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Hybrid Hesitant Fuzzy Multi-Criteria Decision Making Method: A Symmetric Analysis of the Selection of the Best Water Distribution System

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Abstract: Every country's influence and livelihood is centered on that country's water source. Therefore, many studies are being conducted worldwide to improve and sustain water resources. In this research paper, we have selected and researched the water scheme for groundwater recharge and drinking water supply of drought prone areas. The water project is aimed at connecting the drought prone areas of the three districts of Tamil Nadu to filling up the ponds in their respective villages and raising the ground water level and meeting the drinking water requirement. We have chosen a multi-criteria decision method to select the best alternative in a complex situation. When reviewing the implementation of this water project, many experts and people who will benefit from this project may have some hesitation and ambiguity in their suggestion on choosing the best water distribution system. We believe that the benefits of this project can be fully availed of if we choose a water distribution system. Our contribution in this article is to choose the best water distribution system for this project by use of our proposed multi-criteria decision making (MCDM) methods, hesitant fuzzy standard deviation with multi-objective optimization method by ratio analysis (HFSDV-MOORA), hesitant fuzzy standard deviation with technique, for order preference by similarity to an ideal solution (HFSDV-TOPSIS) and hesitant fuzzy standard deviation with Visekriterijumsko Kompromisno Rangiranje (HFSDV-VIKOR), which will provide the best solution for improving the water resource for the drought-prone areas of three districts. Finally, we have identified and compared the correlation coefficient between proposed methods. As a result of the study, it has been found that the best water supply system is closed concrete pipes laid along agricultural land through the rural areas.

Keywords: MCDM; hesitant fuzzy number; water irrigation; groundwater recharge

1. Introduction

In agricultural production, soil, seed and water are basic requirements. Additionally, sunshine, insecticides, fertilizers, suitable climates and human labors are also needed. Of all these, water appears to be the most important requirement of agricultural management. The main impact of irrigation is in terms of the increased agricultural yield. These results affect different human environments, like social, cultural, economic and political, besides the agricultural development. Therefore, many studies are being conducted worldwide to improve and sustain water resources.

In India, the irrigated area consists of about 36% of the net source area. Presently, the agricultural sector account for about 83% of all water uses. Tamil Nadu covers 4% of the geographical area (13.01 million hectares) and caters to 5.96% of the population of the country with 7.21 crores people living along the 17 river basins. Tamil Nadu is a water stressed state and depends heavily on ground water for irrigation. There are various types of systems of irrigation practices in different parts of Tamil Nadu. Direct irrigation method and storage irrigation method, various types of irrigation techniques differ in the way that the water is obtained from the source and distributed within the field. In general, the goal is to supply the entire field uniformly with water, so that each plant has the amount of water it needs, neither too much nor too little. In olden days, irrigation in Tamil Nadu is carried out through gravity irrigation and lift irrigation.

In Tamil Nadu, ground water is considered as a main resource for irrigation. Here, the canal irrigation system is an important primary need of agriculture. It increases the groundwater level. Nowadays in India, ground water is used in the agriculture sector. In drought prone areas, the groundwater level is decreased, because of the rise in the population, development of industries, illegal well pumping, etc. These facts, along with recent drought, led to severe decrease of the groundwater level, especially the drought prone areas. In order to reduce this problem, the initiation of groundwater recharge research can be suitable solution. The water distribution system will help recharge ground water table and benefit farmers in Coimbatore, Erode and Tirupur districts.

2. Literature Review

There are several methods followed in MCDM techniques. See (Table 1), a brief account of some important studies in MCDM methods related to our proposed methods.

Table 1. Literature study of selected MCDM methods.

Authors	Study of MCDM Methods	Applications
Arabsheybani et al. (2018) [1]	An integrated fuzzy MOORA and FMEA technique	Supplier selection
Brauers and Zavadskas, (2006) [2]	MOORA method	Privatization in a transition economy
Karande and Chakraborty, (2012) [3]	Fuzzy-MOORA method	ERP system selection
Cevik Onar et al. (2014) [4]	HF-TOPSIS and Interval Type-2 Fuzzy AHP	Strategic decision selection
Sun and Ouyang, (2015) [5]	HFMAADM Based on TOPSIS With Entropy	An energy police selection
Zhang et al. (2016) [6]	FMCGDM: IVDHF-TOPSIS and IVDHF-VIKOR	Community development
Ozcelik et al. (2014) [7]	Hybrid MOORA-Fuzzy	Special education center selection
Buyukozkan and Gocer, (2017) [8]	An extension of MOORA based on IVIFN	Digital supply chain
Santawy and Ahmed, (2012) [9]	SDV-MOORA	Analysis of project selection
Achebo and Odinikuku, (2015) [10]	SDV and MOORA	Gas metal arc welding process
Myers and Sirois, (2006) [11]	Spearman correlation coefficients	A general view
Beg and Rashid, (2013) [12]	HF-TOPSIS with linguistic	Investment company
Tavana et al. (2018) [13]	An extended VIKOR	Decision maker's risk
Suh Park et al. (2019) [14]	Integrated weighting fuzzy VIKOR	Mobile services
Zhang and Wei, (2013) [15]	Extension of VIKOR method	Enterprises projects
Narayanamoorthy et al. (2019) [16]	IVIHF entropy based VIKOR	Industrial robots selection
Narayanamoorthy et al. (2019) [17]	HF-MAUT with HF-CRITIC method	Water management
Narayanamoorthy et al. (2019) [18]	Normal wiggly dual hesitant fuzzy sets	Energy management
Geetha et al. (2019) [19]	HF-MULTIMOORA method	Healthcare waste management
Narayanamoorthy et al. (2020) [20]	HF-MOOSRA with HF-SOWIA method	Bio-medical waste management
Nadaban et al. (2016) [21]	Fuzzy TOPSIS	Team alternative selection
Senvar et al. (2016) [22]	Hesitant Fuzzy TOPSIS	Hospital Site Selection

Multi-criteria decision making (MCDM) is the process of selecting the most satisfactory optimal solution from a set of assorted types of criteria. One of the extension of fuzzy set is known as HFS, here decision makers psychological behavior are taken in to account when giving membership function. HF was introduced by Torra. Researchers have recently developed and made decisions that make decisions that are unclear and uncertainties in order to find solutions to practical issues recently. Kuang et al. (2015) [23] approached grey numbers to handling uncertainty in MCDA. Zhang et al. (2019) [24] described a MCDM problem in a fuzzy covering approximation space by the method of TOPSIS. Zhan et al. (2018) [25] contributed to the field of fuzzy set theory by proposing some methods that benefit from some novel fuzzy rough set models. Chang, (2014) [26] studied VIKOR method

with linguistic variables parameterized by triangular fuzzy numbers for evaluating the hospital service quality under a fuzzy environment. In multi-objective optimization technique, MOORA procedure is one of the simplest method for selecting best alternative Stanujkic (2014) [27]. Chen, (2000) [28] proposed the compromised solution of fuzzy decision matrix in linear case. Farhadina, (2013) [29] proposed several types of ranking method for hesitant fuzzy values for multiple attribute decision making.

Aggarwal, (2015) [30] proposed new weight finding technique, here weights are calculate through its preferences. Torra, (2010) [31] introduced Hesitant fuzzy set and its properties. Xu et al. (2011) [32] developed some important aggregation operators and distance measure for aggregating the fuzzy information in MAGDM problems by use of HFS. Wu et al. (2018) [33] developed compromise solutions for MAGDM under HFLTS. Xia and Xu, (2011) [34] developed some aggregation operator for correlation of HFS. Kaya et al. (2019) [35] reviewed some MCDM methodologies in energy policy making. Moradian et al. (2018) [36] compared some MCDM methods MOORA, TOPSIS and VIKOR.

In water resource management various mathematical methods are developed with fuzzy decision making, like an interval-based fuzzy chance constrained have been developed for water allocation with double-sided fuzziness by Lee (2000) [37]. In water management, water supply is an important role and scope of improvement in farmers water management decisions. Nowadays many water-saving irrigation models have been developed extremely in many areas. In agricultural water productivity, improving irrigation efficiency is becoming more important, therefore shallow groundwater plays an important role in irrigation water productivity. Some studies have reported a number of problems in irrigation districts. Irrigation has been developed as a strategy of reduce poverty and improve household income. Bjornlund et al. (2019) [38] found the relationship between household decision-making and farm household in small-scale irrigation schemes in southern Africa. In the field of water management, irrigation district managers and farmers used few useful tools like artificial intelligence techniques to forecast irrigation depths at farm level in pressurized irrigation networks project investment. In many water resources problems, DM is the process of choosing and deciding the water supply scheme under the technical and operation risk. Fu et al. (2009) [39] used fuzzy comprehensive model based on entropy weight to project investment. In order to reduce the water demand in drought prone areas, a hybrid TOPSIS-agent-based model is developed to find a solution for this problem (Ghazali et al., 2018) [40].

At present, Tamil Nadu Government is taking new initiatives to provide electricity to the irrigation of agricultural produce in the drought prone areas. It is therefore imperative to analyze the water distribution system for these reasons and to choose the best water distribution system and sort the selected ones according to their importance of criteria. Hence irrigation management is aimed at obtaining electricity from renewable energy, which will be available throughout the year (Senjyu et al., 2005) [41].

In recent times, floods and droughts in various parts of the world are causing serious problems (Ding et al., 2014) [42]. Donaldson, (2013) [43] developed a modernization of irrigation scheme. Opricovic and Tzeng, (2004) [44] addressed the water management problem and gave the compromise solution of MCDM problem by the compared analysis of VIKOR and TOPSIS method. The reutilization of treated waste water for irrigation purposes is becoming more widespread in many drought prone places around the world. In varies countries, various water bodies and ground water irrigation companies have identified drought-prone areas in different districts. In all countries, many water saving irrigation models have been developed extremely in many areas. For example, in China, irrigation water management evaluated and released the new model for analyzed the investment decision making of water saving irrigation project (Fu, 2007) [45]. In order to evaluate the ground water resources, the mathematical tool of Grey relational model based on AHP weight is used for applied to evaluate ground water resources carrying capacity of irrigation district (Gu and Xu, 2011) [46]. It is desirable to create a sustainable water management system that will integrate multi-objective restructuring operation in balancing the water demand and balance of water concentration project in

drought-prone areas. In recent years, there are several fuzzy control technology was separately studied in irrigation systems (Zhang et al., 2018; Walker et al., 2015; Jiang et al., 2020 [47–49]). Therefore, in this research paper, we choose the HFMCMD method has been adopted as a best method in irrigation water management systems.

3. Motivation and Contributions

3.1. Motivation and Objective

Generally, in a water management plan, the selection of projects that involve a number of criteria is implemented under the problem solving section of MCDM. The motivation of this research paper is to provide these decision makers with potential value when making their decision-maker role in a problem-solving environment. These potential values represent the hesitant ambiguous elements of the decision maker. Therefore, to cope with the complexity of the many criteria in this complex problem, which are implemented to meet groundwater and drinking water requirements for drought-prone areas, the decision-maker is allowed to give his or her contribution by several possible values. Therefore, our main motivation of this research paper is to determine the best water supply system through MCDM system with hesitant fuzzy situation. Moreover, in this research paper, we extend the SDV weight finding method in hesitant fuzzy environments, and propose the three MCDM methods with this extended standard weight detection method to find the weights of the five criteria described in the review of literature, since the five weights are selected in hesitant fuzzy environments. In well-known and practiced multi criteria decision-making methods, weight detection methods with the standard deviation method did not consider the hesitant fuzzy situation. In order to fill this gap found in the literature, the present research article proposed three multi-criteria decision making systems with hesitant fuzzy scenarios: HFSDV-MOORA, HFSDV-TOPSIS and HFSDV-VIKOR. We have also analyzed the two cases in terms of the importance of the criteria in the sensitivity analysis.

3.2. Contribution

Our main contribution of this work is that it can capture hesitation thoughts of decision maker. Choosing the best water supply system for drought prone areas involves problem solving methods. In choosing the best water supply scheme, a special mathematical set theory is required in MCDM technique to give value to the hesitant and ambiguous thoughts of the decision maker. For our contribution to this article, we have chosen the hesitant fuzzy set introduced by Tora, (2010) [31] and used it in the selection of water supply for drought prone areas along with the MCDM system. We have used a hesitant fuzzy set with the SDV weight finding technique in the weighting area, which is considered an important part of the MCDM method used here. Another important part of the MCDM system is the use of the hesitant fuzzy set in the MOORA mode in the ranking area. Therefore, our contribution includes the HFE in the selection of the best water supply system.

- The well-known MCDM techniques with hesitant fuzzy environment: MOORA, TOPSIS and VIKOR are used to select the best water distribution system of drought-prone areas.
- The first proposed method is to weights the criteria via extended standard deviation weighting technique, called as HFSDV-MOORA.
- The second proposed method uses the extended standard deviation weighting technique to assign weights to the criteria under HFSDV-TOPSIS.
- The third proposed method uses the extended standard deviation weighting technique to assign weights to the criteria under hesitant fuzzy environment in order to give a compromise solution, because it provides a maximum “group utility” of the “majority” and a minimum of the individual regret of the “opponent” called as HFSDV-VIKOR.
- The ranking result concluded that the proposed SDV-MCDM methods have outperformed the existing WEM-MCDM in the literature review by giving compromise and best solution.

- The comparison study shows that the proposed three methods (SDV-MOORA, TOPSIS, VIKOR) with hesitant fuzzy environment have perfect ranking.
- The sensitivity analysis is to analyze the ranking order when changing the significance of objective weights by making two cases.
- Testing and analyzing the two cases of sensitivity analysis with our proposed MCDM methods and provide the ranking order of comparison study of this sensitive part. And also test the correlation co-efficient between the proposed HFSDV-MOORA, HFSDV-TOPSIS and HFSDV-VIKOR by use of Spearman's rank correlation method.
- Our methods provide the best alternative water distribution system for the ground water irrigation scheme. The solution is rectifying the water problem in drought-prone areas of selected three district. Moreover, we compare our proposed SDV hesitant fuzzy MOORA method with some research articles Li, (2014); Xu, (2013); Liao, (2013) [50–52]. And also compared our proposed weight finding results with entropy weights.

This research paper, we make water distribution system as a numerical problem for selecting the best alternative by use of HF-MCDM method. In Comparison section, we analysis our result with the correlation coefficients results. Discussion and conclusion are given in the final section.

4. Preliminaries

Definition 1. Let T be a finite reference set, a HFS A on T is in terms of a function $h_H(t)$ when applied to T returns a subset of $[0, 1]$ Torra (2010), which is described as the follows,

$$A = \{ \langle t, h_H(t) \rangle / t \in T \}$$

where $h_H(t)$ is a set of some values, which is in $[0, 1]$, denoting the possible membership degrees of the element $t \in T$ to the set A ; Here, hesitant fuzzy element (HFE) is denoted as $h_H(t)$.

Example 1. Let $T = \{t_1, t_2, t_3\}$ and let $h_H(t_1) = H\{0.2, 0.6, 0.9\}$, $h_H(t_2) = H\{0.75, 0.8, 0.85\}$ and $h_H(t_3) = H\{0.1, 0.8, 0.9\}$ considered as the three hesitant fuzzy elements of $t_i (i = 1, 2, 3)$ to a set H . Here, hesitant fuzzy set (HFS) is described as follows,

$$H = \{ \langle t_1, H\{0.2, 0.6, 0.9\} \rangle, \langle t_2, H\{0.75, 0.8, 0.85\} \rangle, \langle t_3, H\{0.1, 0.8, 0.9\} \rangle \}$$

Definition 2. Given three HFEs represented by h, h_1, h_2 are define by Torra (2010) and Xia and Xu (2011) the following operations:

$$\begin{aligned} (i) \quad h^c &= \bigcup_{\gamma \in h} \{1 - \gamma\}; \\ (ii) \quad h_1 \cup h_2 &= \bigcup_{\gamma \in h_1, \gamma \in h_2} \max\{\gamma_1, \gamma_2\}; \\ (iii) \quad h_1 \cap h_2 &= \bigcap_{\gamma \in h_1, \gamma \in h_2} \min\{\gamma_1, \gamma_2\}; \\ (iv) \quad h^\lambda &= \bigcup_{\gamma \in h} \{\gamma^\lambda\}; \\ (v) \quad \lambda h &= \bigcup_{\gamma \in h} \{1 - (1 - \gamma)^\lambda\}; \\ (vi) \quad h_1 \oplus h_2 &= \bigcup_{\gamma \in h_1, \gamma \in h_2} \{\gamma_1 + \gamma_2 - \gamma_1 \gamma_2\}; \\ (vii) \quad h_1 \otimes h_2 &= \bigcup_{\gamma \in h_1, \gamma \in h_2} \{\gamma_1, \gamma_2\}; \end{aligned}$$

Definition 3. For a HFE $h_H(t)$, the score function $S(h_H(t))$ is defined as follows Xu and Xia (2011):

$$S(h_H(t)) = \frac{\sum_{j=1}^{l(h_H(t))} h_H^{\sigma(j)}(t)}{l(h_H(t))}$$

where $S(h_H(t)) \in [0, 1]$.

Based on score function it is possible to make ranking of HFEs according to the following rules:

- $h_1 > h_2$, if $s(h_1) > s(h_2)$;
- $h_1 < h_2$, if $s(h_1) < s(h_2)$;
- $h_1 = h_2$, if $s(h_1) = s(h_2)$.

h_1 and h_2 are two elements in HFEs.

5. The MCDM Methods in Hesitant Environment with Standard Deviation Weights

The proposed method can be seen in Figure 1. It consists of three methods, namely hesitant fuzzy MOORA, hesitant fuzzy TOPSIS and hesitant fuzzy VIKOR with hesitant fuzzy standard deviation, which will be briefed in the following subsections.

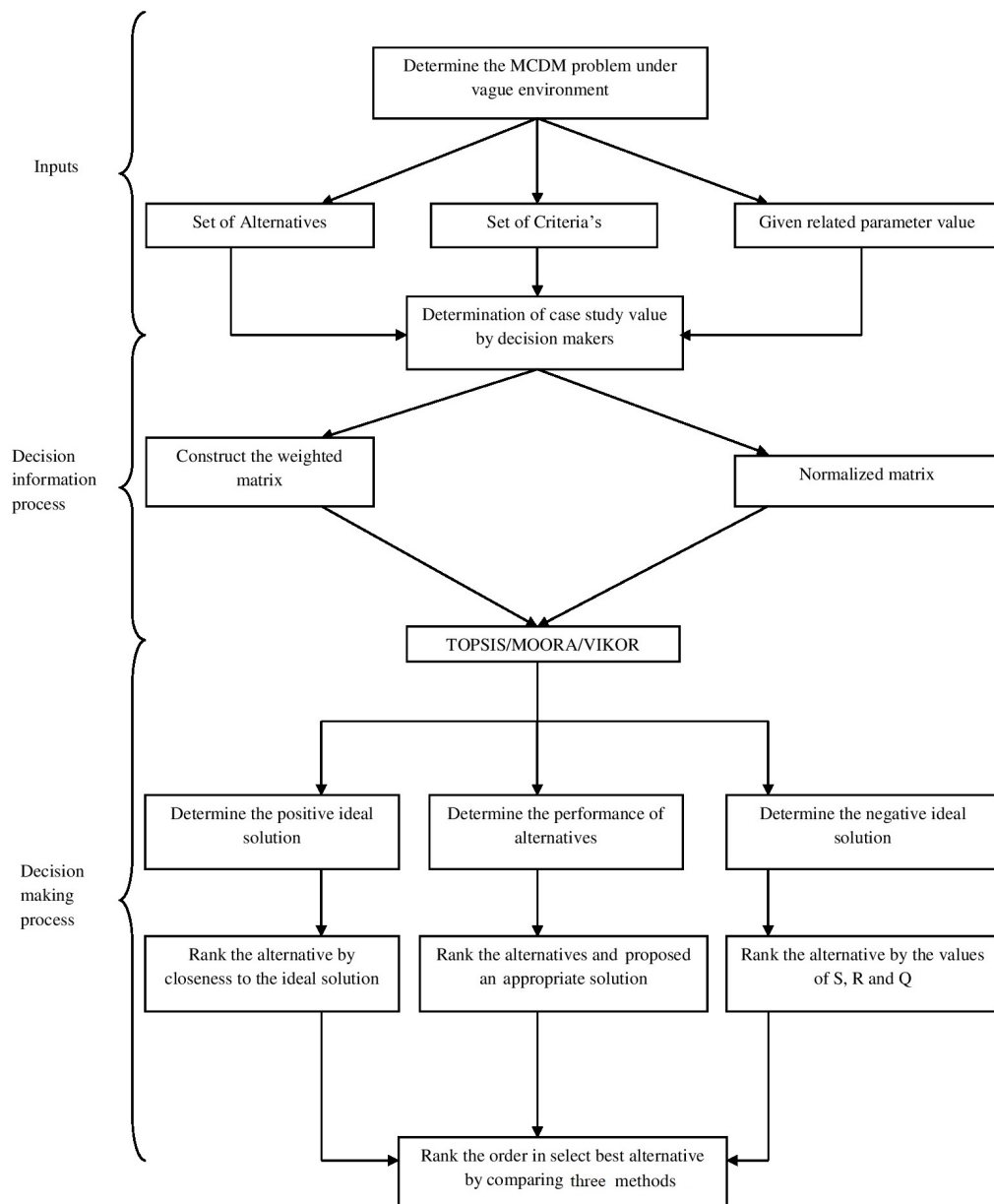


Figure 1. Flowchart of proposed method.

5.1. HFSDV for Calculating Criteria Weights

In MCDM technique, the standard weight finding method measures the weights of each of the selected parameters, based on the standard deviation. In this section, we extend the weight finding method of SDV (Santawy and Ahmed, 2012) [9] in hesitant fuzzy environment. The time of weight comparison, purpose of standard deviation was used to transform different scales and units among various criteria into common measurable units.

The score of HF number reflects the degree of ambiguity, we can calculate the SDV through hesitant fuzzy number score and then obtain the criteria weights. Let $P_i (i = 1, 2, \dots, n)$ be a set of n alternatives, $R_j (j = 1, 2, \dots, m)$ be a set of m criteria. The hesitant fuzzy decision matrix H can be written as:

$$H = \begin{bmatrix} h_{11} & h_{12} & \dots & h_{1n} \\ h_{21} & h_{22} & \dots & h_{2n} \\ \dots & \dots & \ddots & \dots \\ h_{m1} & h_{m2} & \dots & h_{mn} \end{bmatrix}$$

In this research paper, proposed SDV method under HF environment is used to manage hesitation situation of the decision makers in the selection of evaluation criteria. The algorithm of hesitant fuzzy SDV (HFSDV) as follows:

Step 1: In first, construct the score matrix (SM) by using (Definition 3), we get score values of given hesitant fuzzy elements.

$$S(h_H(t)) = \frac{\sum_{j=1}^{l(h_H(t))} h_H^{\sigma(j)}(t)}{l(h_H(t))} \quad (1)$$

Step 2:

$$H(X'_{ij}) = \frac{H(X_{ij}) - \min_{1 \leq j \leq n} H(X_{ij})}{\max_{1 \leq j \leq n} H(X_{ij}) - \min_{1 \leq j \leq n} H(X_{ij})} \quad (2)$$

here, $\max X_{ij}$ denoted as maximum values of the criteria and $\min X_{ij}$ denoted as minimum values of the criteria j . Then, SD of each criteria is calculated by use of the following formula.

$$HSDV_j = \sqrt{\frac{1}{m} \sum_{i=1}^m H(X_{ij} - \bar{X}'_j)^2} \quad (3)$$

Here, \bar{X}'_j is the average values of j^{th} criterion and $j = 1, 2, \dots, n$. The weight of each criteria can be calculated as,

$$W_j = \frac{HSDV_j}{\sum_{j=1}^m HSDV_j} \quad j = 1, 2, \dots, m. \quad (4)$$

where, m denotes the number of indicators; $0 \leq W_j \leq 1$, $\sum_{j=1}^m W_j = 1$.

5.2. HFSDV-MOORA

Consider, water distribution system selection problem. Here, we consider m as alternatives $P_i (i = 1, 2, \dots, n)$ and criteria $R_j (j = 1, 2, \dots, m)$. And proposed SDV-MOORA algorithm is framed with hesitant fuzzy elements (HFE).

Step 1:

Frame the HFDM. Here, the hesitant elements values are in acceding order. It is shown in the following Table 2.

Table 2. Hesitant decision making matrix.

	R_1	R_2	\dots	R_n
P_1	$\langle x(h_{11}) \rangle$	$\langle x(h_{12}) \rangle$	\dots	$\langle x(h_{1n}) \rangle$
P_2	$\langle x(h_{21}) \rangle$	$\langle x(h_{22}) \rangle$	\dots	$\langle x(h_{2n}) \rangle$
\vdots	\vdots	\vdots	\vdots	\vdots
P_m	$\langle x(h_{m1}) \rangle$	$\langle x(h_{m2}) \rangle$	\dots	$\langle x(h_{mn}) \rangle$

Step 2:

Determine score matrix of Hesitant fuzzy matrix, the score value is determined by use of Equation (1).

Step 3:

The process of normalization is framed in to a range of [0, 1] by use of the following equation,

$$h(X_{ij})^* = \frac{h(X_{ij})}{\sqrt{\sum_{i=1}^n x_{ij}^2}} \quad (5)$$

here, $h(X_{ij})^*$ denotes the hesitant fuzzy normalization matrix (HFNM) for i^{th} alternative and j^{th} criteria, here $i = 1, 2, \dots, n$ and $j = 1, 2, \dots, m$.

Step 4:

Next we calculate the criteria weights by use of hesitant fuzzy preference value $w_i = (w_{i1}, w_{i2}, w_{i3}, w_{i4}, w_{i5})$, ($i = 1, 2, \dots, n$) and ($j = 1, 2, \dots, m$) respectively. Then, the score hesitant fuzzy (HF) preference \tilde{X}_{ij} of each alternatives with respect to each criteria can be calculated by use of Equations (3) and (4). Therefore, the numerical water distribution problem can be clearly formatted in the following resultant matrix format

$$\tilde{D} = \begin{bmatrix} \tilde{R}_{11} & \tilde{R}_{12} & \dots & \tilde{R}_{1n} \\ \tilde{R}_{21} & \tilde{R}_{22} & \dots & \tilde{R}_{2n} \\ \dots & \dots & \ddots & \dots \\ \tilde{R}_{m1} & \tilde{R}_{m2} & \dots & \tilde{R}_{mn} \end{bmatrix}$$

$$\tilde{w} = (\tilde{w}_1, \tilde{w}_2, \dots, \tilde{w}_m)$$

Step 5:

The HF beneficial and non-beneficial criteria are denoted as T_j^* and T_j respectively. Finally, the performance score (PS) of all alternatives is founded as y_i . It is calculated by use of hesitant weighted sum of beneficial and non-beneficial criteria.

$$H(y_i) = \sum_{j=1}^m H(w_j)H(X_{ij}^*) - \sum_{j=m+1}^n H(w_j)H(X_{ij}) \quad (6)$$

where, m refer the number of criteria and it is to be maximized; n-m denotes the number of minimized criteria; weights of criteria is w_j .

In some situation, if we assume equal importance to all criteria than the proposed HFSDV-MOORA optimization algorithm becomes,

$$H(y_i) = \sum_{j=1}^m H(T_j^*) - \sum_{j=m+1}^n T_j(X_{ij}) \quad (7)$$

$$H(T_{ij}^*) = \sum_{j=1}^m H(w_j)H(X_{ij}^*); H(T_{ij}) = \sum_{j=m+1}^n H(w_j)H(X_{ij})$$

Step 6:

Based on performance score, we rank the each alternative $H(y_i)$. The best is select by the highest value of performance score value.

5.3. HFSDV-TOPSIS

Technique for Order Preference by Similarity to an Ideal Solution (TOPSIS) can be used to rank the different alternatives to the selected criteria. In the TOPSIS approach, an alternative that is nearest to the FPIS and farthest from the FNIS is chosen as optimal Beg and Rashid, (2013) [12]. TOPSIS method is used for MCDM where the decision makers opinions about the criteria values are expressed in hesitant fuzzy values.

Step 1: Construct normalized score matrix (\tilde{C}_{ij}) .

$$\tilde{X}_{ij} = \frac{X_{ij}}{\sum_{i=1}^n X_{ij}} \quad (8)$$

where $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$.

Step 2: Construct the weighted normalized matrix by multiplying each column of the normalized matrix by its associated weight. The weight values are find by using (4). Hence we get a new matrix

$$T_{ij} = W_j * \tilde{X}_{ij} \quad (9)$$

where $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$.

Step 3: Find the positive ideal solution (PIS)

$$\begin{aligned} A^+ &= \{T_1^+, T_2^+, \dots, T_n^+\} \text{ maximum values} \\ A^+ &= (\{Max(T_{ij})\}) \quad \text{if } j \in J \end{aligned} \quad (10)$$

Step 4: Find the negative ideal solution (NIS)

$$\begin{aligned} A^- &= \{T_1^-, T_2^-, \dots, T_n^-\} \text{ minimum values} \\ A^- &= (\{Min(T_{ij})\}) \quad \text{if } j \in J \end{aligned} \quad (11)$$

Step 5: Determine the separation from positive ideal solution S_i^+

$$S_i^+ = \left[\sum \left\{ (T_j^+ - T_{ij})^2 \right\} \right]^{\frac{1}{2}} \quad (12)$$

where $i = 1, 2, 3, \dots, n$.

Step 6: Determine the separation from negative ideal solution S_i^-

$$S_i^- = \left[\sum \left\{ (T_j^- - T_{ij})^2 \right\} \right]^{\frac{1}{2}} \quad (13)$$

where $i = 1, 2, 3, \dots, n$.

Step 7: Calculate the relative closeness to the ideal solution

$$X_i^+ = \frac{S_i^-}{(S_i^+ + S_i^-)} \quad (14)$$

where $0 \leq X_i \leq 1$ and $i = 1, 2, \dots, n$.

Step 8: Based on the decrease order values of closeness to the ideal solution, alternatives are ranked from higher value to lower value.

5.4. HFSDV-VIKOR

The Visekriterijumsko KOMPromisno Rangiranje (VIKOR) method was developed as an Multi-Criteria Decision Making (MCDM) method to solve a discrete problem with conflicting criteria. The purpose of this method is to determine a feasible solution for ranking criteria. This VIKOR method is used for MCDM where the decision makers opinions about the criteria values are expressed in hesitant fuzzy values Opricovic et al. (2004) [44].

Step 1: Construct normalized score matrix (\tilde{X}_{ij}).

$$\tilde{X}_{ij} = \frac{X_{ij}}{\sum_{i=1}^n X_{ij}} \quad (15)$$

where $i = 1, 2, \dots, n$ and $j = 1, 2, \dots, m$.

Step 2: Calculate the positive ideal solution (PIS)

$$f_i^* = [(Max_j f_{ij} | i \in I_1), (Min_j f_{ij} | i \in I)] \quad (16)$$

Step 3: Calculate the negative ideal solution (NIS)

$$f_i^- = [(Min_j f_{ij} | i \in I_2), (Max_j f_{ij} | i \in I)] \quad (17)$$

Step 4: Calculate utility measures(S_{ij}). The weight values are find by using (4).

$$S_{ij} = W_j \frac{(f_i^* - f_{ij})}{(f_j^* - f_j^-)} \quad (18)$$

Step 5: Calculate utility measure(S) and regard measure(G) for each alternative

$$G_j = Max(S_{ij}) \quad (19)$$

$$S_i = \sum_{j=1}^n W_j \left(\frac{f_i^* - f_{ij}}{f_j^* - f_j^-} \right) \quad (20)$$

Step 6: Computation the index value (Q_i) ith alternative as follows,

$$Q_i = \frac{v(S_i - S^*)}{(S^- - S^*)} + \frac{(1 - v)(G_j - G^*)}{(G^- - G^*)} \quad (21)$$

where $S^* = \min S_j$, $S^- = \max S_j$, $G^* = \min G_j$, $G^- = \max G_j$

Step 7: These values are sorted by rank, it is based on S G and Q in decreasing order.

Step 8: Consider the alternative i , corresponding to $Q_{[i]}$ as a trade-off solution if the two conditions given below are satisfied:

C₁. Acceptable advantage:

$Q(A_2) - Q(A_1) \geq (\frac{1}{n-1})$, with n alternatives.

C₂. Acceptable stability:

If condition C₁ is not met with, then a set of alternate solutions are proposed that consist of:

- (i). Alternatives i_1, i_2 where $i_1 = i_2$ only if the condition is C₂ is not satisfied.
- (ii). Alternatives i_1, i_2, \dots, i_k if the condition C₁ is not fulfilled; i_k is determined by the relation $Q_K - Q_1 < \frac{1}{(n-1)}$ for max k , where $Q_{jk} = Q_k$.

In ranking alternatives, the most preferred alternative should be select to the positive ideal solution for maximization criteria and will negative ideal solution for the minimization criteria.

6. Study Area

We consider drought prone areas of surrounding three districts viz. Erode, Coimbatore and in Tirupur, Tamil Nadu, India. Precipitation is just source of ground water and is separated as open well and bore well for irrigation.

People living in these regions are with agriculture as the fundamental occupation. Because of borewell burrowing and whimsical pumping of water by the populace, water table in the local has gone down enormously from 600 to 1500 feet underneath the ground level. Utilizing engine pumps brought about substantial utilization of electricity. Since water is exhausting at a quick pace, these disturbing circumstance soon will cause a man made calamity. It also will be an irreversible harm to environment and living condition. For rectifying this issue, we acknowledge the course of action from surplus water of Bhavani river.

Athikadavu-Avinashi drinking water supply-cum-irrigation scheme, which aims at managing the water resources in the Kongu region of Tamil Nadu. Athikadavu-Avinashi drinking water supply-cum-irrigation scheme focus to reroute the surplus water of Bhavani river to the lakes, ponds and other water bodies in the drought-prone zones of Coimbatore, Tiruppur and Erode districts as a lift to their ground water level. Around 1.5 TMC(Thousand Million Cubic) of surplus water will be pumped from the Kalingarayan anicut and used to fill the 32 tanks of the Public Works Department, 42 tanks of the Panchayat unions and 971 ponds in the three districts.

River Bhavani, a tributary of Cauvery, is a perennial river that rises in the Nilgiris. The river reaches Pillur dam before joining Cauvery in Erode district in Tamil Nadu and continues its journey to drain into the Bay of Bengal. Bhavani river majorly flows through Coimbatore and Erode districts and is a major resource for irrigation in West Tamil Nadu. The Bhavani river is fed by the rains received mainly during the southwest monsoon, which causes the river to swell during June-September season and flood its banks. Though Pillur dam and Bhavanisagar dam help in avoiding high water flow in the river during monsoon, more often than not, the dams also reach full capacity, leading to flooding downstream due to the shutters being opened. For conducive to overflow and wastage of water due to dams reaching full capacity. To avoid this, the Athikadavu—Avinashi project was proposed [53–60].

In this paper, we selecting the issue of ground water irrigation as a mathematical problem. Here we make a decision about this problem. For that we select decision making process, because decision making provides excellent way to prefer the best alternative from selected alternatives. Here four alternatives are selected depending on various criteria like social, economics, environmental, effectiveness and technical operations. Towards this, we use MCDM method.

In this case, decision making procedure could be unfairly impacted by the information both while assessing alternatives and thinking about gaining from experts. Thus, in an uncertainty decision condition of MCDM process, the experts advice to use the vague set theory. In an accurate way to reflect upon the information given by the different experts, the information is displayed in the type of hesitant fuzzy elements. The root map of water distribution system of selected drought prone areas are shown in Figure 2.

Hence, a HFMCDM approach has been used based on our proposed methods.



Figure 2. Water distribution system of selected drought prone areas.

Description of Problem

The study area consists of four water distribution methods for drought prone areas. Different water resource methods were used in various studies in setting the criteria on decision model. Since the aim of the mathematical problem is to find one of the best water distribution method for water irrigation system, we identify four alternatives and five criteria.

The criteria are separated in two categories viz., the qualitative criteria and the quantitative criteria. The information of criteria are collected through field surveys, questionnaires from the public and interviews of the representatives of the affected region.

The Alternatives are:

(i). Closed concrete pipes laid along agricultural land (P_1)

Comparison when the water transported through this closed concrete pipe line system, the speed of water is uniform, making it an ideal way to distribute feeder water to the pools and ponds located in the panoramic union of each creek. It is also a closed system of transporting water. So water theft and water pollution are unlikely. In addition, the panels that are used in this project can be combined into a series of maps and pipeline through the agricultural lands. So that every village can be used in conjunction with this scheme. It is also considered a cost effective pipeline system to reduce the water transport distance. In this system, the conversation is considered the best alternative system. Since pits are placed beneath the agricultural land and buried in the closed concrete pipeline and carried to the water, thereby causing no harm to agricultural land and agriculture.

(ii). Closed concrete pipes laid along roadsides of national highway (P_2)

In this alternative, closed concrete pipes are consider and laid along the road sides of national highway. The national highway to suit drinking water supplies. Moreover the national highway is the best choice when it comes to drinking and sewerage purposes. The water distribution system along the national highway has to be explored on a number of issues as we examine the criteria taken in this research article. Because of the bridge and road safety systems are found on the national highway, this is considered to be risk of the water distribution system. And it has to be up while receiving permits from the officials of the national highway. Further, the decision to construct this water supply system is important considering on environment safety, costs and the maintenance.

(iii). Concrete canal through agricultural land (P_3)

In our research work, the open canal system with concrete will be considering one of the alternatives. The water supply system with concrete structure on either side of the open canals is consider to be very effective for evaporation loss, seepage, theft of water by illegal pumping. Although this type of water distribution can be seen to be better for agriculture and groundwater recharge. This alternative need to be explored in terms of social acceptance, costs, environment and maintenance.

(iv). Earthen canal through agricultural land (P_4)

The first consideration is when water is transported through an open canal system, which may be used for certain factories rather than for cultivation of agree land and groundwater recharge. This distribution structure is only considered suitable for seasons where the water source is high. When viewed positively, there is a possibility that groundwater will rise in the area on both sides of the canal along the way through agricultural land. When viewed negatively the cost of safety and maintenance of this water distribution system will out weight the other alternative. Also open canal system through agricultural land can result more land actuation.

The description of the alternatives are shown in Figure 3. For evaluating these four alternatives we select five criteria on the basis of expected impacts of canal improvement, and from opinions of public sectors include farmers' benefits of the concerned districts. The hierarchical structure of selected criteria are shown in Figure 4.

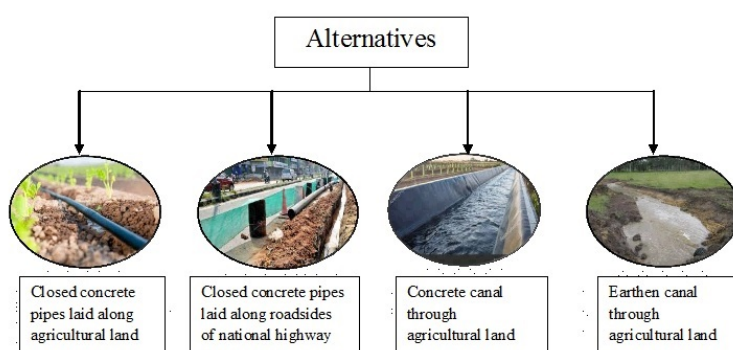


Figure 3. Description of alternatives.

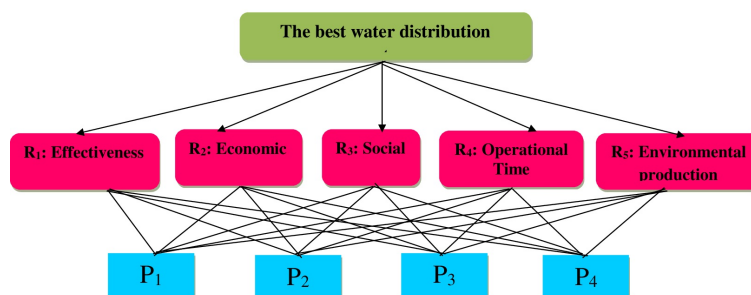


Figure 4. The hierarchical structure of selected criteria.

The criteria are:

(i) Effectiveness (R_1)

This criterion solves the amount of minimizing risk due to the application of the scheme. The effectiveness criterion is a quantitative criterion. The critical assessment model is to improve the living condition of the people in the drought prone area to improve the current situation of the selected location.

(ii) Economic (R_2)

In the economic criteria, we consider rural income and increasing employment opportunity. It is also concentrated among poor agricultural production, poor income, poor standard of living and poverty.

The improvement of dynamic crop production, the increase in livestock capital and build up of processing industries are encouraged also from the improvement of infrastructures.

(iii) Social (R_3)

Social criterion helps assess the level of attitudes of people in the selected area and the rate of the urbanization in the region. This qualitative criteria is based on sale of land, entitlement and migration.

(iv) Operational time (R_4)

This quantitative criterion is one of most difficult criteria to compare the others. This technical operation time considers the risk of selected water distribution method. The possibilities for using renewable energy sources to be taken from the river to project need to be examined.

(v) Environment production (R_5)

Environmental production should be constructed in a way that does not affect people in the environment and to promote the environment. This production among ground water should be unpolluted at the same time it should increase the groundwater level.

7. Numerical Evaluation

In this section, we make a numerical evaluation from the study area. In the drought prone areas of three district, they have planned four types of water distribution system. Here, we consider these four types of water distribution system as the alternatives. The four alternatives are P_1 (Closed concrete pipes laid along agricultural land), P_2 (Closed concrete pipes laid along roadsides of national highway), P_3 (Concrete canal through agricultural land) and P_4 (Earthen canal through agricultural land). These four alternatives choose based among the following five criteria. R_1 (Effectiveness), R_2 (Economic), R_3 (Social), R_4 (Operational time) and R_5 (Environmental production).

The $A_{(ij)}$ is the matrix whose elements reveal the value of different alternatives for the respective criteria. The index i denoted for different alternatives, while the index j denoted for the different criteria. Table 3 shows the numerical calibration for the study area.

Table 3. Numerical Example.

	R_1	R_2	R_3	R_4	R_5
P_1	{0.6, 0.7, 0.8, 0.9}	{0.5, 0.6, 0.7}	{0.9, 0.6, 0.8}	{0.7, 0.8, 0.2, 0.9}	{0.7, 0.9, 0.8}
P_2	{0.5, 0.7, 0.6}	{0.3, 0.4, 0.6, 0.5, }	{0.4, 0.5, 0.3, 0.7}	{0.6, 0.7, 0.4}	{0.7, 0.5, 0.8}
P_3	{0.3, 0.5, 0.6, 0.7}	{0.4, 0.7, 0.5, 0.6, }	{0.2, 0.3, 0.4, 0.6}	{0.3, 0.4, 0.5}	{0.2, 0.4, 0.5, 0.6}
P_4	{0.4, 0.6, 0.7, 0.8}	{0.8, 0.5, 0.7, 0.9}	{0.5, 0.7, 0.9, 0.6}	{0.2, 0.3, 0.1, 0.4}	{0.1, 0.2, 0.3}

The hesitant fuzzy decision matrix (HFDM):

$$H = \begin{bmatrix} \{0.6, 0.7, 0.8, 0.9\} & \{0.5, 0.6, 0.7\} & \{0.6, 0.8, 0.9\} & \{0.2, 0.7, 0.8, 0.9\} & \{0.7, 0.8, 0.9\} \\ \{0.5, 0.6, 0.7\} & \{0.3, 0.4, 0.5, 0.6\} & \{0.3, 0.4, 0.5, 0.7\} & \{0.4, 0.6, 0.7\} & \{0.5, 0.7, 0.8\} \\ \{0.3, 0.5, 0.6, 0.7\} & \{0.4, 0.5, 0.6, 0.7\} & \{0.2, 0.3, 0.4, 0.6\} & \{0.3, 0.4, 0.5\} & \{0.2, 0.4, 0.5, 0.6\} \\ \{0.4, 0.6, 0.7, 0.8\} & \{0.5, 0.7, 0.8, 0.9\} & \{0.5, 0.6, 0.7, 0.9\} & \{0.1, 0.2, 0.3, 0.4\} & \{0.1, 0.2, 0.3\} \end{bmatrix}$$

Calculation of HFSDV and HFSDV-MOORA:

Step 1:

We determine the hesitant score matrix (HSM) C_{ij} based on Equation (1).

Score matrix:

$$C_{ij} = \begin{bmatrix} 0.75 & 0.6 & 0.7667 & 0.65 & 0.8 \\ 0.6 & 0.45 & 0.475 & 0.5667 & 0.6667 \\ 0.525 & 0.55 & 0.375 & 0.4 & 0.425 \\ 0.625 & 0.725 & 0.675 & 0.25 & 0.2 \end{bmatrix}$$

Step 2:

We construct the normalized matrix by using (5).

Normalized score matrix :

$$\tilde{C}_{ij} = \begin{bmatrix} 1 & 0.545 & 1 & 1 & 1 \\ 0.333 & 0 & 0.255 & 0.793 & 0.778 \\ 0 & 0.364 & 0 & 0.375 & 0.375 \\ 0.444 & 1 & 0.765 & 0 & 0 \end{bmatrix}$$

Step 3:

The hesitant SDV for all criteria are calculated by using (3).

$$HFSDV_1 = 0.3219, HFSDV_2 = 0.322, HFSDV_3 = 0.355, HFSDV_4 = 0.3448, HFSDV_5 = 0.3426. \quad (22)$$

Step 4:

By using Equation (4), we calculate each criteria weights as follows and the resultant values of weighted criteria are shown in the following Figure 5

$$w_1 = 0.191, w_2 = 0.191, w_3 = 0.210, w_4 = 0.203, w_5 = 0.205. \quad (23)$$

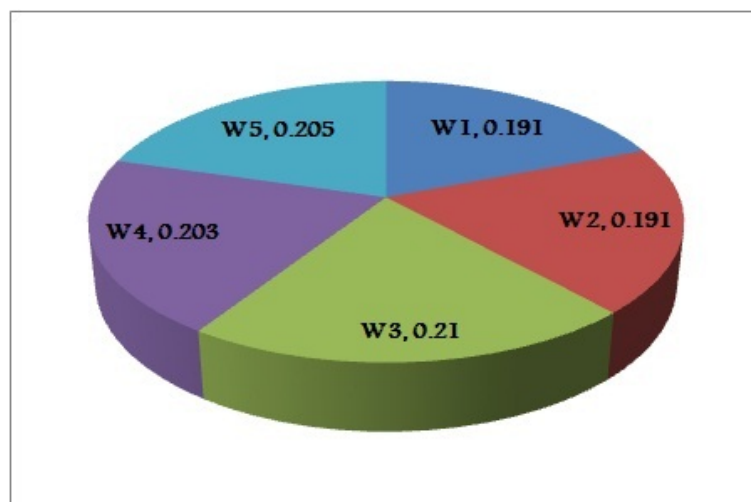


Figure 5. Graphical representation of SDV weights of criteria.

Step 5: Here, the HF- beneficial and HF-non beneficial criteria value are calculated by using (3) & (4). Here, Effectiveness (R_1), Social (R_4) and Environmental production (R_5) are the beneficial criteria; Economic (R_2) and Operational time (R_3) are the non-beneficial criteria values are give as follows,

$$\tilde{T}_{i1} = (0.191, 0.064, 0, 0.085)^T$$

$$\tilde{T}_{i2}^* = (0.104, 0, 0.0695, 0.191)^T$$

$$\tilde{T}_{i3} = (0.210, 0.054, 0, 0.161)^T$$

$$\tilde{T}_{i4}^* = (0.203, 0.161, 0.076, 0)^T$$

$$\tilde{T}_{i5} = (0.205, 0.1595, 0.0769, 0)^T$$

The weighted normalization matrix is calculated as follows,

$$w * h(C_{ij}^*) \quad (24)$$

Then, determine The performance of alternative values are determined with beneficial and non-beneficial criteria by using (6). The resultant values are presented as follows,

$$H(y_1) = 0.299, H(y_2) = 0.117, H(y_3) = -0.069, H(y_4) = 0.055.$$

Step 6: By use of performance of alternatives, the selected alternatives are ranked in decreasing order. The resultant alternative values are shown in Table 4. Here, P_1 is considered as one of the best water distribution system. Also, the result is shown in the Figure 6. Thus, we conclude that one of the most suitable water distribution system is Closed concrete pipes laid along agricultural land.

Table 4. Ranking the alternatives in decreasing order.

Alternatives	Performance of Alternative Values	Rank
P_1	0.299	1
P_2	0.117	2
P_3	-0.069	4
P_4	0.055	3

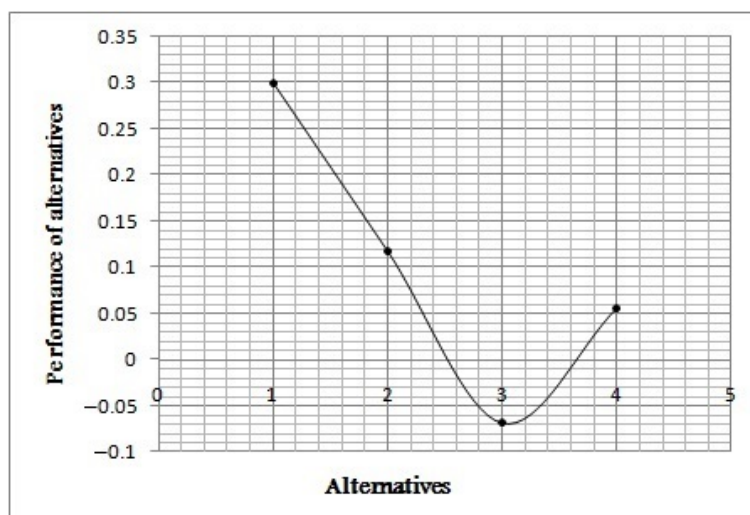


Figure 6. The result of HFSDV-MOORA method.

Calculation of HFSDV-TOPSIS:

Step 1:

In first, the hesitant score matrix (HSM) is framed by use of selected hesitant fuzzy values C_{ij} based on (1).

Step 2:

Next, the normalize score matrix (NHSM) \tilde{C}_{ij} is calculated by use of (8).

Normalized score matrix:

$$\tilde{C}_{ij} = \begin{bmatrix} 0.3 & 0.258 & 0.335 & 0.348 & 0.382 \\ 0.24 & 0.194 & 0.207 & 0.304 & 0.319 \\ 0.21 & 0.237 & 0.164 & 0.214 & 0.203 \\ 0.25 & 0.312 & 0.295 & 0.134 & 0.096 \end{bmatrix}$$

Step 3:

Weights of selected criteria are calculated by using (4) the hesitant fuzzy standard deviation weights are mentioned above in (23).

Step 4:

Then the hesitant normalized score matrix is multiply by standard deviation using(9).

$$T_{ij} = \begin{bmatrix} 0.0573 & 0.0493 & 0.0704 & 0.0706 & 0.0783 \\ 0.0458 & 0.0371 & 0.0435 & 0.0617 & 0.0654 \\ 0.0401 & 0.0453 & 0.0344 & 0.0434 & 0.0416 \\ 0.0478 & 0.0596 & 0.0620 & 0.0272 & 0.0197 \end{bmatrix}$$

Step 5:

Utilize Equations (10) and (11) to calculate the hesitant fuzzy positive ideal solution D^+ and negative ideal solution D^- . The result of positive and negative ideal solution are shown in Figure 6.

$$D^+ = \{0.0573, 0.0596, 0.0704, 0.0706, 0.0783\} \quad (25)$$

$$D^- = \{0.0401, 0.0371, 0.0344, 0.0272, 0.0197\} \quad (26)$$

Step 6:

Utilize Equations (12) and (13) to calculate the separation measures alternative from the hesitant fuzzy positive ideal solution A_i^+ and negative ideal solution A_i^- . The separation measures of positive and negative ideal solution are given as follows,

$$S_i^+ = \{0.01, 0.04, 0.062, 0.0742\} \quad (27)$$

$$S_i^- = \{0.0837, 0.0034, 0.03, 0.0361\} \quad (28)$$

Step 7:

Utilize equation(17) to calculate the relative closeness coefficient of an alternative P_i with respect to the hesitant fuzzy positive ideal solution C_i^+

$$C_i^+ = \{0.893, 0.0783, 0.033, 0.037\} \quad (29)$$

Rank the alternatives (P_i) according to the closeness coefficient:

$$P_1 > P_2 > P_4 > P_3$$

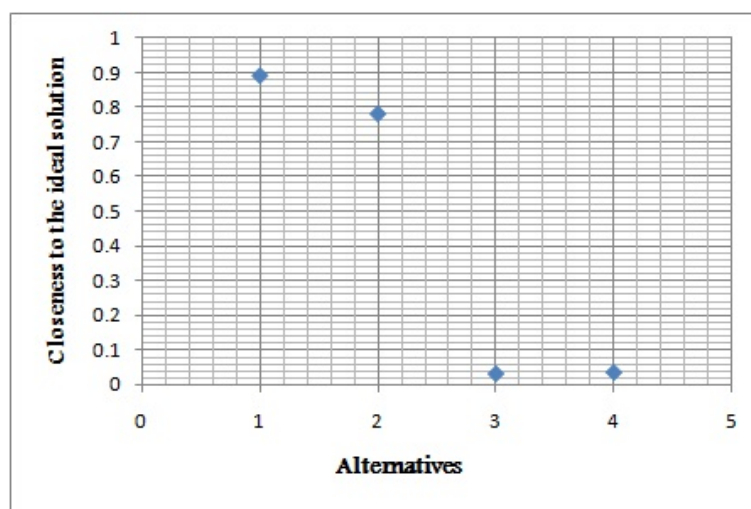


Figure 7. Graphical representation of TOPSIS values.

Thus the highly preferable alternative is P_1 so, closed concrete pipes laid along agricultural land is best for the selected problem. The result of TOPSIS values are shown in Figure 7.

Calculation of HFSDV-VIKOR:

Step 1:

By using normalized score matrix with hesitant fuzzy elements, we calculate each positive and negative ideal solution of (16) and (17).

$$F^* = \{0.3, 0.312, 0.335, 0.348, 0.382\} \quad (30)$$

$$F^- = \{0.21, 0.194, 0.164, 0.134, 0.096\} \quad (31)$$

Step 2:

The utility measures for the founded score matrix S_{ij} is determined by using (21) given in the following Table 5.

Table 5. The utility measures matrix S_{ij} .

	R_1	R_2	R_3	R_4	R_5
P_1	0	0.087	0	0	0
P_2	0.127	0.191	0.157	0.042	0.045
P_3	0.191	0.121	0.21	0.127	0.128
P_4	0.106	0	0.049	0.203	0.205

Step 3:

We determined the regret measures (G) by using (22).

$$G_1 = 0.087, \quad G_2 = 0.191, \quad G_3 = 0.191, \quad G_4 = 0.205. \quad (32)$$

Step 4:

Here, we determined utility measures (S) by using (23).

$$S_1 = 0.0807, \quad S_2 = 0.562, \quad S_3 = 0.777, \quad S_4 = 0.563. \quad (33)$$

Step 5:

And, we determined the value of index (Q_i) for each selected alternative by using (24).

$$Q_1 = 0, \quad Q_2 = 0.785, \quad Q_3 = 0.941, \quad Q_4 = 0.845. \quad (34)$$

Step 6:

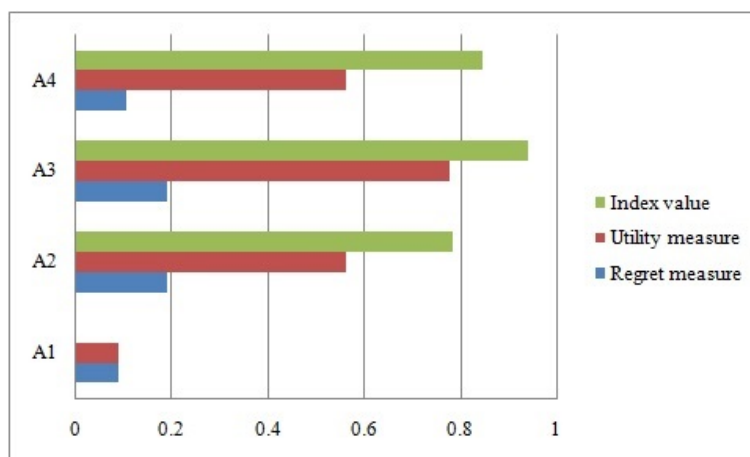
Finally, rank the alternatives by use of Q value in decreasing order, it gives

$$P_1 > P_2 > P_3 > P_4$$

Here, P_1 is best alternative for selected water distribution system. The final results are given in Table 6. The result of regret measure(G_i), utility measure(S_i) and index value(Q_i) as shown in Figure 8.

Table 6. The evaluation value of each alternative.

	G_i	S_i	Q_i
P_1	0.087	0.087	0
P_2	0.191	0.562	0.785
P_3	0.191	0.777	0.941
P_4	0.106	0.563	0.845

**Figure 8.** Graphical representation of HFSDV-VIKOR result.**Step 7:**

According to the condition (C_1) in Section 5.4, We can get,

$$\begin{aligned}
 Q(P_2) - Q(P_1) &= 0.6251 - 0 \\
 &= 0.6251 \geq \frac{1}{4-1} \quad (\text{here } n = 4) \\
 &= 0.6251 > \frac{1}{3}.
 \end{aligned}
 \tag{35}$$

According to the condition (C_2) in Section 5.4, rankings of S and R coincide with the classification of Q and thus satisfies the condition (C_2).

8. Comparison Analysis

In this comparison part, we give some detailed comparison analysis between our proposed HFSDV-MCDM methods with the existing HFWEM-MCDM methods.

In weight detection techniques, the SDV method is considered the best method compared to the WEM method because the SDV weighing system that distributes the weight between the selected criteria. The standard deviation weight is appropriate to provide the most stable weights for the basic criteria of effectively selecting the options available at the end of the problem solving process. Therefore, in MCDM techniques, the SDV weighing method is better than the WEM method in the case of the efficient approach.

we compare the ranking results of proposed HFSDV-MOORA with existing HFWEM-MOORA, HFSDV-TOPSIS with HFWEM-TOPSIS and HFSDV-VIKOR with HFWEM-VIKOR, the results give the same ranking order. The comparison ranking results are given in the Tables 7 and 8, from these comparison table, the four alternative ranking of proposed hesitant fuzzy MCDM methods with standard deviation weight finding technique is perfectly parallel to the sequence of existing hesitant fuzzy MCDM with weighted entropy measure weighting technique. Also the final results are shown in the following Figures 9–11. And also it exposed the reliability and authenticity of our proposed method. From the comparison Table 7, the ranking places of proposed HFSDV-MCDM method is as follows,

- In HFSDV-MOORA $P_1 > P_2 > P_4 > P_3$
- In HFSDV-TOPSIS $P_1 > P_2 > P_4 > P_3$
- In HFSDV-VIKOR $P_1 > P_2 > P_4 > P_3$

Moreover, in HFWEM-MCDM methods, the ranking results having minor differences when comparing MOORA with another two methods TOPSIS and VIKOR. In VIKOR method (P_3 is prefer than P_4). In HFSDV-MOORA, P_4 is prefer than P_3 . The reason for the smallest difference is that, in MCDM problems, criteria weights are make an important role. Here, the weighting of criteria reflects their relative importance in the DM process. While the calculation of HFWEM-MOORA method, the ratio analysis gives importance of both beneficial criteria and non-beneficial criteria. In HFWEM-TOPSIS and HFWEM-VIKOR the ranking results are depends on the closeness to the “ideal” solution. Based on the comparison Table 7, the ranking order of existing HFWEM-MCDM is as follows,

- In HFWEM-MOORA $P_1 > P_2 > P_4 > P_3$
- In HFWEM-TOPSIS $P_1 > P_2 > P_3 > P_4$
- In HFWEM-VIKOR $P_1 > P_2 > P_3 > P_4$

Table 7. Comparison of ranking result for each alternative by using weighted entropy.

Alternatives	HFWEM-TOPSIS	Rank	HFWEM-VIKOR	Rank	HFWEM-MOORA	Rank
P_1	0.901	1	0	1	0.125	1
P_2	0.177	2	0.6251	2	0.058	2
P_3	0.193	3	0.8414	3	0.0101	4
P_4	0.245	4	0.8763	4	0.0324	3

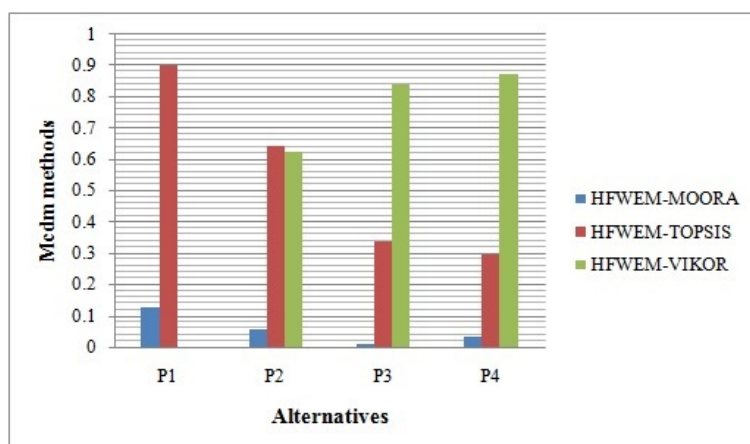


Figure 9. The ranking result of each alternative by using weighted entropy

The ranking result for each alternative is shown in Table 8.

Table 8. The ranking result for each alternative by using SDV weight.

Alternatives	HFSDV-TOPSIS	Rank	HFSDV-VIKOR	Rank	HFSDV-MOORA	Rank
P_1	0.893	1	0	1	0.299	1
P_2	0.0783	2	0.785	2	0.117	2
P_3	0.033	4	0.941	4	-0.069	4
P_4	0.037	3	0.845	3	0.055	3

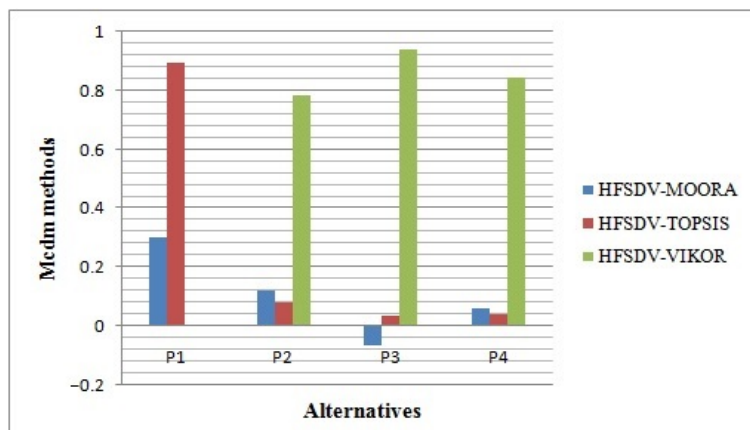


Figure 10. Comparison ranking result of each alternative by using standard deviation weight.

In this analysis part, Spearman's rank correlation method Myers, (2006) was utilized to calculate the correlation coefficients between proposed MCDM methods with SDV weights. The resultant values are shown in the following Table 9.

Table 9. Spearman's rank correlation coefficients between proposed three methods with SDV weights.

Proposed Methods	HFSDV-MOORA	HFSDV-TOPSIS	HFSDV-VIKOR
HFSDV-MOORA	-	1	1
HFSDV-TOPSIS		-	1
HFSDV-VIKOR			-

It's shows that in this part, the value of correlation coefficient between all proposed methods gets the value 1, this is the value of high score of correlation coefficients. It shows that the proposed methods get perfect similar ranking results. The Spearman's rank correlation method was utilized to find the correlation coefficients between comparison methods with WEM weights. The results are shown in the following Table 10.

Table 10. Spearman's rank correlation coefficients between proposed three methods with WEM weights.

Methods	HFWEM-MOORA	HFWEM-TOPSIS	HFWEM-VIKOR
HFWEM-MOORA	-	0.8	0.8
HFWEM-TOPSIS		-	0.8
HFWEM-VIKOR			-

It's shows that in this part, the value of correlation coefficient between all comparison methods gets the value 0.8, this is the acceptable score of correlation coefficients. It shows that the proposed methods get slightly similar ranking results. This test demonstrates the validity and reliability of our proposed methods.

Li, (2014) has already proposed the HF-MOORA for MCGDM, here we compare our HF-MCDM ranking results with proposed SDV weights with Li, (2014), the comparison results have been showing in the following Table 11.

Table 11. The Comparison results with Li, (2014).

Alternatives	Li, (2014) Weights	Rank of Alternatives	Proposed SDV Weights	Rank of Alternatives
A_1	$w_1 = 0.3$	2	$w_1 = 0.2470$	2
A_2	$w_2 = 0.25$	1	$w_2 = 0.2430$	4
A_3	$w_3 = 0.25$	3	$w_3 = 0.2429$	3
A_4	$w_4 = 0.2$	4	$w_4 = 0.2698$	1

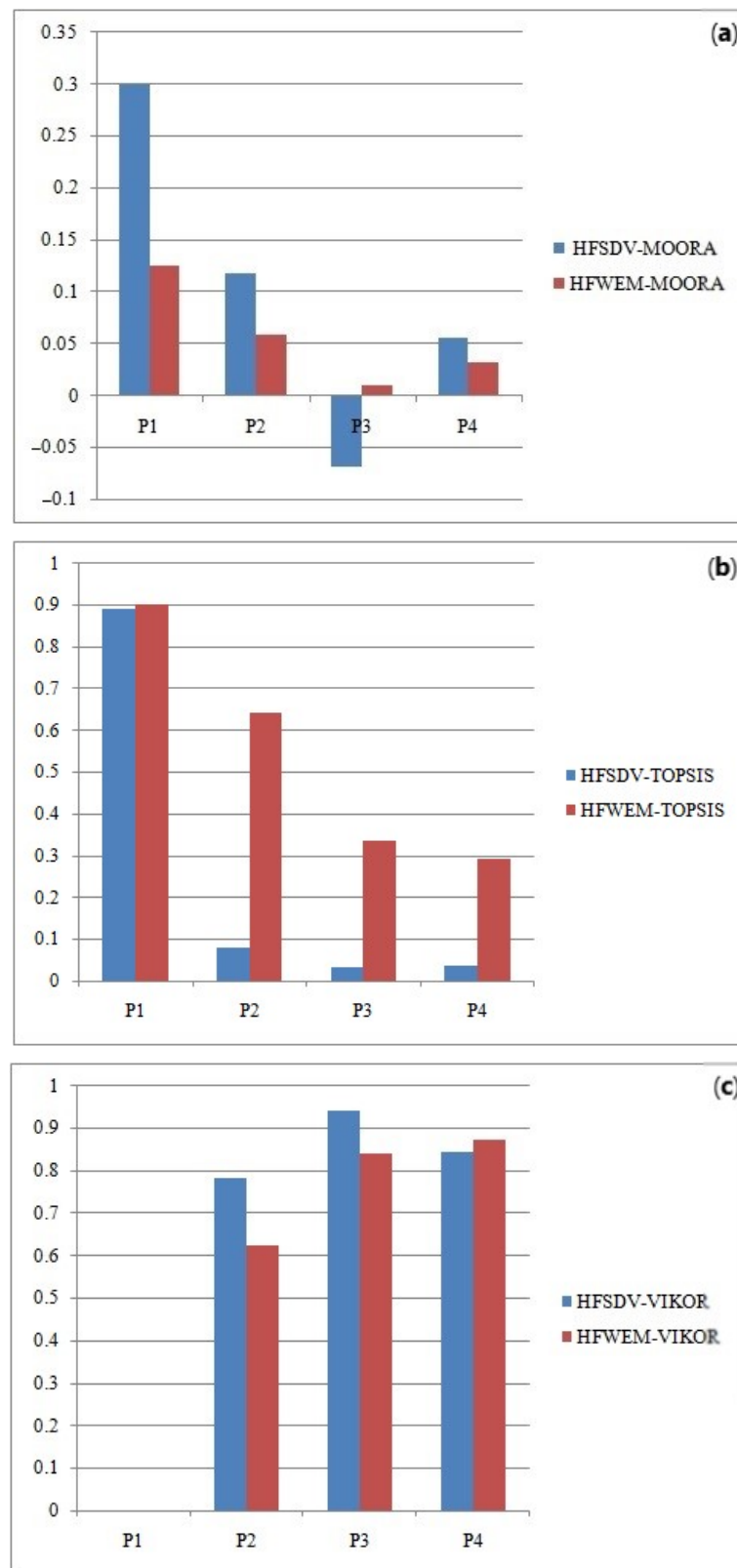


Figure 11. Comparison results: (a)—Comparison results between HFSDV-MOORA and HFWEM-MOORA, (b)—Comparison results between HFSDV-TOPSIS and HFWEM-TOPSIS, (c)—Comparison results between HFSDV-VIKOR and HFWEM-VIKOR

From the comparison table, we can see the SDV weights altered the ranking results of Li, (2014). And also he propose hesitant fuzzy MOORA method for group decision making. Here, we consider the hesitant fuzzy MOORA under single decision making problem.

Moreover, here we compare our HF-MCDM ranking results with proposed weights with Xu and Zhang, (2013). The comparison results have been showing in the following Table 12, and the proposed standard deviation weights are altered the ranking results as follows.

Table 12. The Comparison results with Xu and Zhang, (2013).

Alternatives	Xu and Zhang, (2013) Weights	Rank of Alternatives	Proposed SDV Weights	Rank of Alternatives
A_1	$w_1 = 0.2341$	5	$w_1 = 0.25$	5
A_2	$w_2 = 0.2474$	3	$w_2 = 0.24$	3
A_3	$w_3 = 0.3181$	2	$w_3 = 0.30$	2
A_4	$w_4 = 0.2004$	1	$w_4 = 0.20$	4
A_5	-	4	-	4

And also we compare our HF-MCDM ranking results with proposed weights with Liao and Xu, (2013), the comparison results have been showing in the following Table 13.

Table 13. The Comparison results with Liao and Xu, (2013) [52].

Alternatives	Liao and Xu, (2013) Weights	Rank of Alternatives	Proposed SDV Weights	Rank of Alternatives
A_1	$w_1 = 0.1$	4	$w_1 = 0.229$	2
A_2	$w_2 = 0.2$	1	$w_2 = 0.23$	1
A_3	$w_3 = 0.4$	3	$w_3 = 0.268$	3
A_4	$w_4 = 0.3$	2	$w_4 = 0.273$	4

9. Sensitivity Analysis

In this sensitivity part, sensitivity analysis is conducted to investigate the impact levels of objective criteria weights. Our main purpose of sensitivity analysis is to analyze the ranking order when changing the significance of objective weights. When changing the importance of selected criteria, we have explain that what changes are happened in the order of alternatives and here, it has been showed by sensitivity analysis.

In our research paper, the sensitivity analysis is shown through following two cases.

Case 1: Equal importance for assigning weights: positive criteria = 0.5 and negative criteria = 0.5.

In our selected water distribution system problem, we have to analyze the alternatives belonging to the five criteria. In these considered criteria, effectiveness(R_1), social(R_3) and environmental production(R_5) are viewed as positive criteria. And economic(R_2), operational time(R_4) are taken as negative criteria. In this Case, the decision-maker selects the best water supply for drought-prone areas, with an emphasis on equity for two negative criteria, the cost and the time to select the best water distribution system. For this case, both positive and negative criteria have consider equal influence and obtain weight values. The weight values of the each criteria is assumed to be as follows,

$$W_1 = 0.20, W_2 = 0.25, W_3 = 0.15, W_4 = 0.25 \text{ and } W_5 = 0.15$$

for each criteria R_1, R_2, R_3, R_4 and R_5 respectively.

The weight values obtained at the end of this case are used in our proposed MCDM methods and the data results are shown in the Table 14. And also the ranking result of this case are shown in the Figure 12.

Table 14. When $W_1 = 0.20$, $W_2 = 0.25$, $W_3 = 0.15$, $W_4 = 0.25$, $W_5 = 0.15$.

Alternatives	HFSDV-MOORA	Rank	HFSDV-TOPSIS	Rank	HFSDV-VIKOR	Rank
P_1	0.2137	2	0.8461	1	0	1
P_2	0.3684	1	0.5174	2	0.1391	2
P_3	−0.1283	4	0.3485	4	0.1844	3
P_4	0.0460	3	0.3505	3	0.2429	4

Case 2: Assigning weights when importance to positive criteria: positive criteria = 1 and negative criteria = 0.

We analyses here on basis of influence of criteria and measure of effectiveness when only positive criteria have maximum weights and others are in minimum weight. In other words, the sensitivity analysis is done by setting the values of positive criteria as 1 and for negative is 0.

In the case of water distribution application, we need to examine the back ground of several criteria to select the best water distribution system. The water scheme mentioned here focuses on the ground water recharge of the drought prone areas, therefore in this case, the higher weight (1) is given to the importance of effectiveness (R_1), social (R_3) and environmental production (R_5). In this Case, the decision-maker selects the best water supply for drought-prone areas, with only three positive criteria: effectiveness, social and environmental protection. Moreover, in this area, the importance of both negative and negative economic parameters, both economic and operational, has been completely ignored. The weight values of the each criteria is assumed to be as follows,

$$W_1 = 0.5, W_2 = 0, W_3 = 0.25, W_4 = 0 \text{ and } W_5 = 0.25$$

for each criteria R_1, R_2, R_3, R_4, R_5 respectively.

Now we identifying the ranking order by use of our proposed MCDM methods with these modified criteria weights, the results of ranking values and alternatives order is given by the Table 15. And also the final results are shown in the Figure 12.

In this research paper we have proposed three methods to select the best method following several positive and a few negative criteria regarding the selection of ground water recharge and drinking water supply systems. And here is where the economy and utility of society should be taken into consideration. This is because these are considered the most central criteria for solving the problem of the peasantry. Both cost and time are a component of the water supply system and are not considered central. When choosing a water supply system for drought prone areas in general, we need to place more emphasis on Effectiveness, Social and Environmental protection to meet the area's water needs. Conversely, both Cost and Operation Time are taken into account in implementing the project, but both are considered insignificant in terms of positive criteria. This is because the decision-maker considers both the Cost and Operation Time of the water supply system to be less important for drinking water and groundwater recharge. Therefore, Case- 2 of the sensitivity analysis we have taken here is considered correct. Weights for the criteria have been identified with the help of the proposed SDV weighting technique with emphasis on positive criteria in the region. And with the help of the HF-MOORA method one of the best water distribution system is selected.

Table 15. When $W_1 = 0.5$, $W_2 = 0$, $W_3 = 0.25$, $W_4 = 0$ and $W_5 = 0.25$.

Alternatives	HFSDV-MOORA	Rank	HFSDV-TOPSIS	Rank	HFSDV-VIKOR	Rank
P_1	1	1	1	1	0	1
P_2	0.427	2	0.0.560	2	0.654	2
P_3	0.093	4	0.258	4	0.950	4
P_4	0.411	3	0.337	3	0.820	3

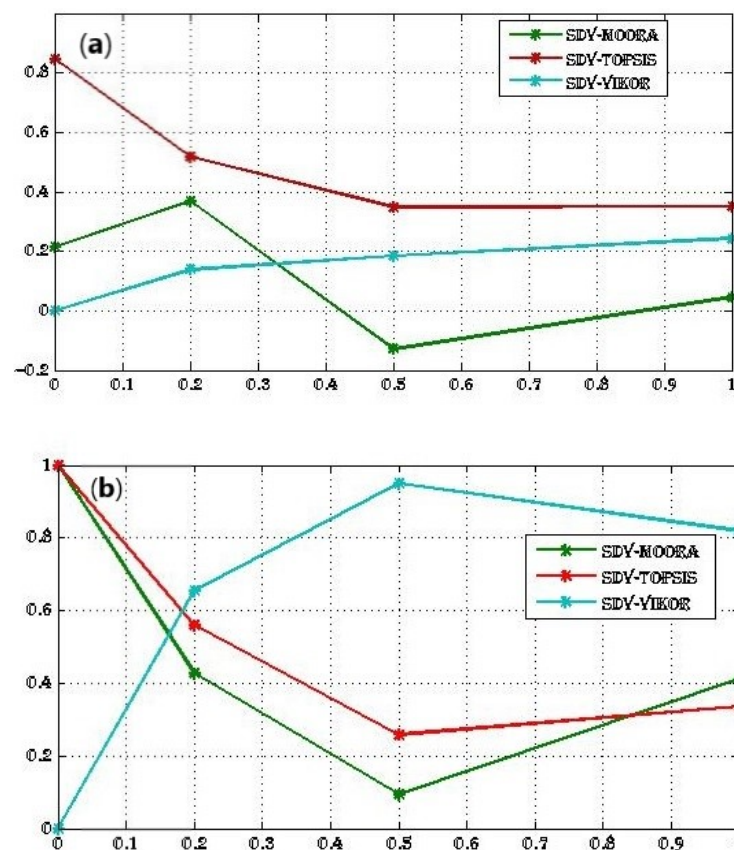


Figure 12. Sensitivity analysis results: (a)—When $W_1 = 0.20, W_2 = 0.25, W_3 = 0.15, W_4 = 0.25, W_5 = 0.15$, (b)—When $W_1 = 0.5, W_2 = 0, W_3 = 0.25, W_4 = 0, W_5 = 0.25$.

10. Conclusions

In this paper, we proposed HFSDV- MOORA, HFSDV-TOPSIS and HFSDV-VIKOR for solving MCDM problems with HFE. The MOORA method is concentrate both positive and negative objectives and giving raking values. The VIKOR method gives compromised solution based on their performance and give utility measure, regret measure and index value. The TOPSIS method is based on PIS and NIS. Here, the above three methods are giving the same ranking for selecting the best alternative. In this regard, the mathematical model is an important tool for the evaluation of water distribution systems and illustrates an efficient implementation of groundwater recharge approach to the drought-prone areas.

HFS is recommended to handle the situation in uncertainty of the unstable set theory. DM gave their idea about the criteria of their corresponding alternatives by HFS. In this research paper, an MCDM is applied to a groundwater irrigation system in the hesitant fuzzy environment. Here, we evaluate four alternatives among five criteria in a hesitant fuzzy environment by use of hesitant fuzzy MOORA, hesitant fuzzy TOPSIS and hesitant fuzzy VIKOR. In our numerical evaluation, we conclude the best water distribution system of the selected problem. Based on a ranking model, we conclude the best water distribution system is closed concrete pipes laid along agricultural land.

Also, we proposed and investigated the SDV weight for manage integrating the values of measurements in different units. Also, we proposed three new MCDM methods with the extension with HFSDV weight. The results of comparison with existing weight finding method proved that the proposed approach is compatible with other MCDM methods. Moreover the proposed methods with HFSDV approach was more reasonable than other weight finding method. These proposed

methods provide the best alternative water distribution system for the ground water irrigation scheme. The solution is rectifying the water problem in drought-prone areas of selected three districts.

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Nomenclature

DM	Decision Making
MCDM	Multi-Criteria Decision Making
MAGDM	Multi- Attribute Group Decision Making
HMCDM	Hesitant Multi-Criteria Decision Making
TMC	Thousand Million Cubic
HFLTS	Hesitant Fuzzy Linguistic Term Set
HFS	Hesitant Fuzzy Set
HFE	Hesitant Fuzzy Element
HFMCDM	Hesitant Fuzzy Multi-Criteria Decision Making
MOORA	Multi objective optimization on the basis of ratio analysis
TOPSIS	Technique for order preference by similarity to an ideal solution
VIKOR	Vlsekriterijumsko KOmpromisno Rangiranje
HFSDV	Hesitant Fuzzy Standard Deviation
SDV	Standard Deviation
WEM	Weighted Entropy Measure
SAW	Simple Additive Weighting
SA	Similarity Analysis
DA	Decision Analysis
DA	Decision Analysis
IVIFN	Interval Valued Intuitionistic Fuzzy Numbers
FUDS	FUzzy Decision System
AASP	Aringar Anna Sugar Project
PIS	Positive Ideal Solution
PIS	Positive Ideal Solution
NIS	Negative Ideal Solution
FPIS	Fuzzy Positive Ideal Solution
FNIS	Fuzzy Negative Ideal Solution

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