

Research Article

Comparison of Fuzzy AHP and Fuzzy TOPSIS for Road Pavement Maintenance Prioritization: Methodological Exposition and Case Study

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Received 30 April 2015; Revised 15 June 2015; Accepted 16 June 2015

Academic Editor: Samer Madanat

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For road pavement maintenance and repairs prioritization, a multiattribute approach that compares fuzzy Analytical Hierarchy Process (AHP) and fuzzy Technique for Order Preference by Ideal Situation (TOPSIS) is evaluated. The pavement distress data was collected through empirical condition surveys and rating by pavement experts. In comparison to the crisp AHP, the fuzzy AHP and fuzzy TOPSIS pairwise comparison techniques are considered to be more suitable for the subjective analysis of the pavement conditions for automated maintenance prioritization. From the case study results, four pavement maintenance objectives were determined as road safety, pavement surface preservation, road operational status and standards, and road aesthetics, with corresponding depreciating significance weights of $W = [0.37, 0.31, 0.22, 0.10]^T$. The top three maintenance functions were identified as Thin Hot Mix Asphalt (HMA) overlays, resurfacing and slurry seals, which were a result of pavement cracking, potholes, raveling, and patching, while the bottom three were cape seal, micro surfacing, and fog seal. The two methods gave nearly the same prioritization ranking. In general, the fuzzy AHP approach tended to overestimate the maintenance prioritization ranking as compared to the fuzzy TOPSIS.

1. Introduction

Road pavement condition survey and assessment are an important component in the decision making procedure in pavement management systems (PMS). The two main reasons for the assessment of pavement condition are to recognize and prioritize maintenance requirements in specific road sections and to identify the pavement network conditions and, if any, the rehabilitation requirements [1]. In urban road maintenance and rehabilitation (M&R) projects, the evaluation of road pavement conditions is the initial phase that guarantees suitable and sustainable maintenance programs [2].

For pavement M&R projects, specific pavement maintenance management systems (PMMS) should be proposed, as suitable systems for the improvement of the efficiency of decision making. Furthermore, such systems should be able to provide feedbacks on the impacts of the decisions in terms of short and long term maintenance strategies and with

respect to cost-benefit analysis [3]. A typical structure that integrates PMS, PMMS, and M&R is depicted in Figure 1.

From Figure 1, it is seen that a PMS is supposed to be a coordinated and systematic process for carrying out all activities related to providing and maintaining pavements. On the other hand a PMMS should be able to predict the pavement condition and the cost associated with its M&R over a given time frame and also to aid in the planning and programming of the associated M&R works. In implementing PMMS, multicriteria priority ranking and analysis can be used to rank and select pavement sections due for maintenance in the order of urgency. PMMS is considered to be a timely, cost-beneficial, and preventive M&R approach in the maintenance as compared to the traditional reactive M&R approach.

Pavement Maintenance Prioritization Using Multicriteria Decision Approach. The main multicriteria decision making approach is the Analytic Hierarchy Process (AHP) which was

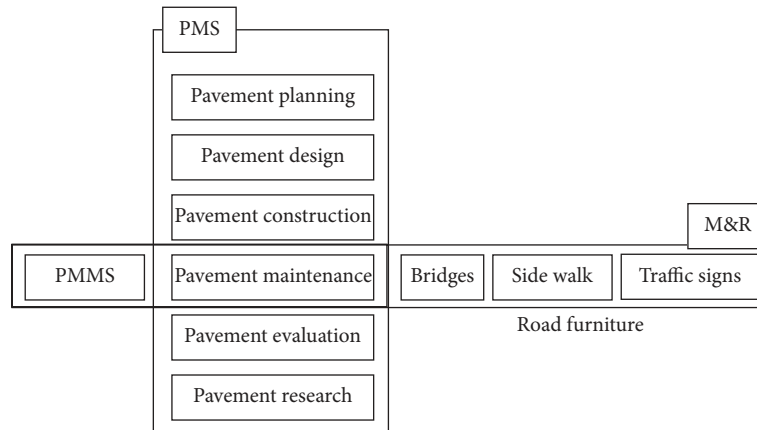


FIGURE 1: Schematic structure of a typical pavement management system (PMS) and pavement maintenance and management system (PMMS).

introduced by Saaty in the 1970s [4, 5]. Saaty developed AHP to primarily meet the complexities and challenges of decision situations that are brought out by multiple and conflicting criteria as a multicriteria decision making (MCDM) tool. AHP aids decision makers in determining the best criteria out of multiple sets of criteria, based on crisp or exact numerical values.

On road pavement maintenance priority ranking and analysis, AHP has been widely used (e.g., [6, 7]). However, due to the fact that in most practical situations the human preference is marred by uncertainty, decision makers would find it difficult to explicitly assign exact numerical values to the comparison judgments [8]. Therefore the decision maker would theoretically find it very difficult to express the strength of his preferences and confidence in terms of the AHP in pairwise comparison judgments. As such AHP has been argued to be ineffective when employed to ambiguous or vaguely represented real-life problems that consist of uncertainty and subjectivity [9].

In decision making processes, uncertainty, imprecision, and subjectivity can be handled by means of probability theory and or fuzzy set theory (FST) [10]. While probability focuses on the stochasticity of the decision making process, FST tends to formalize the subjective and imprecise nature of human behavior by representing and quantifying vague information through a membership grade function [10]. As such, probabilistic theory is not adequately suited for subjectivity and imprecision in the human decision making process [11–14].

The advantage of fuzzy based pairwise comparison is that it allows decision makers to be more flexible in their judgments. This is attained through the different degrees of fuzzification. Furthermore if there is need for consideration of the attitude toward risk, FTS has the possibility of overlapping preferences of criteria in case the expert is not confident about the degree of importance within a set of decisions and there is the provision for interval decision as expressed by a fuzzy membership function [15]. FST can also be considered useful in decision problems where the degrees of uncertainty are expected to increase or change in time,

for example, in pavement failure and deterioration, where different conditions are expected to change with time, but there is no predictive information on the future state.

In order to accommodate uncertainty in experts' heuristic and experiential judgments, Shah et al. [3] used the ranking of road maintenance factors based on subjective rating. And to take care of subjectivity in judgment, fuzzy pairwise comparison deduction technique can be integrated with AHP [3, 16, 17]. Used independently, fuzzy AHP results are directly ranked according to the normalized weights. However, in real life decision, decision makers' degrees of confidence and attitudes to risks should be taken into consideration.

Fuzzy Technique for Order Preference by Ideal Situation (TOPSIS) technique operates on the basis that the most preferred decision alternative should not only have the shortest distance from a positive ideal solution (PIS), but also have the farthest distance from the negative ideal solution (NIS), hence its capability and efficiency in handling uncertainty [18, 19]. However the use of Euclidean Distance as determined from PIS and NIS in TOPSIS does not consider the correlation of attributes and that makes it difficult to derive the weights of the decision attributes while keeping the judgment consistency [20]. Thus the integration fuzzy TOPSIS with AHP is considered in this study, in addition to taking into account the degree of confidence in decision making.

In this study, fuzzy approach, allowing experts to use linguistic variables, is considered for prioritization by applying and comparing the results from fuzzy AHP and fuzzy TOPSIS. The classical AHP is utilized to determine the weight vector of road maintenance factors with respect to the set objectives. In principle, therefore, the two approaches differ primarily in the way the fuzzy ratings are aggregated and normalized but do impact on the data collection procedure.

First, a methodological overview and approach for fuzzy AHP and fuzzy TOPSIS for the formulation and automation of cost-effective PMMS for consistent and reliable prioritization of M&R of road pavements in urban areas are presented. In the implementation, a case study in Eldoret Town in Kenya is used. For thematic display of the road network, a graphical user interface was designed in GIS and

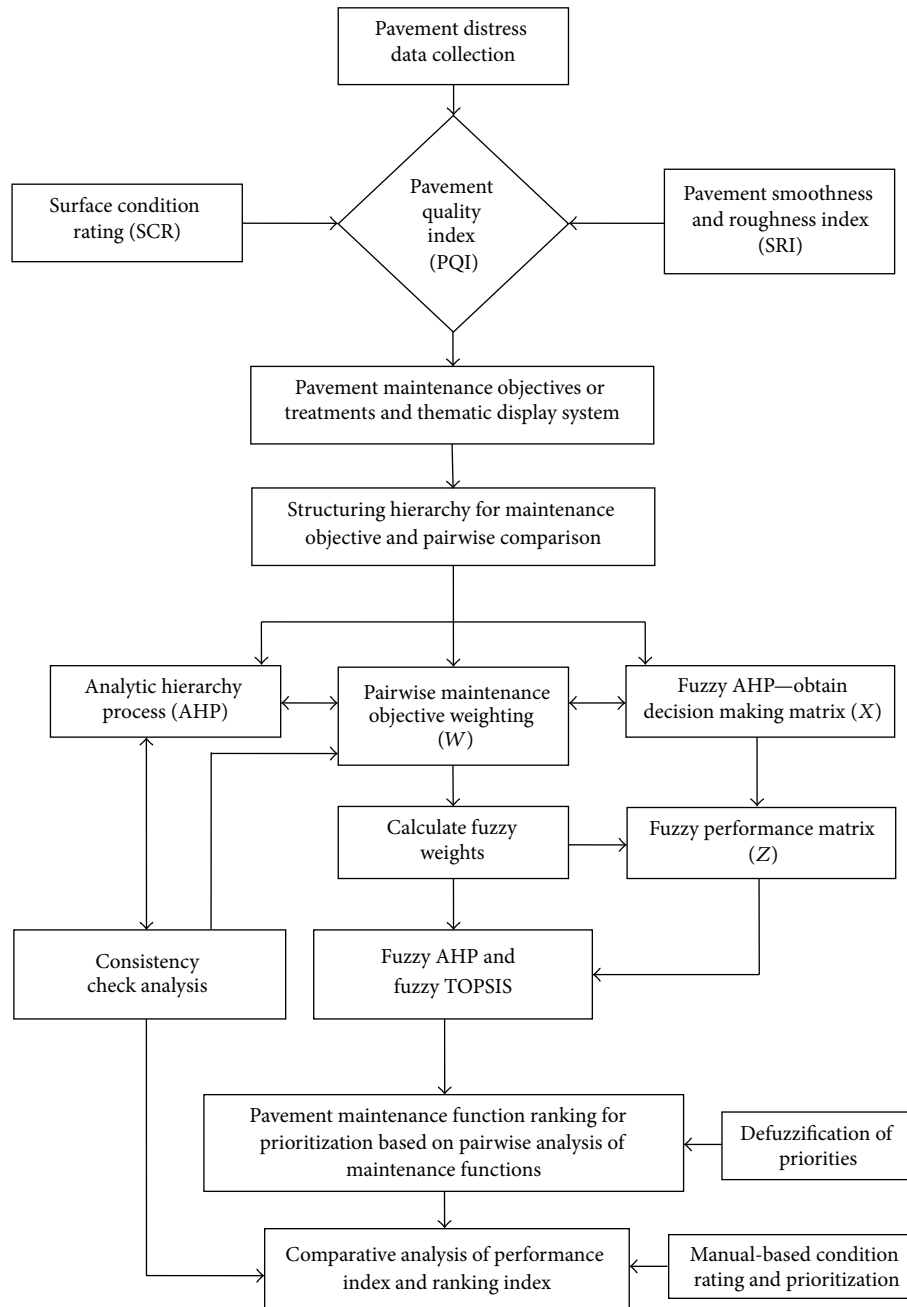


FIGURE 2: Schematic framework for road pavement prioritization using fuzzy AHP and fuzzy TOPSIS.

SQL database management systems. In summary, Figure 2 presents the schematic framework for the methodological approach in this study.

The rest of the paper is organized as follows. Section 2 is on the case study pavement distress data collection, cataloging, and display. Section 3 presents the methodological overview and approach based on fuzzy AHP and fuzzy TOPSIS. Section 4 presents the results of the case study, and Section 5 presents the analysis and discussions of the results. The study conclusions and recommendations are presented in Section 6.

2. Case Study Pavement Distress Catalogue and Display

There are two main approaches for the collection of pavement distress data: manual (visual inspection) distress data collection and automated distress data collection [21]. For cost-effective and simplified pavement condition surveys, this study adopted empirically manual-based inspection approach to identify the pavement failures or distresses. GPS was used to obtain the chainages of the observed pavement failures and to locate the associated pavement distresses.

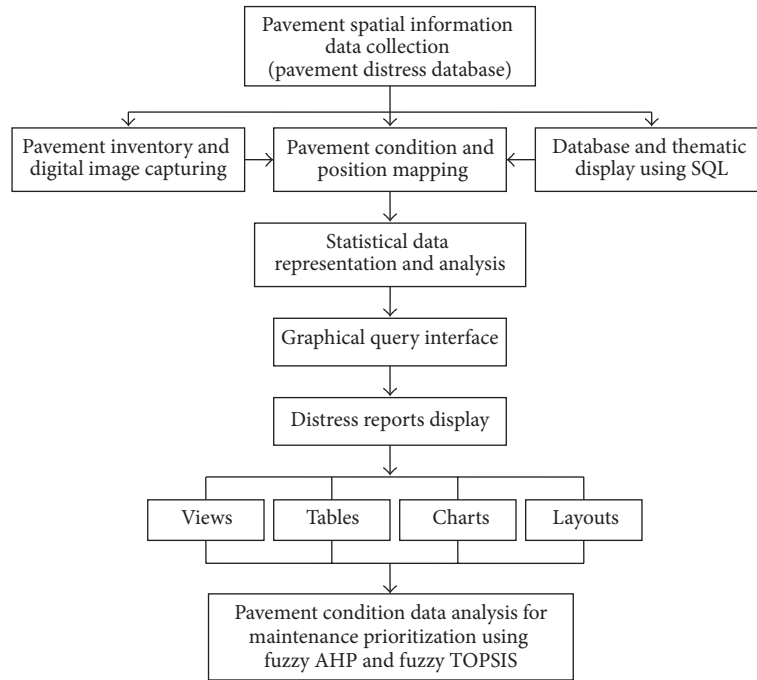


FIGURE 3: Proposed framework for pavement inventory and condition survey thematic display system.

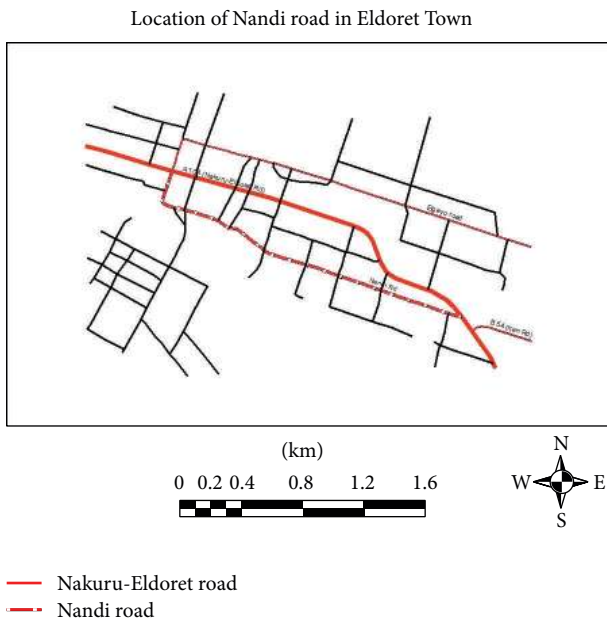


FIGURE 4: Study road and the adjoining roads digitized and georeferenced from Google Maps image and GPS field observations.

A summary of the structure for the pavement distress catalogue and display system is presented in Figure 3. The result of this stage of mapping is presented in Figure 4 for the case study of Nandi road in Eldoret Town (Kenya).

2.1. Pavement Distress Inventorizing and Failure Index. Since the collection of detailed pavement condition data is very

costly and time-consuming, innovative approaches for rapid data collection are in increasing demand among highway agencies with limited PMS budgets [22]. For the case study, visual based empirical inspection and rating (ECR), for distresses data collection, was adopted for the 3 km Nandi road, with observations made at 300 m segments. For each inspected segment of the road, a digital image of the observed distresses was also captured for thematic display.

In order to determine the degree of distress failure two approaches are used. One is a surface condition rating (SCR) to define the failure criteria for cracking and pothole based distresses. According to the proposed SCR, a rating scale of 0–3 was adopted, with a new road surface (no pothole, no cracking) having an SCR of 3. As the amount and severity of the defect increase, the SCR also drops. That implies that low cracking has an SCR of 3, while medium to severe cracking ranges from 2 to 0. In terms of the defect extent, defined by the depth, area, and/or length, an SCR of 3 implies negligible defects, while SCR index values of 2–0 imply wide or long defects that require repairs.

The second component of pavement failure is the pavement surface ride quality. This was characterized by patching and raveling distresses indicators and expressed in terms of the degree of smoothness or roughness index (SRI). The SRI is also ranked between 3 and 0, such that a new road pavement would have an SRI of 3. SRI indices of between 2 and 1 imply that the road has deteriorated to the extent where motorists feel that it is uncomfortable driving on and therefore needs rehabilitation. An SRI of 0 then means that there is complete surface failure requiring major rehabilitation, as is characterized by continuous cracking, potholes, patching, and raveling distresses. The pavement distresses that make up the general pavement condition are determined by the expert judges

comprising of pavement engineers and road inspectors from the road management department.

From the experts' judgments and analyses, it was deduced that the overall road pavement quality index (PQI) indicator is defined by the geometric mean of SCR and SRI as expressed in (1). In (1), PQI values range from 3 to 0 with a PQI value of between 2 and 0 implying that the pavement is subject to maintenance and/or rehabilitation:

$$PQI = \sqrt{SCR * SRI}. \quad (1)$$

In the implementation, it is recommended that PQI indices should not be rounded off; that is, a PQI of $\sqrt{1 * 3} = \sqrt{3} = 1.732$ is not equivalent to 2. This precaution is taken in order to avoid minor repair maintenances at any given time in the pavement life-span for economic reasons, at the expense of more immediate and larger M&R cases. From the PQI, the road pavement maintenance objectives and functions can then be derived.

A summary of the observed distress results for the 3 km road is presented in Table 1. The results in Table 1 show that the distresses were mainly characterized by cracking, potholes, patching, and raveling. The results in Table 1 will be subjected to pairwise comparison and prioritization by the pavement management experts in order to determine the priority areas for M&R activities. Figure 5 characteristically shows some of the functional and structural failures that are observed on the case study road, as summarized in Table 1.

3. Methodology

3.1. AHP for Decision Prioritization. AHP is often used to guide decision makers coping with multicriteria decisions involving multiattribute situations [5, 23–25]. Decision judgments as articulated by pairwise comparisons are the fundamental inputs for facilitating the AHP procedure. Every pairwise comparison results in a numerical value a_{ij} representing the ratio between the weights of the two decision criteria, i and j . The Saaty preference scale is used to assign numerical values to different levels of preferences. As a standard, the Saaty scale used for AHP ranges from 1 to 9 and reflects the importance of one factor over another as represented in Table 2.

Supposing that $C = \{C_j \mid j = 1, 2, \dots, n\}$ is the set of criteria, then the evaluation matrix can be obtained, in which every element a_{ij} ($i, j = 1, 2, \dots, n$) represents the relative weights of the criteria illustrated:

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}. \quad (2)$$

In (2), n is the number of criteria in a decision process, such that A is the judgment which contains the pairwise comparison value a_{ij} for all $i, j \in \{1, 2, \dots, n\}$.

For multiple decision makers or experts as is often the case, if h is the number of decision makers and a_{ij}^k is the

TABLE 1: Road condition survey data showing pavement segment chainage and the observed distresses, as derived from the PQI calculations.

Segment or chainage (m)	Observed distresses
0–300	(i) Potholes (ii) Longitudinal cracks (iii) Transverse cracks (iv) Alligator cracks
300–600	(i) Longitudinal cracks (ii) Transverse cracks (iii) Block cracking
600–900	(i) Potholes (ii) Longitudinal cracks (iii) Transverse cracks (iv) Alligator cracks (v) Patching
900–1200	(i) Potholes (ii) Longitudinal cracks (iii) Alligator cracks
1200–1500	(i) Potholes (ii) Longitudinal cracks (iii) Edge cracking (iv) Raveling
1500–1800	(i) Potholes (ii) Longitudinal cracks (iii) Transverse cracks (iv) Block cracking
1800–2100	(i) Potholes (ii) Longitudinal cracks (iii) Alligator cracks
2100–2400	(i) Potholes (ii) Longitudinal cracks (iii) Transverse cracks (iv) Block cracking
2400–2700	(i) Potholes (ii) Longitudinal cracks (iii) Alligator cracks
2700–3000	(i) Potholes (ii) Longitudinal cracks (iii) Transverse cracks (iv) Alligator cracks (v) Block cracking (vi) Raveling

pairwise comparison value of criteria i and j given by decision maker k , where $k = 1, 2, \dots, h$, then by using geometric mean of the a_{ij}^k conducted by each decision maker, a new compound judgment matrix is derived, with the corresponding elements determined as follows:

$$a_{ij} = (a_{ij}^1 * a_{ij}^2 * \cdots * a_{ij}^k * \cdots * a_{ij}^h)^{1/h} = \left(\prod_{k=1}^h a_{ij}^k \right)^{1/h}. \quad (3)$$

The basic procedure for AHP approach as represented by the mean of normalized values method can be summarized in the following steps according to Saaty [26]: (i) normalization of each column to get a new judgment matrix (A'), (ii) summing up of each row of normalized judgment matrix

TABLE 2: The nine-point Saaty scale of relative importance [4].

Intensity of importance	Definition	Explanation
1	Equal importance	Two activities contribute equally to the objective.
3	Weak importance of one over the other	Experience and judgment slightly favour one over the other.
5	Essential or strong importance	An activity is strongly favoured and its dominance is demonstrated in practice.
7	Demonstrated importance	The evidence favouring over another is of highest possible order of affirmation.
9	Absolute importance	When compromise is needed.
2, 4, 6, 8	Intermediate values	



FIGURE 5: Photographs of typically observed road pavement distresses.

(A') to get weight vector (V), and (iii) finally defining the final normalization weight vector (W). Detailed and stepwise implementation of the AHP techniques can be found in Ouma and Tateishi [25].

According to the relative importance in Table 2, a decision maker who, for example, assigns a value of 3 to a pairwise comparison implies that the judgment slightly favours one

alternative over the other. However, because of the subjective nature in linguistic decisions (e.g., very high, high, moderate, low, and very low), such an assessment cannot be accurate. Furthermore, when the decision maker's judgments are uncertain, determining such precise Saaty's static nine-point scale values may be very difficult. This means that the static crisp values may lack the ability to capture the decision

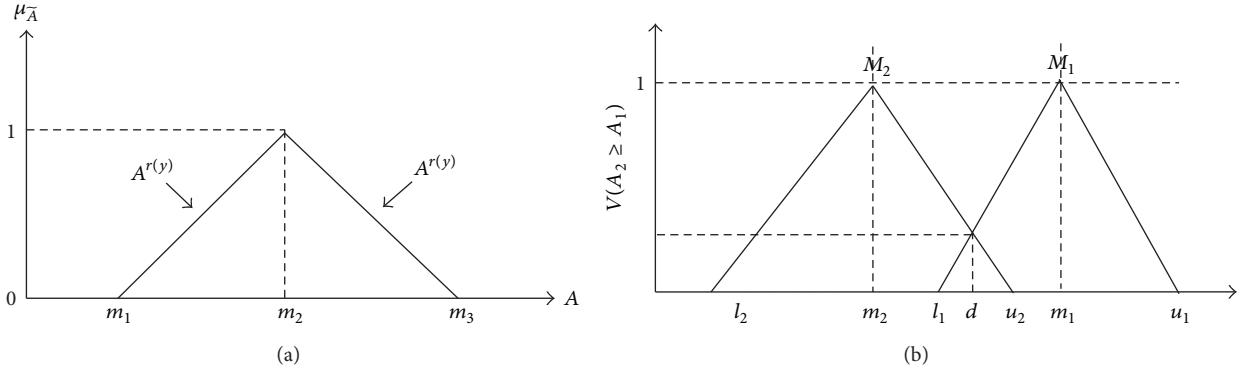


FIGURE 6: (a) A triangular fuzzy number, M , and (b) the intersection between M_1 and M_2 in extent fuzzy analysis.

makers' ambiguous and/or vague preferences. The logical way to overcome this limitation is to define the comparison ratios in terms of fuzzy numbers.

3.2. Fuzzy Sets Theory and Fuzzy AHP. In contrast to classical set theory, FST allows for the assessment of the membership of elements with respect to a set represented by $\mu = [0, 1]$ [10]. The fuzzy AHP is an extension of AHP technique for multicriteria decision making in pairwise based comparisons that deals with qualitative and imprecise real-world decisions problems.

In essence fuzzy AHP is a synthetic extension of the classical AHP method when the fuzziness of the decision maker's is considered. Fuzzy AHP is comprised of the following steps in implementation: (i) structuring of the levels of hierarchy for decision making, (ii) prioritization based on fuzzy pairwise comparisons, (iii) checking for consistency of the preference judgments by the experts, (iv) synthesis of pairwise priorities, and (v) defuzzification of the determined priorities according to Chen and Hsieh [27]. Defuzzification is necessary because the fuzzy arithmetic means are not crisp values and hence cannot directly be ranked. Defuzzification thus refers to the fuzzy ranking method that is employed so as to obtain the desired nonfuzzy or crisp values [28].

A fuzzy AHP decision problem consists of a number of alternatives $[M_i (i = 1, 2, \dots, m)]$, a set of evaluation criteria $[C_j (j = 1, 2, \dots, n)]$, a linguistic judgment (r_{ij}) representing the relative importance of each pair of criteria, and a weighting vector $[W = (w_1, w_2, \dots, w_n)]$. Like the classical AHP, fuzzy AHP also has a judgment matrix. However the extent fuzzy approach uses the triangle fuzzy numbers (TFNs) instead of a constant pairwise comparison value [29]. The extent analysis based fuzzy AHP depends on the degree of possibilities of each criterion used in the decision process. TFNs for the linguistic variables from the judgments are placed according to the responses and for a particular level on the decision hierarchy, from which the pairwise comparison matrix is constructed [30, 31].

According to Kaufmann and Gupta [32], a fuzzy number \tilde{F} is a triangular fuzzy number if its membership function $\mu_{\tilde{F}}(x) : \mathfrak{R} \rightarrow [0, 1]$, is defined according to (4), and is depicted in Figure 6. As in (4), the comparison ratios between

the success factors i and j are characterized with TFN which describes the judgment about a_{ij} and are denoted by \tilde{a}_{ij} . Hence it is possible to describe some degree of vague or ambiguous human perception and decision in pairwise comparisons. In Figure 6(a), a TFN is denoted by either $(m_1/m_2, m_2/m_3)$ or (m_1, m_2, m_3) . For a fuzzy event, the parameters m_1 , m_2 , and m_3 in Figure 6(a), respectively, denote the smallest possible value, the most promising value, and the largest possible value. Each TFN has linear representations on its left and right sides such that its membership function can generally be defined as follows:

$$\mu_{\tilde{A}}(x) = \begin{cases} \left(x - \frac{l}{m} - l\right), & l \leq x \leq m, \\ \left(u - \frac{x}{u} - m\right), & m \leq x \leq u, \\ 0, & \text{otherwise,} \end{cases} \quad (4)$$

where $l (= m_1)$ and $u (= m_3)$, respectively, represent the lower and upper bounds of the fuzzy number \tilde{A} , while $m (= m_2)$ is the median value as shown in Figure 6(a). Thus a TFN is generally denoted by $\tilde{A} = (l, m, u)$.

As shown in Figure 6(a), a fuzzy number can be given by its corresponding left and right representation of each degree of membership, defined by

$$\begin{aligned} \tilde{A} &= (A^{l(y)}, A^{r(y)}) \\ &= (m_1 + (m_2 - m_1)y, m_3 + (m_2 - m_3)y), \end{aligned} \quad (5)$$

$\forall y \in [0, 1],$

where $l(y)$ and $r(y)$ denote the left side representation and the right side representation of a fuzzy number, respectively. Matrix \tilde{A} represents an $(n \times n)$ judgment matrix containing TFNs (\tilde{a}_{ij}) for all $i, j \in \{1, 2, \dots, n\}$, as expressed in the following:

$$\tilde{A} = (\tilde{a}_{ij}) = \begin{bmatrix} (1, 1, 1) & \tilde{a}_{12} & \cdots & \tilde{a}_{1n} \\ \tilde{a}_{21} & (1, 1, 1) & \cdots & \tilde{a}_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \tilde{a}_{n1} & \tilde{a}_{n2} & \cdots & (1, 1, 1) \end{bmatrix}. \quad (6)$$

TABLE 3: (a) Fuzzy linguistic variables and corresponding fuzzy numbers. (b) Fuzzy number memberships of pairwise comparisons [8].

(a)			
Linguistic variable	Fuzzy number	Membership function	Definition
Equally important	$\tilde{1}$	(1, 1, 2)	Practical knowledge and experience imply that factor i is equally important when compared to factor j .
Moderately important	$\tilde{3}$	(2, 3, 4)	Practical knowledge and experience imply that factor i is moderately more important when compared to factor j .
More important	$\tilde{5}$	(4, 5, 6)	Practical knowledge and experience imply that factor i is more important when compared to factor j .
Strongly important	$\tilde{7}$	(6, 7, 8)	Practical knowledge and experience imply that factor i is strongly important when compared to factor j .
Extremely important	$\tilde{9}$	(8, 9, 9)	Practical knowledge and experience imply that factor i is extremely important when compared to factor j and totally outweighs it.

(b)			
a	Fuzzy number	\tilde{a}	Fuzzy number
1	(1, 1, 2)	0.50	(0.33, 0.5, 1)
2	(1, 2, 3)	0.33	(0.25, 0.33, 0.5)
3	(2, 3, 4)	0.25	(0.2, 0.25, 0.33)
4	(3, 4, 5)	0.20	(0.17, 0.2, 0.25)
5	(4, 5, 6)	0.17	(0.14, 0.17, 0.2)
6	(5, 6, 7)	0.14	(0.13, 0.14, 0.17)
7	(6, 7, 8)	0.13	(0.11, 0.13, 0.14)
8	(7, 8, 9)	0.11	(0.11, 0.11, 0.13)
9	(8, 9, 9)	Diagonal elements	(1, 1, 1)

In general, $\tilde{a}_{ij} = (l_{ij}, m_{ij}, u_{ij})$, with l_{ij} being the lower limit, u_{ij} the upper limit, and m_{ij} the most likely value, such that $m_{ij} = \sqrt{l_{ij} * u_{ij}}$. Assume that \tilde{A}_1 and \tilde{A}_2 are two triangular fuzzy numbers with $\tilde{A}_1 = (l_1, m_1, u_1)$ and $\tilde{A}_2 = (l_2, m_2, u_2)$. Some of the fuzzy functional operators (additive, multiplicative, and inverse) are represented in the following:

$$\tilde{A}_1 \oplus \tilde{A}_2 = (l_1 + l_2, m_1 + m_2, u_1 + u_2), \tag{7}$$

$$\tilde{A}_1 \otimes \tilde{A}_2 = (l_1 l_2, m_1 m_2, u_1 u_2), \tag{8}$$

$$\tilde{A}_1^{-1} = \left(\frac{1}{u_1}, \frac{1}{m_1}, \frac{1}{l_1} \right). \tag{9}$$

The outcome of each set of pairwise comparisons is expressed as a fuzzy positive reciprocal (9) matrix: $\tilde{A} = \{\tilde{a}_{ij}\}$, such that if $\tilde{a}_{ij} = 1$ then $\tilde{a}_{ji} = 1/\tilde{a}_{ij}$ for all $i, j \leq n$, where n denotes the number of alternatives being compared within one set of pairwise comparisons, \tilde{a}_{ij} denotes the importance of alternative i over alternative j , \tilde{a}_{ji} denotes the importance of alternative j over alternative i , and $\tilde{a}_{ij} = (l_{ij}, m_{ij}, u_{ij})$.

The definitions of TFN values in Table 3 are used to facilitate the use of pairwise comparisons in fuzzy based decision processes. TFN (\tilde{a}) in Table 3 means that the decision maker

thinks the importance ratio of two alternatives is about a . In Table 3(a), particular linguistic assessment terms called fuzzy linguistic variables are introduced to represent the underlying fuzzy numbers employed for factor evaluations. For this study, five fuzzy linguistic variables are used to capture the subjective judgments about the relative importance of a factor versus another. The linguistic variables are summarized in Table 3(a) as equally important, moderately important, more important, strongly important, and extremely important. In Tables 3(a) and 3(b), a summary of the fuzzy linguistic variable set with lower, most likely, and upper values of underlying TFN and their definitions are given.

After the decision criteria have been determined as depicted in Figure 6(b), a judgment approach for determining the importance levels of these criteria is established. To evaluate the decision judgments, the decision makers select the related linguistic variables; then for calculations they are converted into the TFN and generalized for such analysis as presented in Table 3. The fundamental step in fuzzy AHP methodology is the prioritization procedure, whereby the prioritization problem is defined as deriving the unknown priority vector from the judgment set \tilde{A} . The basic procedure of the fuzzy AHP can be summarized in the following four main steps [33].

Step 1 (developing and structuring of the decision hierarchy). This first step comprises the restructuring the complex decision making problem into a hierarchical structure. The derived structural framework is significant in understanding the interactions amongst the elements involved in each decision level and aids the decision makers in exploring the impacts of the different decision components on the evaluation system.

Step 2 (building of the fuzzy pairwise comparison matrices). First, the relative importance of the criteria is determined through pairwise comparison of the decision criteria. After expert evaluations, the relative importance is transformed into triangular fuzzy numbers [25]. By considering a prioritization problem at a decision level with n elements, each set of comparisons for a given level requires $n(n - 1)/2$ judgments, which are used to construct a positive fuzzy reciprocal comparison matrix $\tilde{A} = \{\tilde{a}_{ij}\}$ as given in (6).

Step 3 (consistency check and fuzzy weights priorities derivation). This step checks for decision consistency and determines the priorities from the pairwise comparison matrices [25]. A fuzzy comparison matrix $\tilde{A} = \{\tilde{a}_{ij}\}$ is consistent if $\tilde{a}_{ik} \otimes \tilde{a}_{kj} \approx \tilde{a}_{ij}$, where $i, j, k = 1, 2, \dots, n$ [34]. After the consistency check, the fuzzy priorities \tilde{w}_i are calculated, from which the priority vectors $(w_1, w_2, \dots, w_n)^T$ are obtained from the comparison matrix by applying a prioritization ranking approach [31].

Step 4 (defuzzification—conversion of the fuzzy weights to crisp weights). Defuzzification can be solved using the graded mean integration representation (GMIR) method by Chen and Hsieh [27] and Hsieh et al. [35]:

$$\begin{aligned} \tilde{r}_i &= (\tilde{a}_{i1} \otimes \dots \otimes \tilde{a}_{ij} \otimes \dots \otimes \tilde{a}_{in})^{1/n}, \\ \tilde{w}_i &= \tilde{r}_i \otimes [\tilde{r}_1 \oplus \dots \oplus \tilde{r}_i \oplus \dots \oplus \tilde{r}_n]^{-1}, \end{aligned} \quad (10)$$

where \tilde{a}_{ij} is fuzzy comparison value of dimension i to criterion j , \tilde{r}_i is a geometric mean of fuzzy comparison value of criterion i to each criterion, and \tilde{w}_i is the fuzzy weight of the i th criterion, which can be indicted by a triangular fuzzy number $\tilde{w}_i = (lw_i, mw_i, uw_i)$ [35].

Step 5 (calculation and normalization of the weight vector and alternatives ranking). The crisp values are normalized in this stage. This final step aggregates the local priorities obtained at the different levels of the decision structure hierarchy into the composite priorities for the alternatives according to the weighted sum $A_i = \sum_{j=1}^n w_j a_{ij}$, where w_j is the weight of criterion j . The higher is the value of A_i , the more preferred is the alternative.

3.3. Fuzzy TOPSIS. The fundamental principle of the TOPSIS method is that the chosen alternative should have the shortest distance from the positive ideal solution (PIS) and the farthest distance from the negative ideal solution (NIS). If in a MCDM

problem decision process there are n criteria (C_1, C_2, \dots, C_n) amongst m alternatives (A_1, A_2, \dots, A_m) , then according TOPSIS principle, the performance of the alternative i at the criterion j , is such that the best alternative should have the shortest distance from the PIS and the farthest distance from the NIS. The TOPSIS considers the PIS and NIS as the reference points and does not take into consideration the relative importance of the distances from these points [36].

Fuzzy TOPSIS, where TOPSIS refers to Technique for Order Preference by Similarity to Ideal Situation, is a multicriteria decision evaluation method for selected criteria. The principle of TOPSIS approach is that an alternative that is nearest to the Fuzzy Positive Ideal Solution (FPIS) and farthest from the Fuzzy Negative Ideal Solution (FNIS) is chosen as optimal [37, 38].

The following steps give a summary of the fuzzy TOPSIS multicriteria decision making method [36, 39].

Step 1 (assignment of ratings to the criteria and the alternatives). If there are J possible candidates denoted by $A = \{A_1, A_2, \dots, A_j\}$, to be evaluated against n criteria $C = \{C_1, C_2, \dots, C_j\}$, and the criteria weights are denoted by w_i $\{i = 1, 2, \dots, m\}$, then the performance ratings of each decision maker D_k $\{k = 1, 2, \dots, K\}$ for each alternative A_j $\{j = 1, 2, \dots, n\}$ with respect to criteria C_i $\{i = 1, 2, \dots, m\}$ are denoted by $\tilde{R}_k = \tilde{x}_{ijk}$ $(i = 1, 2, \dots, m; j = 1, 2, \dots, n; k = 1, 2, \dots, K)$ with the membership function $\mu_{\tilde{R}_k}(x)$.

Step 2 (compute and aggregate the fuzzy ratings for the criteria and the alternatives). If the fuzzy ratings of all decision makers are described as triangular fuzzy number $\tilde{R}_k = (a_k, b_k, c_k)$, then the aggregated fuzzy rating is given by $\tilde{R} = (a, b, c)$, where $k = 1, 2, \dots, K$, where

$$\begin{aligned} a &= \min_k \{a_k\}, \\ b &= \frac{1}{K} \sum_{k=1}^K b_k, \\ c &= \max_k \{c_k\}. \end{aligned} \quad (11)$$

If the fuzzy rating and importance weight of the k th decision maker are $\tilde{x}_{ijk} = (a_{ijk}, b_{ijk}, c_{ijk})$ and the corresponding weights are $\tilde{w}_{ijk} = (w_{i1k}, w_{i2k}, w_{i3k})$, then the aggregated fuzzy ratings x_{ij} of the alternatives with respect to each criterion are given by $\tilde{x}_{ij} = (a_{ij}, b_{ij}, c_{ij})$, where

$$\begin{aligned} a_{ij} &= \min_k \{a_{ijk}\}, \\ b_{ij} &= \frac{1}{K} \sum_{k=1}^K b_{ijk}, \\ c_{ij} &= \max_k \{c_{ijk}\}. \end{aligned} \quad (12)$$

The aggregated fuzzy weights \tilde{w}_{ij} of each criterion are then calculated by $\tilde{w}_j = (w_{j1}, w_{j2}, w_{j3})$, where

$$\begin{aligned} w_{j1} &= \min_k \{w_{jk1}\}, \\ w_{j2} &= \frac{1}{K} \sum_{k=1}^K w_{jk2}, \\ w_{j3} &= \max_k \{w_{jk3}\}. \end{aligned} \quad (13)$$

Step 3 (compute the fuzzy decision matrix). The fuzzy performance/decision matrix for the alternatives (\tilde{D}) and the criteria (\tilde{W}) is constructed as follows. This is followed by the choice of the appropriate linguistic variables for the alternatives with respect to criteria:

$$\tilde{D} = \begin{matrix} & C_1 & C_2 & \cdots & C_n \\ \begin{matrix} A_1 \\ A_2 \\ \vdots \\ A_m \end{matrix} & \begin{bmatrix} \tilde{x}_{11} & \tilde{x}_{12} & \cdots & \tilde{x}_{1n} \\ \tilde{x}_{21} & \tilde{x}_{22} & \cdots & \tilde{x}_{2n} \\ \cdots & \cdots & \cdots & \cdots \\ \tilde{x}_{m1} & \tilde{x}_{m2} & \cdots & \tilde{x}_{mn} \end{bmatrix} \end{matrix}, \quad (14)$$

where $\tilde{x}_{ij} = (1/K)(\tilde{x}_{ij}^1 \oplus \cdots \oplus \tilde{x}_{ij}^k \oplus \cdots \oplus \tilde{x}_{ij}^K)$ with \tilde{x}_{ij}^k being the performance rating of alternative A_i with respect to C_j as evaluated by the k th decision maker, $\tilde{x}_{ij}^k = (l_{ij}^k, m_{ij}^k, u_{ij}^k)$ and $\tilde{W} = (\tilde{w}_1, \tilde{w}_2, \dots, \tilde{w}_n)$ [40].

Step 4 (normalize the fuzzy decision matrix). The raw data are normalized using linear scale transformation in order to bring the various criteria scales into a comparable scale. The normalized fuzzy decision matrix \tilde{R} is given by

$$\tilde{R} = [\tilde{r}_{ij}]_{m \times n}, \quad i = 1, 2, \dots, m; \quad j = 1, 2, \dots, n, \quad (15)$$

where $\tilde{r}_{ij} = (a_{ij}/c_j^*, b_{ij}/c_j^*, c_{ij}/c_j^*)$ and $c_j^* = \max_i c_{ij}$ are called the benefit criteria and $\tilde{r}_{ij} = (a_j^-/c_j^*, a_j^-/b_j^*, a_j^-/a_j^*)$ and $a_j^- = \min_i a_{ij}$ are called the cost criteria.

Step 5 (compute the weighted normalized matrix). The weighted normalized matrix (\tilde{V}) for criteria is computed by multiplying the weights (\tilde{w}_j) of evaluation criteria with the normalized fuzzy decision matrix r_{ij} :

$$\begin{aligned} \tilde{V} &= [\tilde{v}_{ij}]_{m \times n} = [\tilde{r}_{ij}(\cdot) w_{ij}], \\ & \quad i = 1, 2, \dots, m; \quad j = 1, 2, \dots, n. \end{aligned} \quad (16)$$

Step 6 (compute the fuzzy positive ideal solution (FPIS) and fuzzy negative ideal solution (FNIS)). The FPIS and FNIS of the alternatives are computed as follows. From the weighted normalized fuzzy-decision matrix, the elements \tilde{v}_{ij}

are normalized positive TFNs within the closed interval [0, 1]. Then, FPIS (A^+) and FNIS (A^-) are calculated [41]:

$$\begin{aligned} A^+ &= (\tilde{v}_1^+, \tilde{v}_2^+, \dots, \tilde{v}_n^+), \\ & \text{where } \tilde{v}_j^+ = \max_i \{v_{ij3}\}, \quad i = 1, 2, \dots, m; \quad j = 1, 2, \dots, n, \\ A^- &= (\tilde{v}_1^-, \tilde{v}_2^-, \dots, \tilde{v}_n^-), \\ & \text{where } \tilde{v}_j^- = \max_i \{v_{ij1}\}, \quad i = 1, 2, \dots, m; \quad j = 1, 2, \dots, n. \end{aligned} \quad (17)$$

Step 7 (compute the distance of each alternative from FPIS and FNIS). The distance (d_i^+, d_i^-) of each weighted alternative $i = 1, 2, \dots, m$ from the FPIS and the FNIS is computed as follows, using the area compensation method:

$$\begin{aligned} d_i^+ &= \sum_{j=1}^n d_v(\tilde{v}_{ij}, \tilde{v}_j^+) = \left[\frac{1}{3} \sum_{j=1}^n (\tilde{v}_{ij} - \tilde{v}_j^+)^2 \right]^{1/2}, \\ & \quad i = 1, 2, \dots, m, \\ d_i^- &= \sum_{j=1}^n d_v(\tilde{v}_{ij}, \tilde{v}_j^-) = \left[\frac{1}{3} \sum_{j=1}^n (\tilde{v}_{ij} - \tilde{v}_j^-)^2 \right]^{1/2}, \\ & \quad i = 1, 2, \dots, m, \end{aligned} \quad (18)$$

where $d_v(\tilde{a}, \tilde{b})$ is the distance measurement between two fuzzy numbers \tilde{a} and \tilde{b} .

Step 8 (compute the closeness coefficient (CC_i) of each alternative). The closeness coefficient (CC_i) or the degree of relative gaps that represents the distances to the fuzzy positive ideal solution (A^+) and the fuzzy negative ideal solution (A^-) is computed. The closeness coefficient of each alternative is calculated as

$$\begin{aligned} CC_i &= \frac{d_i^-}{d_i^- + d_i^+} = \left(1 - \frac{d_i^+}{d_i^- + d_i^+} \right), \\ & \quad \text{for } i = 1, 2, \dots, m, \end{aligned} \quad (19)$$

where $d_i^-/(d_i^- + d_i^+)$ is the fuzzy satisfaction degree in the i th alternative and $d_i^+/(d_i^- + d_i^+)$ is the fuzzy gap degree in the i th alternative.

Step 9 (ranking the alternatives). In the last step, the different alternatives are ranked according to the closeness coefficient (CC_i) in decreasing order. The best alternative is closest to the FPIS and farthest from the FNIS.

4. Case Study Results and Analysis

4.1. GIS-SQL Database for Thematic Displays of Pavement Distress Data. Figure 7 shows the typical output of the results for the GIS-SQL based graphical thematic display of the road condition survey for the Nandi road distress between segments 2100 m and 2400 m. The segment query results

TABLE 4: Case study maintenance functions for the observed distresses.

Distress treatment (maintenance function) for observed distress	Fog seal	Micro surfacing	Slurry seal	Cape seal	Chip seal	Thin HMA overlay	Resurfacing
Rutting		×				×	
Alligator and block cracking	×	×		×	×	×	×
Longitudinal and transverse cracking		×		×	×	×	
Pothole		×				×	×
Raveling	×	×	×	×	×		
Resurfacing		×				×	×

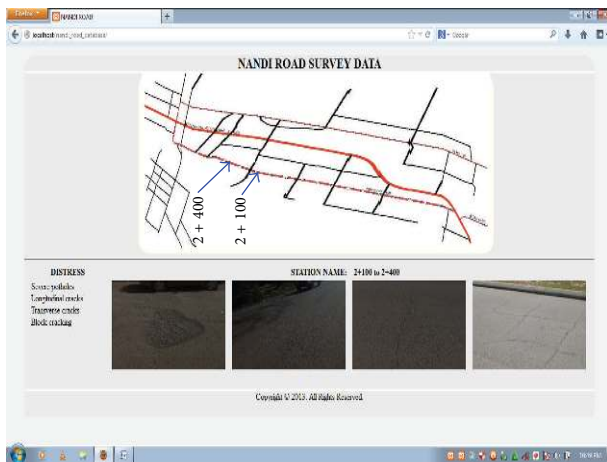


FIGURE 7: An SQL based thematic graphical display of road pavement distress at specified segments.

display the following information regarding the segment: the chainage of the road segment surveyed, the distresses observed within the segment, and the distress images. Such pavement condition data representation can easily be leveraged into a web access platform, so as to be easily accessible to the different levels of urban road maintenance and management.

A summary statistic on the frequency of occurrence of distress types is also generated for the entire surveyed road pavement as presented in Figure 8. Results in Figure 8 show the most and least frequent distresses on the road as longitudinal cracks and patching, respectively. A plot of the degree of distress that is low, medium, or high can also be derived from the empirically observed and categorized severity of distresses, though such linguistic information is captured in the distress survey by the condition assessment experts.

4.2. Case Study Road Maintenance Treatments. From the observed distresses presented in Table 1 and in Figure 5, the possible maintenance treatments and functions are determined, with the results presented in Table 4. From the condition survey results, it was observed that there were seven main pavement distresses or maintenance functions along the 3 km road. From the results in Table 4 it was noted that if the distresses identified in the pavement condition survey are related to structural deficiencies, then the road

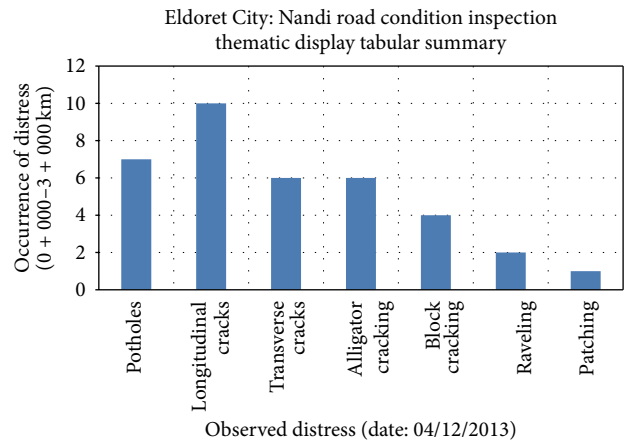


FIGURE 8: Summary of the distresses report of for the case study.

section is most likely not to be a candidate for a preventive basic maintenance treatment but rather for rehabilitation.

The results in Table 4 are summarized as follows: (i) fog seal: it is an application of diluted emulsion (normally 1 to 1) to enrich the pavement surface and hinder raveling and oxidation and is considered as a temporary application; (ii) crack sealing: this treatment is used to prevent water and debris from entering cracks in the pavement; the treatment might include routing to clean the entire crack and to create a reservoir to hold the sealant; (iii) chip seal: this treatment is used to waterproof the surface, seal small cracks, and improve friction; although it is typically used on low volume roads and streets, it can also be used on high volume roads and highways; (iv) thin cold mix seals: these treatments include slurry seals, cape seals, and micro surfacing which are used on all types of facilities to fill cracks, improve friction, and improve ride quality; (v) Thin Hot Mix Asphalt (HMA) overlays: these include dense-, open-, and gap-graded mixes (as well as surface recycling) that are used to improve ride quality, provide surface drainage and friction, and correct surface irregularities; Thin HMA overlays are generally 37 mm in thickness; and (vi) resurfacing: resurfacing of a road pavement in totality or section thereof is normally carried out in order to add strength to the asphalt layer and to prolong its life and or to correct a specific section or the entire surface profile and in order to improve on the riding quality and surface water drainage.

TABLE 5: Round 1 of decision based on importance of allocation or weighting.

Road pavement maintenance objectives	Pavement expert (engineer/inspector/planner)										Maintenance relative weights
	1	2	3	4	5	6	7	8	9	10	
Road safety	35	45	45	30	40	45	35	50	30	35	0.39
Pavement surface preservation	40	30	20	40	30	25	30	20	30	40	0.31
Road operational status and standards	15	15	20	20	15	20	25	20	20	15	0.19
Road aesthetics	10	10	15	10	15	10	10	10	10	10	0.11

TABLE 6: Round 2 of judgment based on incremental linear ranking with 1.0 as the lowest importance.

Road pavement maintenance objectives	Pavement expert (engineer/inspector/planner)										Maintenance relative weights based on linear scale
	1	2	3	4	5	6	7	8	9	10	
Road safety	3	4	5	5	4	2	2	2.5	4	4	0.36
Pavement surface preservation	2.5	3	3	4	4	3	2	4	3	2	0.30
Road operational status and standards	2	3	3	4	3	1.5	2	3	3	1	0.25
Road aesthetics	1	1	1	1	1	2	1	1	1	1	0.12

TABLE 7: Round 3 of decision according to crisp scale based ranking (with 1 being the least important and 5 the most important).

Road pavement maintenance objectives	Pavement expert (engineer/inspector/planner)										Maintenance relative weights based on 1–5 crisp scale
	1	2	3	4	5	6	7	8	9	10	
Road safety	5	5	4	5	5	4	5	5	5	5	0.39
Pavement surface preservation	4	4	3	4	5	4	4	5	4	4	0.32
Road operational status and standards	3	2	3	3	4	3	2	3	2	3	0.22
Road aesthetics	1	1	1	1	1	2	1	1	1	1	0.07

4.3. *Maintenance Objectives Determination, Weighting, and AHP Hierarchical Structure.* To determine the maintenance objectives, from the maintenance functions, ten experts comprising pavement engineers, inspectors, and transport planners from Kenya Urban Roads Authority (KURA) gave their assessments and judgments on the maintenance objectives weighting for routine urban road maintenance. This was carried out by first selecting the maintenance objectives, where the experts were asked to nominate a number of maintenance objectives based on the observed distresses (Table 4). From the expert evaluation results, the top four maintenance objectives are adopted as the case study maintenance criteria.

The selected road maintenance objectives were determined, respectively, as road safety (RS); pavement surface preservation (PSR); road operational status and standards (ROS) or quality; and road aesthetics (RA) (Table 5). After the maintenance objectives are obtained, the weights of the selected maintenance objectives are then determined.

The same group of experts gave their estimates on the four maintenance objectives weighting. Each of the experts assigned weights to the maintenance objectives in three rounds, each round using a different approach comprising the following elements. The rationale of using the three rounds is to vary and objectively, without biases, determine the weights based on different ranking scales, from percentages, incremental decimals linear scale (1.0–5.0), and whole number linear scale (1–5).

- (i) Round 1: in this round weighting is carried out in terms of percentages and applied to each maintenance objective.

- (ii) Round 2: by using linear incremental scale with steps of 0.25, the objective with the lowest importance is set at 1.0 and the one with the greatest importance is recorded at 5.0, depending on the function.

- (iii) Round 3: on a linear scale of crisp values of 1–5 (increasing at intervals of 1), the importance of the objectives is ranked with 1 representing the least important objective and 5 representing the most important.

Because of human objectivity, reasoning, and subjectivity, at the end of three rounds, the results will vary based on the individual decision maker's perception of the criterion and in accordance with ranking scale used. At the end, the results are aggregated to minimize on subjectivity and to objectively determine the overall weights. Results for the three ranking steps are presented in Tables 5–7 for the four maintenance objectives. In the pavement expert columns, 1 to 10 represent the ten decision experts.

From the averaged AHP and fuzzy weights based evaluation matrices represented, respectively, in Tables 8 and 9, the final aggregated weighting vector was determined as $W = [0.37, 0.31, 0.22, 0.10]^T$ for the four maintenance objectives and is presented in Table 10. The results in Table 10 show that, according to the experts, road safety and surface condition accounted for 68% of the maintenance objectives preferences, while quality and aesthetics together contributed to 32% of the maintenance preferences.

From the above results the final AHP hierarchical structure of the maintenance prioritization levels is derived for

TABLE 8: Averaged AHP based evaluation matrix for the maintenance objectives.

Road maintenance objectives	Road safety	Pavement surface preservation	Operational status and standards	Road aesthetics
Road safety	1	2	1	2
Pavement surface preservation	0.5	1	0.5	0.75
Operational status and standards	1	2	1	1
Road aesthetics	0.5	1.33	1	1

TABLE 9: Equivalent fuzzy based evaluation matrix for maintenance objectives.

Road maintenance objectives	Road safety	Pavement surface preservation	Operational status and standards	Road aesthetics
Road safety	(1, 1, 1)	(1.33, 2, 4)	(0.8, 1, 1.33)	(1, 2, 3)
Pavement surface preservation	(0.33, 0.5, 0.75)	(1, 1, 1)	(0.25, 0.5, 1)	(0.5, 0.75, 1)
Operational status and standards	(0.75, 1, 1.25)	(1, 2, 3)	(1, 1, 1)	(0.5, 1, 1.25)
Road aesthetics	(0.33, 0.5, 1)	(1, 1.33, 2)	(0.8, 1, 1.33)	(1, 1, 1)

TABLE 10: Average maintenance objectives weighting vector (W) for the case study.

Road maintenance objectives	Road safety	Pavement surface preservation	Operational status and standards	Road aesthetics
Mean relative maintenance objective weights	0.37	0.31	0.22	0.10

the case study as presented in Figure 9. As already stated, hierarchical structuring of a pavement maintenance prioritization problem is an efficient way of dealing with the decision complexity and is useful in identifying the major components of the PMMS. From Figure 9, it is seen that the structure for the case study comprises three levels, with the final level being the overall goal, and the other two levels: the first, lower hierarchy, is the maintenance functions, and the second, higher hierarchy, is the maintenance objectives. For realistic cost-benefit evaluation, every M&R project should have its own hierarchical structure.

4.4. Multiattribute Analysis for Pairwise Comparisons of Maintenance Objectives. For the pairwise comparative evaluation, ten pavement engineers from KURA gave their judgments using pairwise comparisons of the four maintenance objectives. All the ten engineers gave pairwise comparisons to the seven maintenance functions with respect to the predetermined four maintenance objectives. The pairwise comparisons are based on nine-point Saaty scale of relative importance [26] as presented in Table 2. Using the fuzzification and defuzzification rules as presented in Table 3, substitutes for the crisp numbers and the fuzzy numbers are obtained.

The optimal level of confidence is the degree of accuracy with which one is able to execute a given decision making task. That is, due to the subjective nature of the exercise, an optimal level from experienced researchers is empirically considered to be neither too high nor too low, and therefore a value above mean threshold of 60% is chosen. On the other hand, the attitude towards risk is the effect of executing the particular task within an error margin. For this case study, the error margin for the decision makers is set a maximum of 50%

chance, and the errors are attributed to mistakes (gross), systematic and random occurrences. This means that an optimal performer should be able to execute the task with an accuracy of at least 60% and is liable to maximum error level of 50%.

By choosing an optimal level of confidence of ($\alpha = 60\%$) and assuming a moderate attitude to risk of ($\lambda = 0.5$), the fuzzy performance matrix (Z) is defuzzified, under the optimal boundary conditions of [$\alpha = 0.6, \lambda = 0.5$]. The matrix is normalized to satisfy the condition that the summation of all the seven maintenance functions under each criterion is 1. Therefore the crisp normalized performance matrix Z_{α}^{λ} is obtained as below for the four pavement maintenance objectives and as determined by the ten experts:

$$Z_{\alpha}^{\lambda} = \begin{matrix} & \begin{matrix} \text{RS} & \text{PSR} & \text{ROS} & \text{RA} \end{matrix} \\ \begin{pmatrix} (0.199) & (0.428) & (0.349) & (0.349) \\ (0.464) & (0.415) & (0.377) & (0.064) \\ \dots & \dots & \dots & \dots \\ (0.170) & (0.142) & (0.161) & (0.229) \end{pmatrix} & (20) \end{matrix}$$

The final maintenance functions ranking is then derived and the results are presented in Tables 11 and 12 for fuzzy AHP and fuzzy TOPSIS. The final performance index for each maintenance function is obtained by averaging the respective priorities as determined by each of the decision makers. It is observed that the results in Tables 11 and 12 are nearly the same except for the rankings 5 and 6. Compared with the manual-based evaluations and ranking, which are based on the set out guiding principles such distress type, the dimensions (area, length, and depth), and significance of the road, for the aggregation of individual ranking of each

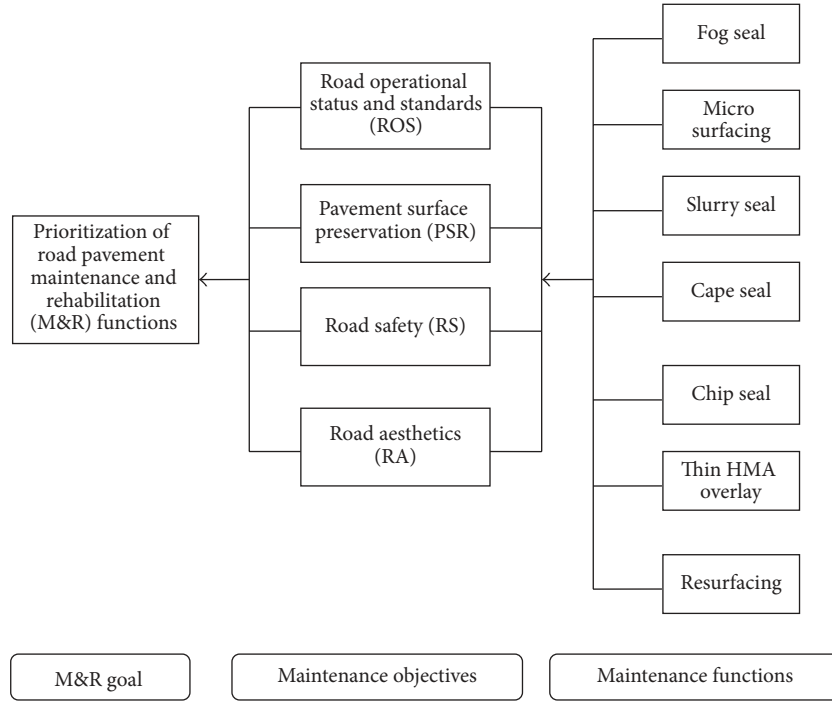


FIGURE 9: Three-level hierarchical structure of maintenance function prioritization process for the case study road.

TABLE 11: Maintenance function ranking and performance ranking using fuzzy AHP.

Function number	Maintenance function	Performance index	Ranking index
1	Fog seal	0.755	7
2	Micro surfacing	0.798	6
3	Slurry seal	0.870	3
4	Cape seal	0.816	5
5	Chip seal	0.852	4
6	Thin HMA overlay	0.883	1
7	Resurfacing	0.879	2

distress and scoring, the fuzzy TOPSIS results in Table 12 were found to be consistent with the prioritization experiences in practice. The top three maintenance functions (Thin HMA overlay, resurfacing, and slurry seal) are all pavement related functions and are related to the pavement deficiencies and have more direct impacts on the road pavement functions.

5. Comparison of Fuzzy AHP and Fuzzy TOPSIS Results

To compare variance in the decisions, the fuzzy AHP and fuzzy TOPSIS were plotted on the same scale, by rescaling the performance indices of all the seven maintenance functions to the same scale such that the normalized sum equals 1. Figure 10 presents the results of the comparative ranking

indices. The results in Figure 10 show that the fuzzy AHP tends to overestimate the ranking process as compared to the fuzzy TOPSIS and the manual-based decision validation procedure. The results in Figure 10 show that the fuzzy TOPSIS results are quite close to manual-based approach, which implies that the TOPSIS approach is much more able to capture the user degree of confidence. Nonetheless, different case studies can be carried out in order to generalize the effectiveness of two fuzzy based prioritization techniques, especially in different geographical regions.

There are however similarities and differences in the rankings as obtained from the two prioritization techniques as seen in Figure 10. For example, the top three maintenance functions are the same and only in the bottom three is there a slight difference. The results also show that there is consistency in the fuzzy TOPSIS ranking process, as compared to the variability in ranking values in fuzzy AHP.

Although fuzzy AHP and fuzzy TOPSIS have the same objective of prioritizing the maintenance objectives and functions through MCDM principles, they have fundamental differences. According to fuzzy AHP, pairwise comparisons by decision makers are made for the criteria and alternatives under each criterion. The resulting comparisons are integrated and decision makers' pairwise comparison values are transformed into triangular fuzzy numbers. The priority weights of the criteria and alternatives are then derived. According to the combination of the priority weights of criteria and alternatives, the best alternative is determined.

In fuzzy TOPSIS, the decision makers use the linguistic variables to evaluate each alternative with respect to each criterion in order to assess the importance of the criteria. The linguistic variables are converted into triangular

TABLE 12: Maintenance function ranking and performance index ranking using fuzzy TOPSIS.

Function number	Maintenance function	$D_{\alpha}^{\lambda+}$	$D_{\alpha}^{\lambda-}$	Separation score CC_i	Ranking
1	Fog seal	0.0580	0.0220	0.2750	7
2	Micro surfacing	0.0497	0.0485	0.4255	5
3	Slurry seal	0.0310	0.0270	0.4655	3
4	Cape seal	0.0459	0.0340	0.3921	6
5	Chip seal	0.0378	0.0291	0.4350	4
6	Thin HMA overlay	0.0347	0.0420	0.5480	1
7	Resurfacing	0.0432	0.0477	0.5248	2

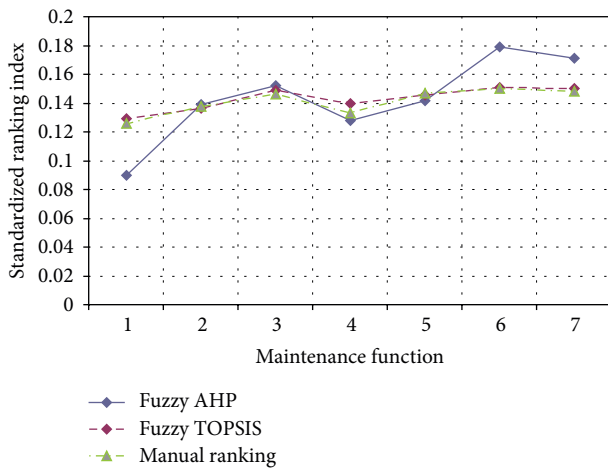


FIGURE 10: Comparison of maintenance prioritization results from fuzzy AHP and fuzzy TOPSIS.

fuzzy numbers, from which the fuzzy decision matrix is determined. The normalized fuzzy decision matrix and the weighted normalized fuzzy decision matrix are then formed. After determining the FPIS and FNIS, the distance (d) of each alternative to FPIS and FNIS is also determined. Therefrom the closeness coefficient of each alternative is individually computed. From the seven alternatives, the closeness coefficient revealed the results in Table 12 for the fuzzy TOPSIS and in Table II for the fuzzy AHP.

The fundamental differences between these two methods could be based on the fact that TOPSIS considers the linguistic variables at the onset of the pairwise comparison and therefore the results are not affected by the AHP drawbacks in fuzzy AHP. Nonetheless, according to the results, the top three results were similar for both methods. This in part shows that, apart from the subjectivity that is not critically captured by AHP in the pairwise comparison process, both TOPSIS and AHP as multicriteria decision making techniques are complete in performance. But, as presented in Figure 10, the subjectivity in AHP is noticed when the results are compared on standardized scale.

In overall, the appropriate preventive maintenance strategy will largely be influenced by the type, severity, and extent of the pavement surface distresses and the structural and functional condition of the pavement. In this regard,

the influences of the categories of pavement maintenance and the performance of preventive maintenance treatments, as recommended by Johanns and Craig [42], should be taken into consideration in balancing between preventive, corrective, and emergency maintenances schedules.

6. Conclusions

The current case study presents a methodological overview on the use of multiattribute decision making using fuzzy AHP and fuzzy TOPSIS for the prioritization of pavement maintenance alternatives. By using different sets of fuzzy membership functions, the current approach enables decision makers to capture the often unpredictable and complex pavement maintenance prioritization procedure, in a more objective way.

For the case study, it is observed that maintenance measures relate to the following pavement failures: cracking, potholes, raveling, and patching. From the prioritization results, and with manual-based opinion, it is concluded that the top three maintenance objective functions in the case study urban roads are characterized by Thin HMA overlay, resurfacing, and slurry seals, for the following pavement maintenance objectives: road safety conditions, pavement surface preservation, road operational status and standards, and the road aesthetics. The results show that road safety is closely related to the actual pavement preservation status, and both contribute in nearly the same ratio to more than 60% of the maintenance objectives.

The ranking procedures yielded nearly similar results, with fuzzy TOPSIS performing slightly better than fuzzy AHP, when the results were compared against actual empirical-manual-based prioritization. Further, the results show that, at optimal degrees of confidence and risk, the fuzzy AHP tended to overestimate the prioritization ranking process, as compared to the fuzzy TOPSIS approach.

To improve on the current study, it is recommended that automated distress surveys and measurements be considered. This is important in improving on the speeds of condition surveys, accuracy and reliability, and cost-effectiveness. Developments on pavement maintenance prioritization approach that can enable decision makers to quantify and minimize the total risk due to canceling or deferring maintenance functions during budget cuts should also be considered in the prioritization process. Additionally, studies on the

spatial-temporal modelling of the rate of pavement deterioration are recommended in order to integrate the maintenance prioritization approach with the asset's whole-life economic analysis. And, in addition, detailed studies on the variations and relationships between the decision makers' level of confidence and attitude to risk in multicriteria decision making process can also be a detailed subject of investigation.

Conflict of Interests

The authors declare that there is no conflict of interests regarding the publication of this paper.

Acknowledgments

The authors would like to acknowledge all those who were involved in the field surveys on the road condition inspection project. Further acknowledgements go to the contributions in terms of comments by the anonymous reviewers who took their time to pinpoint key areas to improve on in this paper.

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