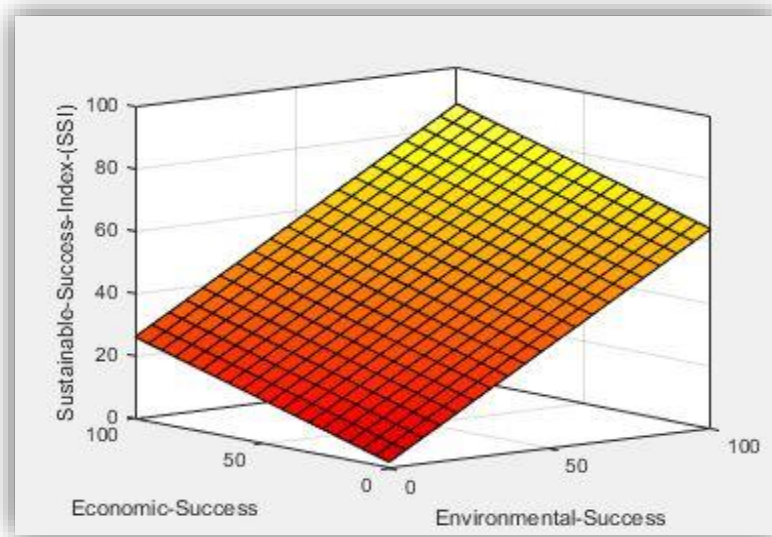


Fig. 5. The output surface of the FIS for the project B

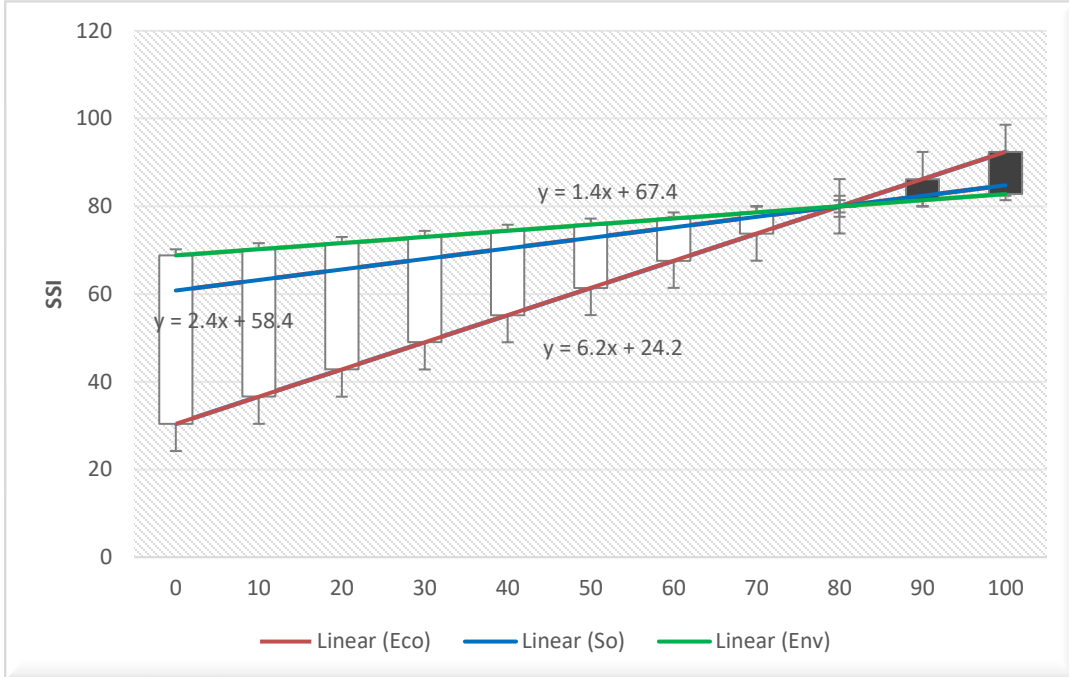


Fig.6. sensitivity analysis based on different scenarios in sustainability areas.

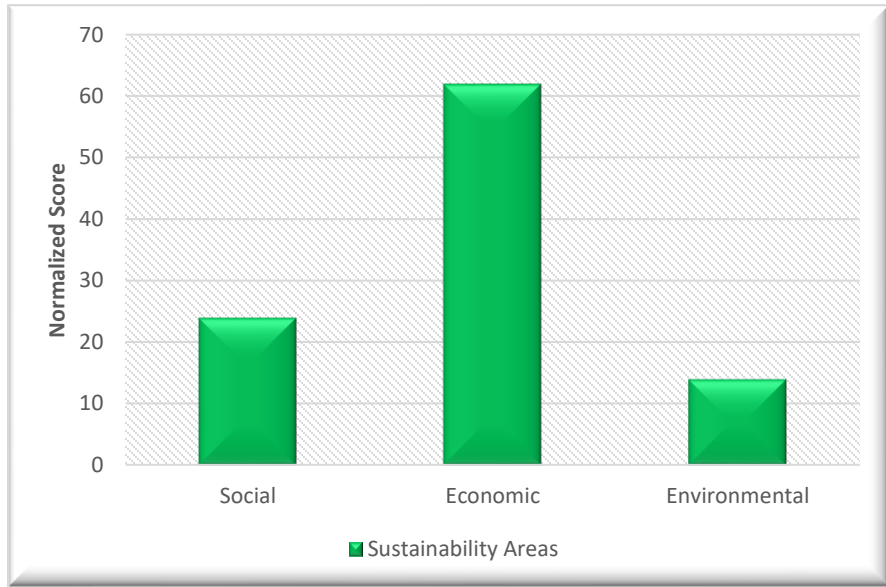


Fig.7. Sustainability areas score

18	Environment protection measures in project design	(Shen et al., 2010),(Chen et al., 2008)
19	Cleaning up Contaminated Water and Land	(Bourdeau, 1999),(Huang and Hsu, 2011)
20	Using Clean Technologies and Materials	(Bennett et al., 1999), (Chen et al., 2008), (Ross et al., 2010)

Table 2 Summary of available previous studies on sustainable success criteria

group	code	Success criteria	sources
Economical	1	Project completion within time	(Adinyira et al., 2012) , (Atkinson, 1999) , (Elattar, 2009) , (Ahadzie et al., 2008)
	2	Project completion within budget	(Adinyira et al., 2012) , (Atkinson, 1999) , (Elattar, 2009) , (Ahadzie et al., 2008)
	3	Project quality	(Adinyira et al., 2012) , (Atkinson, 1999) , (Elattar, 2009) , (Ahadzie et al., 2008)
	4	Internal return ratio (IRR)/Return on investment	(Elattar, 2009) , (Shen et al., 2010)
	5	Respond to project risks/Overall risk containment	(Adinyira et al., 2012) , (Ahadzie et al., 2008)
	6	Employer satisfaction/Client satisfaction/owner satisfaction	(Yeung et al., 2007) , (Pheng and Chuan, 2006) , (Ashley, 1986) , (Krajangsri and Pongpeng, 2016) , (Skibniewski and Ghosh, 2009) , (Ahadzie et al., 2008) , (Chan and Chan, 2004)
Social	7	Satisfaction of people in project neighbourhood/ End user satisfaction/ Customer's satisfaction	(Adinyira et al., 2012) , (Elattar, 2009) , (Yuan et al., 2011) , (Yeung et al., 2007) , (Müller and Turner, 2007) , (Ahadzie et al., 2008) , (Chan and Chan, 2004)
	8	Provision of employment opportunities	(Bennett et al., 1999) , (Shen et al., 2010) , (Chen et al., 2010) , (Alnaser et al., 2008)
	9	Overall health and safety measures/Accident rate	(Adinyira et al., 2012) , (Shen et al., 2010) , (Fernández-Sánchez and Rodríguez-López (2010) , (Ahadzie et al., 2008) , (Chan and Chan, 2004)
	10	Satisfaction of staffs/team satisfaction	(Müller and Turner, 2007) , (Chan and Chan, 2004)
	11	Environmental degradation	(Adinyira et al., 2012) , (Ahadzie et al., 2008) , (Chan and Chan, 2004)
Environmental	12	Noise pollution	(Shen et al., 2010) , (Fernández-Sánchez and Rodríguez-López (2010)
	13	Effect on air and land quality	(Chen et al., 2010) , (Shen et al., 2010)
	14	Adverse impact on historical sites and cultural heritage	(Krajangsri and Pongpeng, 2016) , (Shen et al., 2010)
	15	Energy consumption	(Alnaser et al., 2008) , (Bennett et al. (1999)

Table 3. Filtered rules extracted from Rosetta software

Success Perspective	Rule	LHS Support	RHS Support	RHS Accuracy	LHS Coverage	RHS Coverage	Class
Economical	F1(4) AND F2(4) AND F3(4) AND F4(3) => Economical Success(4)	4	4	1	0.21	0.57	3
	F1(3) AND F2(3) AND F3(4) AND F4(4) => Economical Success(3)	3	3	1	0.16	0.50	2
	F1(3) AND F2(4) AND F3(4) AND F4(2) => Economical Success(4)	2	2	1	0.11	0.29	3
	F1(1) AND F2(2) AND F3(1) AND F4(3) => Economical Success(2)	2	2	1	0.11	0.50	1
	F1(2) AND F2(1) AND F3(4) AND F4(1) => Economical Success(2)	2	2	1	0.11	0.50	1
	F1(1) AND F2(3) AND F3(4) AND F4(2) => Economical Success(3)	2	2	1	0.11	0.33	2
	F1(4) AND F2(5) AND F3(5) AND F4(3) => Economical Success(5)	2	2	1	0.11	1.00	4
	F1(5) AND F2(3) AND F3(4) AND F4(3) => Economical Success(3) OR (4)	2	1, 1	0.5, 0.5	0.11	0.17, 0.14	2,3
Social	F12(4) AND F10(3) AND F13(3) AND F14(4) => social success(3) OR (4)	4	3, 1	0.75, 0.25	0.22	0.5, 0.14	2,3
	F12(3) AND F10(4) AND F13(4) AND F14(4) => social success(4)	2	2	1	0.11	0.29	3
	F12(4) AND F10(5) AND F13(4) AND F14(5) => social success(5)	2	2	1	0.11	1.00	4
	F12(5) AND F10(4) AND F13(3) AND F14(5) => social success(4)	2	2	1	0.11	0.29	3
	F12(5) AND F10(4) AND F13(4) AND F14(4) => social success(4)	2	2	1	0.11	0.29	3
	F12(2) AND F10(2) AND F13(3) AND F14(2) => social success(2)	2	2	1	0.11	0.67	1
	F12(2) AND F10(3) AND F13(2) AND F14(4) => social success(2) OR (3)	2	1, 1	0.5, 0.5	0.11	0.34, 0.17	1,2
	F12(3) AND F10(2) AND F13(2) AND F14(4) => social success(3)	2	2	1	0.11	0.33	2
Environmental	F15(4) AND F16(3) AND F18(2) => environmental success(4) OR (3)	3	1, 2	0.34, 0.67	0.16	0.14, 0.5	2,3
	F15(1) AND F16(1) AND F18(3) => environmental success(2)	3	3	1	0.16	0.60	1
	F15(4) AND F16(3) AND F18(3) => environmental success(3)	2	2	1	0.11	0.50	2
	F15(5) AND F16(4) AND F18(4) => environmental success(5) OR (4)	5	3, 2	0.6, 0.4	0.26	1.0, 0.29	3,4
	F15(4) AND F16(4) AND F18(3) => environmental success(4)	2	2	1	0.11	0.29	3
	F15(4) AND F16(4) AND F18(5) => environmental success(4)	2	2	1	0.11	0.29	3
	F15(3) AND F16(2) AND F18(2) => environmental success(2)	2	2	1	0.11	0.40	1

Table 4. The linguistic variables for inputs and outputs at first stage

Inputs at first stage		Outputs at first stage		Inputs at second stage	
Very Bad (VB)	(1,1,2)	Very unsuccessful (VUS)	(0,0,25)	Unsuccessful(U)	(0,0,20,40)
Slightly Bad (SB)	(1,2,3)	Unsuccessful (US)	(0,25,50)	Moderate(M)	(30,50,50,70)
Moderate(M)	(2,3,4)	Moderately successful (MS)	(25,50,75)	Successful(S)	(60,80,100,100)
Good(G)	(3,4,5)	Very successful (VS)	(50,75,100)		
Excellent (E)	(4,5,5)	Extremely successful (ES)	(75,100,100)		

Table 5. Fuzzy rule-based matrix in the second stage

First Input	Second Input	Third Input	Output	First Input	Second Input	Third Input	Output
U	U	U	U	M	M	S	$0.86*U+0.14*M$
U	U	M	$0.86*U+0.14*M$	M	S	U	$0.62*U+0.24*M+0.14*S$
U	U	S	$0.86*U+0.14*S$	M	S	M	$0.74*M+0.24*S$
U	M	U	$0.76*U+0.24*M$	M	S	S	$0.62*M+0.38*S$
U	M	M	$0.62*U+0.38*M$	S	U	U	$0.62*S+0.38*U$
U	M	S	$0.62*U+0.24*M+0.14*S$	S	U	M	$0.62*U+0.24*M+0.14*S$
U	S	U	$0.76*U+0.24*S$	S	U	S	$0.76*S+0.24*U$
U	S	M	$0.62*U+0.24*M+0.14*S$	S	M	U	$0.62*U+0.24*M+0.14*S$
U	S	S	$0.62*U+0.38*S$	S	M	M	$0.62*S+0.38*M$
M	U	U	$0.62*M+0.38*U$	S	M	S	$0.76*S+0.24*M$
M	U	M	$0.76*M+0.24*U$	S	S	U	$0.86*S+0.14*U$
M	U	S	$0.62*U+0.24*M+0.14*S$	S	S	M	$0.86*S+0.14*M$
M	M	U	$0.86*M+0.14*U$	S	S	S	S
M	M	M	M				

Table 6. Decision makers' perception toward status of success factors in project B

Category	Success Factors	Decision Makers' Perception
Economical Perspective	Time management	4
	Cost management	4
	Quality management	3
	Risk management	2
Social Perspective	Leadership	4
	Competent project team	2
	Teamwork	3
	Motivation	3
Environmental Perspective	Attempt to preserve environment	3
	Waste Management	3
	Environment protection measures in project design	2

Table 7. Validation of proposed model at first stage

Project	Output	Assessment results using different defuzzification methods				
		COM	MOM	SOM	LOM	BOM
A	Economical success score	85	87.5	90	100	87
	Social success score	91	91	82	100	82
	Environmental success score	82	82	82	87	82
B	Economical success score	76	76	65	96	69
	Social success score	67	67	53	76	67
	Environmental success score	52	52	48	63	55
C	Economical success score	20	20	4	10	20
	Social success score	10	10	2	15	20
	Environmental success score	15	15	7	20	20

Scenario. No	Eco	So	Env	SSI (out of 100) using Trapezoidal MF
1	20	20	20	20
2	20	20	50	24.2
3	20	20	80	28.4
4	20	50	20	27.2
5	20	50	50	31.4
6	20	50	80	35.6
7	20	80	20	34.4
8	20	80	50	38.6
9	20	80	80	42.8
10	50	20	20	38.6
11	50	20	50	42.8
12	50	20	80	47
13	50	50	20	45.8
14	50	50	50	50
15	50	50	80	54.2
16	50	80	20	53
17	50	80	50	57.2
18	50	80	80	61.4
19	80	20	20	57.2
20	80	20	50	61.4
21	80	20	80	65.6
22	80	50	20	64.4
23	80	50	50	68.6
24	80	50	80	72.8
25	80	80	20	71.6
26	80	80	50	75.8
27	80	80	80	80

Table 8.
Prediction of SSI
based on
Scenario analysis