

Predicting the Impact Size of Uncertainty Events on Construction Cost And Time of Highway Projects Using ANFIS Technique

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Abstract. This study examines the ability and performance of Adaptive Neuro-Fuzzy Inference System (ANFIS) as an intelligent machine learning technique in the prediction of the impact size of uncertainty events on construction cost and time of highway projects. The rationale for the study stems from widespread reports of cost and time overruns on highway construction projects and the knowledge that the cost and time of projects are affected significantly by uncertainty events. Thus, the prediction of the impact size of uncertainty events during the design phase will enable project managers in preparing a proper plan with sufficient contingencies to deal with these uncertainty events. The success or failure of prediction depends on the credibility of the prediction method. In this study, the impact size of 76 uncertain events on the construction cost and time of highway projects were predicted using ANFIS technique, and the accuracy and reliability of ANFIS prediction were assessed. The results of R-Squared and four error tests proved that ANFIS is an accurate and reliable technique for predicting the impact size of uncertainty events on the cost and time of construction projects. Based on these findings, the study concludes that the use of intelligent machine learning methods such as ANFIS will minimise the potential inconsistency of correlations in construction cost and time prediction, improved accuracy in estimated project cost and time and reduced overruns.

Keywords: Accuracy, ANFIS, Construction cost and time, Impact size, Reliability, Uncertainty events.

1 Introduction

Construction of transportation infrastructure and particularly highway construction projects are one of the most unreliable projects in terms of estimating construction cost and time [1]. According to Abdullah, Rahman, et al. [2] 90% of the People's Trust Council highway construction projects experienced cost and time underestimation. Providing accurate estimates of cost and time on the construction of highway projects is a difficult task because construction projects are subject to risks and uncertainties, especially in the planning phase of the project when insufficient data and information about the project is available [3]. Typically, in the estimation of construction project

38 resources, several variables are not known since uncertainties populate construction
39 projects.

40 There are many common uncertainty events (also known as risk events) that
41 influence the cost and time of construction activities of construction projects that affect
42 construction performance differently [4]. According to Flyvbjerg [5], the occurrence of
43 uncertainty events in the construction of infrastructure projects, particularly in highway
44 projects, is higher than other construction projects, due to unique features of highway
45 projects, which include complexity between major construction activities, long duration
46 of construction, dynamic processes, repetitive linear projects, and mobile construction
47 sites. The project is regarded as successful if adequate allowance is made for
48 uncertainty events and a proper response is provided for their impacts on cost and time
49 of project which may or may not occur [6].

50 Estimating construction cost and time at the early stage of project development
51 represents a prediction provided by the estimator based on available information and
52 data. Estimating in construction is defined as that area of construction practice where
53 the estimator's experience and judgment are utilised in the application of scientific
54 principles and techniques to the problem of predicting and controlling cost and time of
55 projects [7]. The ability to accurately predict the impact size of uncertainty events on
56 construction time has always been one of the most critical challenges of estimators [8].
57 The success or failure of prediction depends on the credibility of the prediction
58 technique.

59 The impacts of risk and uncertainty events on cost and time of construction projects
60 have been approached in two major ways: qualitative techniques and quantitative
61 techniques. A qualitative technique such as contingency application and fault tree
62 analysis are more basic techniques, and the subjective judgment could not determine
63 the exact impacts of events [4]. Quantitative techniques such as regression modelling,
64 neural networks and Monte Carlo simulation are more common and preferable than the
65 qualitative techniques because of the capability of quantifying the impacts of such
66 events [9]. However, quantification techniques are complex and hard and furthermore
67 require exact data, and such data either do not exist at all or are hard to obtain in the
68 realisation of construction projects [10]. Furthermore, the applicability of these
69 techniques in predicting the impact size of uncertainty events in real construction
70 projects is limited due to the unpredictable nature of construction projects and their
71 dependence on the estimator's thinking prototype in the process of uncertainty analysis
72 [10].

73 On the other hand, hybrid intelligence machine learning methods based on Fuzzy
74 Inference System (FIS) is used in modelling the qualitative aspects without employing
75 precise quantitative analyses, it provides standard practical methods for transformation
76 into rule-based as well as effective methods for turning Membership Functions (MF)
77 for better performance index [11]. The Adaptive Neuro-Fuzzy Inference System
78 (ANFIS) combines the strengths of Artificial Neural Network (ANN) with Fuzzy
79 Inference Systems (FIS) to create an efficient method for analysing the complex
80 problems. Also, neural network fuzzy systems interpret human knowledge and deduce
81 it into a mathematical model. These two tools were combined to achieve readability
82 and learning ability at the same time. ANFIS, trained to develop fuzzy rules and

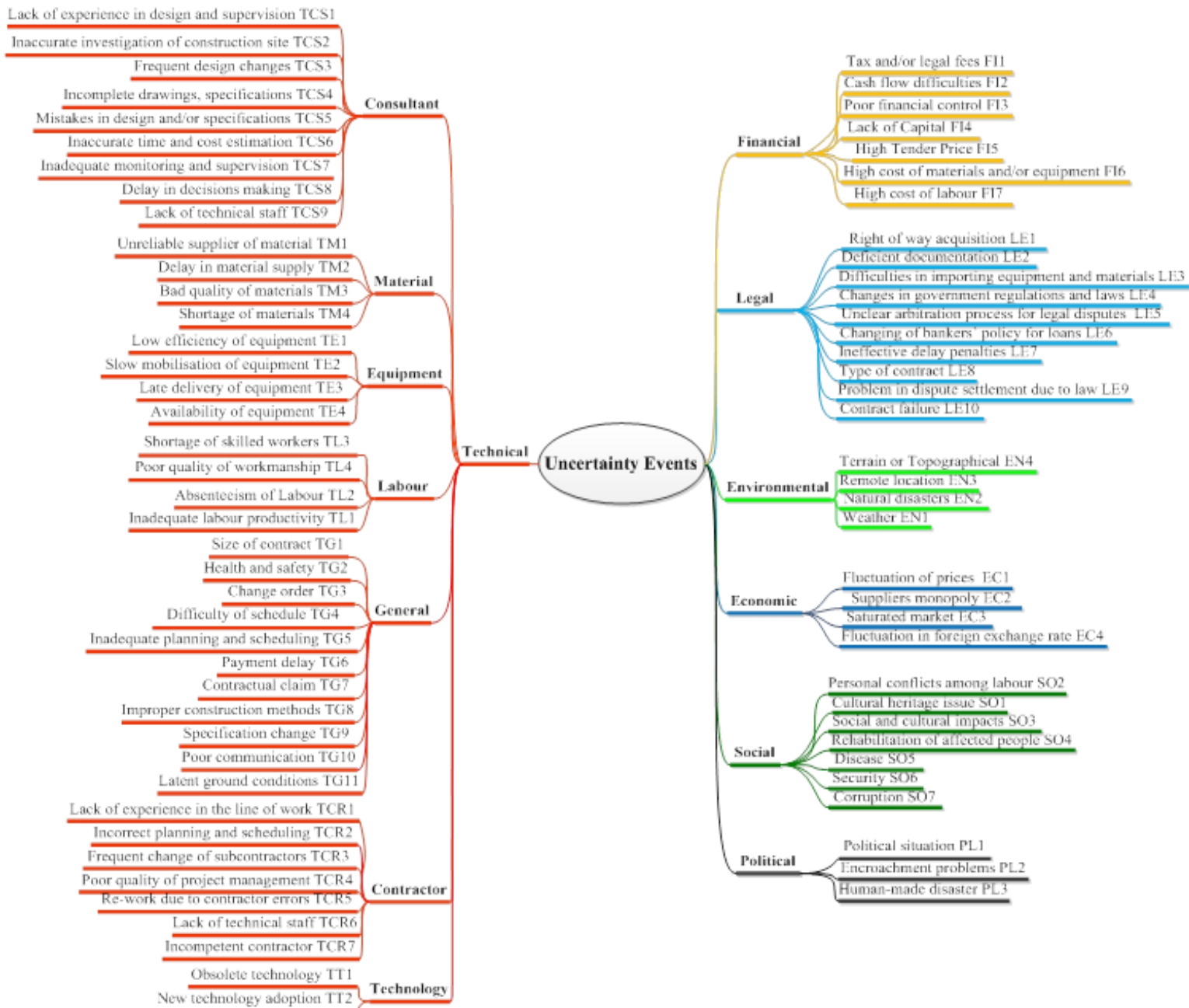
83 determine Membership Functions (MF) for input and output variables of the system is
84 an intelligent system which can estimate the variables and fuzzy rules intelligently and
85 does not require a systematic method for design of fuzzy systems. ANFIS can handle
86 nonlinearity, uncertainty, and complex problems which are involved in predicting the
87 impact size of uncertainty events on construction cost and time [12]. Generally, Neuro-
88 Fuzzy has two major categories namely: fuzzy linguistic modelling which is focused
89 on interpretability (Mamdani), and precise fuzzy modelling which is focused on
90 accuracy (Takagi-Sugeno) [13].

91 ANFIS has been used in various fields of engineering. For instance, Ugur [14] has
92 developed ANFIS to estimate the costs of the residential building. While Fragiadakis,
93 Tsoukalas, et al. [15] assessed the occupational risk in the shipbuilding industry; Ebrat
94 and Ghodsi [10] applied ANFIS to evaluate the risk in construction projects; Li, Su, et
95 al. [16] forecasted building energy consumption using hybrid ANFIS; Güneri, Ertay,
96 et al. [17] used ANFIS to overcome supplier selection problem in construction projects;
97 Shahhosseini and Sebt [18] used ANFIS to establish a fuzzy adaptive decision-making
98 model for selection and assignment of human resources to construction projects based
99 on competency; and Wang and Elhag [19] developed an ANFIS based risk assessment
100 model for bridge maintenance projects.

101 However, there are limited researches undertaken using ANFIS in the field of
102 construction management and limited studies have evaluated the performance of
103 ANFIS as a technique for predicting the impact size of uncertainty events on
104 construction cost and time of highway projects. Therefore, this study examines the
105 predicted impact size of uncertainty events on the construction cost and time of
106 highway projects using ANFIS as an intelligence method, and after that assessed the
107 ability and accuracy of prediction of ANFIS.

108 **2 Uncertainty Events on highway projects**

109 An uncertainty event (also known as risk or unforeseen event) is an event with a
110 substantial impact on the construction activities and process and occurs with some
111 probability of occurrence, and severely disrupt the construction process if it occurs [20].
112 Uncertainty events are associated with the project characteristics and location and has
113 no root causes that can be generalised [21]. Therefore, there is an obvious need to
114 effectively anticipate, identify and classify the uncertainty events on different locations
115 and projects to assess their influence on the objectives of construction projects.
116 Moghayedi and Windapo [22] identified 76 uncertainty events as affecting the
117 construction of South African highway projects under seven major groups namely
118 Economic, Environmental, Financial, Legal, Political, Social and Technical as
119 presented in Figure 1. The 76 uncertainty events identified by Moghayedi and Windapo
120 [22] was used in the current study to evaluate the performance of proposed techniques
121 for predicting the uncertainty events.
122



123 **Fig. 1.** Uncertainty events on South African highway construction projects
 124 distributed according to causative factors

125 3 Methodology

126 This paper developed An ANFIS model and evaluated its performance in predicting
 127 the impact size of uncertainty events on construction cost and time of highway projects.
 128 The hierarchy structure of the model developed for predicting the impact size of
 129 uncertainty events consists of four main steps; determination of uncertainty event
 130 attributes, data collection, developing the predicting models, evaluating the
 131 performance of the developed model and recommendations.

132 The process begins with a determination of criteria for uncertainty assessment. The
 133 impact size of uncertainty is assessed by two parameters, probability of occurrence and
 134 severity [23]. To model, the impact of uncertainty events on construction time, the ISO
 135 31000 (International Standard Organization) impact matrix was used [23]. Figure 2
 136 shows the probability of occurrence and severity as two input variables and relevant
 137 impact size as the output of the predicting model.

		Severity				
		Insignificant (1)	Minor (3)	Moderate (5)	Major (7)	Catastrophic (9)
Probability of Occurrence	Rare (.1)	Minimal (1)	Minimal (1)	Low (2)	Low (2)	Moderate (3)
	Unlikely (.3)	Minimal (1)	Low (2)	Moderate (3)	Moderate (3)	High (4)
	Possible (.5)	Low (2)	Moderate (3)	Moderate (3)	High (4)	High (4)
	Likely (.7)	Low (2)	Moderate (3)	High (4)	High (4)	Extreme (5)
	Almost Certain (.9)	Moderate (3)	High (4)	High (4)	Extreme (5)	Extreme (5)

138 **Fig. 2.** Uncertainty impact matrix

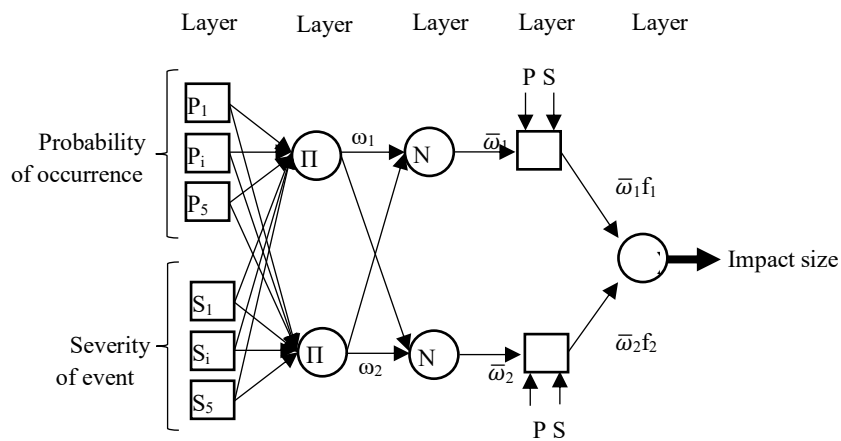
139 In the second step, the uncertainty event attributes (probability of occurrence and
 140 severity) were obtained from 32 project managers with a minimum of 20 years of work
 141 experience in South African highway construction projects recommended by The South
 142 African National Roads Agency SOC Ltd (SANRAL) using the five-point linguistic
 143 Likert scale questionnaires.

144 In the third step, an ANFIS model is developed in MATLAB based on the impact
 145 matrix to predict the impact size of uncertainty events on construction cost and time of
 146 highway projects to assess the impact size of uncertainty events in construction cost
 147 and time of highway projects. In the last step, the fitness of predicted impact sizes to
 148 the real data are evaluated by R-Square, and the reliability of ANFIS on predicting
 149 impact size are evaluated using Root Mean Square Error (RMSE) and Mean Absolute
 150 Percentage Error (MAPE). The predicted impact size and the performance values of
 151 ANFIS on construction cost and time of highway projects are presented in the next
 152 section.

153 The RMSE, MAPE and of predicted impact size of uncertainty events are calculated
 154 to evaluate the performance of ANFIS. The result of the selected 20 events is presented
 155 in Table 1.

156 4 Results

157 The study developed a first-order Takagi-Sugeno fuzzy inference system to accurately
 158 assess the impact size of uncertainty events on construction cost and time of highway
 159 projects. In this inference system, the output of each rule is a linear combination of two
 160 input variables added by a linear term of “AND” logic. The final output is the weighted
 161 average of each rule’s output [24]. Figure 3 illustrates the Takagi-Sugeno ANFIS
 162 structure which was developed for this study. To model this ANFIS following 25 fuzzy
 163 rules “If-Then” are considered.



164 **Fig. 3.** Developed Takagi-Sugeno ANFIS structure
 165 Eighty per cent (80%) of the data collected from the research participants was used for
 166 training of the FIS, while the twenty per cent (20%) was used for checking and testing
 167 the neural network which set the system parameters. The RMSE, MAPE and R-Square
 168 value of predicted impact size of uncertainty events are calculated to evaluate the
 169 ANFIS performance and the result for the selected 20 events are presented in Table 1.