

Research Article

An Evaluation Model of Quantitative and Qualitative Fuzzy Multi-Criteria Decision-Making Approach for Location Selection of Transshipment Ports

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The role of container logistics centre as home bases for merchandise transportation has become increasingly important. The container carriers need to select a suitable centre location of transshipment port to meet the requirements of container shipping logistics. In the light of this, the main purpose of this paper is to develop a fuzzy multi-criteria decision-making (MCDM) model to evaluate the best selection of transshipment ports for container carriers. At first, some concepts and methods used to develop the proposed model are briefly introduced. The performance values of quantitative and qualitative subcriteria are discussed to evaluate the fuzzy ratings. Then, the ideal and anti-ideal concepts and the modified distance measure method are used in the proposed model. Finally, a step-by-step example is illustrated to study the computational process of the quantitative and qualitative fuzzy MCDM model. The proposed approach has successfully accomplished our goal. In addition, the proposed fuzzy MCDM model can be empirically employed to select the best location of transshipment port for container carriers in the future study.

1. Introduction

Shipping and port are two major aspects in the sea transport. The issues of shipping and port logistics distribution centres are discussed in the academic literature [1–7] for years. The increasing container transport in liner shipping market has had a vast expansion on the world shipping is growing in importance. In the recent years, due to the blooming development of container shipping in the world, the needs of the transshipment containers in and out of the hub loading centre have been growing rapidly in the Far East. In addition, the governments of Far Eastern countries actively pushed their container ports to become a transshipment centre. Particularly, they engaged in improving computer hardware and software, inland transport systems, and the customs clearance operation and reducing the container handling charges in order to attract more carriers to call their ports as well as to

obtain more transshipment quantity of containers to enlarge their port capacity.

The scopes of shipping logistics services [8] are covering forwarding and consolidation services, logistics operations services, value-added services, warehousing and distribution services, intermodal transport services, information technology (IT) solutions, processing of customs clearance, and specialized services. The container port [9] is a nodal point to handle container cargo to offer value-added services such as collection, warehousing, packing, and distribution among international trade and logistics systems. When the hub-and-spoke network of global container shipping is gradually emerged, the container port in the nodal points has been starting to strengthen its competitive ability to withstand the keen environment. In particular, container transport demands required efficient integrated moves, premium package services, and making the best use of available model

transport operations and container terminals. Due to the importance of container logistics centres as home bases for merchandise transport, the container carriers need to select a suitable location of a transshipment port to meet the customers' requirements.

Furthermore, because a good location [2, 10] will effectively help expand agglomeration economy effects and increase competitive advantages for container ports, as well as the location can help container carriers to be swift the commodities in order to reduce the logistics cost and increase customers' satisfaction. Hence, the container carriers will invest considerable sources for software and hardware facilities subsequently once the location is decided, in which its planning, design, construction, and operation will be also time consuming. In order to satisfy the needs of the container carriers and their customers, there is a need to proceed with a study on the effects from various perspectives and evaluate proper location. Therefore, the location selection of transshipment port for the container carriers is an important issue to study.

The container carriers take many evaluation criteria into consideration while facing the uncertainty environment with keen competition. Due to the quantitative and qualitative characteristics of multiple criteria decision-making (MCDM) [11–15] of location selection of transshipment port and a change in various criteria upon group decision environment, the evaluation problem of location selection of transshipment port is essential to study. Besides, the decision information is hard to come by and is often vague, particularly regarding the linguistic terms. The fuzzy set theory [16] was therefore designed to sort through the uncertainties of vague linguistic terms [17] and helped generate a single possible outcome. Finally, we will propose a quantitative and qualitative fuzzy MCDM method to assist with improving the decision-making quality in this paper.

In summary, the main purpose of this paper is to develop a quantitative and qualitative fuzzy MCDM method to evaluate the problem of location selection of transshipment port for the container carriers. Section 2 presents the research methodologies. The proposed fuzzy MCDM method for evaluating the location of transshipment port is constructed in Section 3. A numerical example is studied in Section 4. Finally, a conclusion is made in the last section.

2. Research Methodologies

In this section, some concepts used to develop an integrated fuzzy MCDM method are introduced. These include the triangular fuzzy numbers and algebraic operations, the linguistic values, the graded mean integration representation (GMIR) method, and the modified distance measure approach, respectively.

2.1. Triangular Fuzzy Numbers and the Algebraic Operations. In a universe of discourse X , a fuzzy subset A of X is defined by a membership function $f_A(x)$, which maps each element x in X to a real number in the interval $[0, 1]$. The function value $f_A(x)$ represents the grade of membership of x in A .

A fuzzy number A [18] in real line \mathfrak{R} is a triangular fuzzy number if its membership function $f_A : \mathfrak{R} \rightarrow [0, 1]$ is as follows

$$f_A(x) = \begin{cases} \frac{(x-l)}{(m-l)}, & l \leq x \leq m, \\ \frac{(x-u)}{(m-u)}, & m \leq x \leq u, \\ 0, & \text{otherwise,} \end{cases} \quad (1)$$

with $-\infty < l \leq m \leq u < \infty$. A triangular fuzzy number can be denoted by (l, m, u) .

The triangular fuzzy numbers are easy to use and easy to interpret. The parameter m gives the maximal grade of $f_A(x)$; that is, $f_A(m) = 1$; it is the most probable value of the evaluation data. In addition, “ l ” and “ u ” are the lower and upper bounds of the available area for the evaluation data. They are used to reflect the fuzziness of the evaluation data. The narrower the interval $[l, u]$, the lower the fuzziness of the evaluation data.

The Zadeh's extension principle [16] and the Chen's function principle [19] are employed to proceed with the algebraic operations of fuzzy numbers. The merit of the function principle not only does not change the type of membership function of fuzzy number after operations, but also can reduce the troublesomeness and tediousness of operations. Hence, we used the Chen's function principle in this paper. Let $A_1 = (l_1, m_1, u_1)$ and $A_2 = (l_2, m_2, u_2)$ be fuzzy numbers. The algebraic operations of any two fuzzy numbers A_1 and A_2 can be expressed as follows.

(1) *Fuzzy Addition.* $A_1 \oplus A_2 = (l_1 + l_2, m_1 + m_2, u_1 + u_2)$, where $l_1, m_1, u_1, l_2, m_2,$ and u_2 are any real numbers.

(2) *Fuzzy Multiplication.* $A_1 \otimes A_2 = (l_1 l_2, m_1 m_2, u_1 u_2)$, where $l_1, m_1, u_1, l_2, m_2,$ and u_2 are all nonzero positive real numbers.

(3) *Fuzzy Division.* (i) $(A_1)^{-1} = (l_1, m_1, u_1)^{-1} = (1/u_1, 1/m_1, 1/l_1)$, where $l_1, m_1,$ and u_1 are all positive real numbers or all negative real numbers.

(ii) $A_1 \oslash A_2 = (l_1/u_2, m_1/m_2, u_1/l_2)$, where $l_1, m_1, u_1, l_2, m_2,$ and u_2 are all nonzero positive real numbers.

2.2. Linguistic Values. In fuzzy decision environments, two preference ratings can be used. They are fuzzy numbers and linguistic values (LVs) characterized by fuzzy numbers [17]. Depending on practical needs, decision makers (DMs) may apply one or both of them. In this paper, the weighting set and preference rating set are used to analytically express the LV and describe how important and how good the involved criteria, subcriteria, and alternatives against various subcriteria above the alternative level are.

In this paper, the weighting set is defined as $W = \{VL, L, M, H, VH\}$ and rating set as $S = \{VP, P, F, G, VG\}$; where VL = Very Low, L = Low, M = Medium, H = High, VH = Very High, VP = Very Poor, P = Poor, F = Fair, G = Good, and VG = Very Good. Here, we define the LVs as VL = VP = (0, 0, 0.2), L = P = (0, 0.2, 0.4), M = F = (0.3, 0.5, 0.7), H = G = (0.6, 0.8, 1), and VH = VG = (0.8, 1, 1), respectively.

2.3. *Graded Mean Integration Representation Method.* In a fuzzy decision-making environment, a defuzzification method of the triangular fuzzy numbers for ranking the alternatives is essential. To match the fuzzy MCDM method developed in this paper and to solve the problem powerfully, the graded mean integration representation (GMIR) method, proposed by Chen and Hsieh [20], is employed to defuzzify the triangular fuzzy numbers.

Let $A_i = (l_i, m_i, u_i)$, $i = 1, 2, \dots, n$, be n triangular fuzzy numbers. By the GMIR method, the GMIR $G(A_i)$ of A_i is as follows:

$$G(A_i) = \frac{l_i + 4m_i + u_i}{6}. \quad (2)$$

Suppose $G(A_i)$ and $G(A_j)$ are the GMIR of the triangular fuzzy numbers A_i and A_j , respectively. We define the following:

$$\begin{aligned} A_i > A_j &\iff G(A_i) > G(A_j), \\ A_i < A_j &\iff G(A_i) < G(A_j), \\ A_i = A_j &\iff G(A_i) = G(A_j). \end{aligned} \quad (3)$$

2.4. *Modified Distance Measure Approach.* Two famous distance measure approaches between two fuzzy numbers, that is, mean and geometrical distance measures, were introduced by Heilpern [21] in 1997. However, Heilpern's method cannot satisfy some special cases between two fuzzy numbers. Hsieh and Chen [22] had proposed the modified geometrical distance approach to improve the drawback of Heilpern's method. To match the integrated fuzzy MCDM method developed in this paper, this modified geometrical distance approach is used to measure the distance of two fuzzy numbers.

Let $A_i = (l_i, m_i, u_i)$ and $A_j = (l_j, m_j, u_j)$ be fuzzy numbers. Then, the Hsieh and Chen's modified geometrical distance can be denoted by the following:

$$\begin{aligned} MD(A_i, A_j) &= \left\{ \frac{1}{4} \left[(l_i - l_j)^2 + 2(m_i - m_j)^2 + (u_i - u_j)^2 \right] \right\}^{1/2}. \end{aligned} \quad (4)$$

3. The Proposed Fuzzy MCDM Method

A stepwise description of the hybrid fuzzy MCDM method for selecting hub location of transshipment port for the container carriers is proposed in the following.

3.1. *Developing a Hierarchical Structure.* A hierarchy structure is the framework of system structure. It can not only be utilized to study the interaction among the elements involved in each layer but also help decisionmakers (DMs) to explore the impact of different elements against the evaluated system. The concepts of hierarchical structure analysis with three

distinct levels; that is, criteria level, subcriteria level, and alternatives level, are used in this paper. Figure 1 shows the complete hierarchical structure of selecting location of transshipment port with k criteria, $n_1 + \dots + n_t + \dots + n_k$ subcriteria, and m alternatives.

With regard to the evaluation criteria and subcriteria, the authors referred to some of the literature, which are made known in academic and management publications [2, 10, 23–34]. Finally, six criteria and twenty-nine subcriteria are suggested. Their codes are shown in the parentheses. In this paper, six quantitative subcriteria (i.e., C_{15} , C_{21} , C_{22} , C_{23} , C_{24} , and C_{25}) are negative, whereas three quantitative ones (i.e., C_{31} , C_{32} , and C_{33}) are positive. However the other twenty subcriteria are qualitative and positive.

- (1) Geographical condition (C_1). This criterion includes “level of closeness to the import/export area (C_{11}),” “level of proximity of the feeder port (C_{12}),” “level of closeness to main navigation route (C_{13}),” “level of frequency of ship calls (C_{14}),” and “delivery time (C_{15}).”
- (2) Cost (C_2). This criterion includes “transportation cost (C_{21}),” “cargo operation cost (C_{22}),” “land cost (C_{23}),” “labour cost (C_{24}),” and “port charges (C_{25}).”
- (3) Economy condition (C_3). This criterion includes “volume of import containers (C_{31}),” “volume of export containers (C_{32}),” “volume of transshipment containers (C_{33}),” “level of economic growth (C_{34}),” and “the other trade variables (C_{35}).”
- (4) Government, law, and social conditions (C_4). This criterion includes “level of private ownership of enterprise (C_{41}),” “level of efficiency of customs (C_{42}),” “level of efficiency of government department (C_{43}),” and “level of social stability (C_{44}).”
- (5) Investment conditions (C_5). This criterion includes “level of tax break and preferential treatment (C_{51}),” “level of law on investment restrictions (C_{52}),” “level of political stability (C_{53}),” “level of land availability and expansion possibility (C_{54}),” and “level of labour quality (C_{55}).”
- (6) Infrastructure conditions (C_6). This criterion includes “level of port facilities (C_{61}),” “level of loading and discharging facilities (C_{62}),” “level of intermodal link (C_{63}),” “level of cargo handling efficiency (C_{64}),” and “level of computer information system (C_{65}).”

3.2. *Estimation of Fuzzy Weights of All Criteria and Subcriteria.* In this paper, the arithmetic mean method is used to obtain the average fuzzy weights of all criteria and subcriteria as well as the fuzzy ratings of alternatives versus all qualitative subcriteria. The LVs of the weighting set and preference rating set, mentioned in the Section 2.2, are assisted in obtaining the fuzzy weights and fuzzy ratings. This is done as follows.

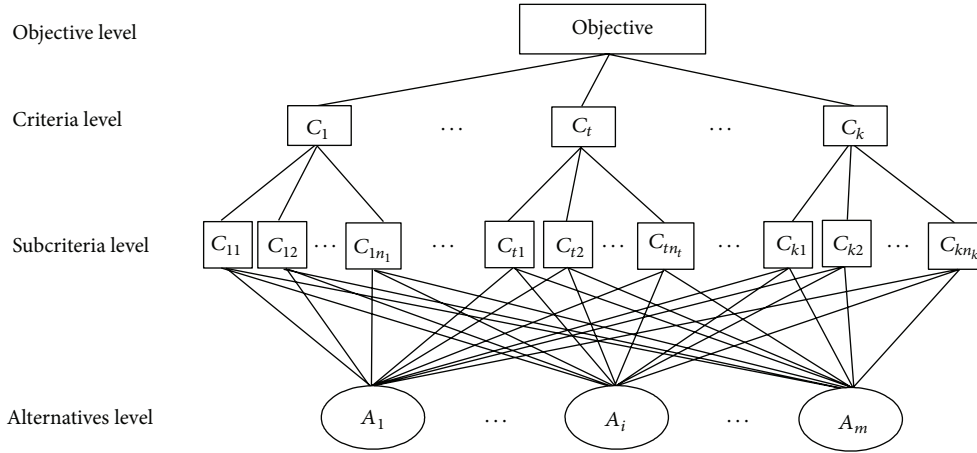


FIGURE 1: The hierarchy structure.

Let $W_t^h = (l_t^h, m_t^h, u_t^h)$, $t = 1, 2, \dots, k$; $h = 1, 2, \dots, E$, be the weight given to criterion C_t by h th DM. Then, the average fuzzy weight of C_t can be represented as follows:

$$W_t = \frac{1}{E} \otimes (W_t^1 \oplus W_t^2 \oplus \dots \oplus W_t^E) \quad (5)$$

$$= (l_t, m_t, u_t),$$

where $l_t = (1/E) \sum_{h=1}^E l_t^h$, $m_t = (1/E) \sum_{h=1}^E m_t^h$, $u_t = (1/E) \sum_{h=1}^E u_t^h$.

Let $W_{tj}^h = (l_{tj}^h, m_{tj}^h, u_{tj}^h)$, $t = 1, 2, \dots, k$; $j = 1, 2, \dots, p_t$; $h = 1, 2, \dots, E$, be the weight given to subcriterion SC_{tj} by h th DM. Then, the average fuzzy weight of SC_{tj} can be represented as follows:

$$W_{tj} = \frac{1}{E} \otimes (W_{tj}^1 \oplus W_{tj}^2 \oplus \dots \oplus W_{tj}^E) \quad (6)$$

$$= (l_{tj}, m_{tj}, u_{tj}),$$

where $l_{tj} = (1/E) \sum_{h=1}^E l_{tj}^h$, $m_{tj} = (1/E) \sum_{h=1}^E m_{tj}^h$, $u_{tj} = (1/E) \sum_{h=1}^E u_{tj}^h$.

3.3. Estimation of Fuzzy Ratings of All Alternatives Versus All Subcriteria. In this paper, the subcriteria are classified into two categories: (1) the subjective criteria, which have linguistic/qualitative definition, for example, level of closeness to the import/export area; and (2) the objective ones, which are defined in monetary/quantitative terms, for example, delivery time or transport cost.

That is, let $S = \{s_1, \dots, s_t, \dots, s_q\}$ and $O = \{o_1, \dots, o_r, \dots, o_p\}$ be the sets of all q qualitative subcriteria and p quantitative ones above the alternatives level. When we measure the fuzzy ratings of all alternatives versus all subcriteria, we face two cases as follows.

Case I (for the qualitative subcriteria). In this paper, the arithmetic mean method is used to obtain the average fuzzy

ratings of alternatives versus all qualitative subcriteria. The LVs of the preference rating set, mentioned in the Section 2.2, are assisted in obtaining the fuzzy ratings. This is done as follows.

Let $S_{itj}^h = (l_{itj}^h, m_{itj}^h, u_{itj}^h)$, $i = 1, 2, \dots, m$; $t = 1, 2, \dots, k$; $j = 1, 2, \dots, p_t$; $h = 1, 2, \dots, E$, be the rating assigned to alternative A_i by h th DM for subcriterion SC_{tj} . Then, the average fuzzy rating of alternative A_i can be represented as follows:

$$S_{itj} = \frac{1}{E} \otimes (S_{itj}^1 \oplus S_{itj}^2 \oplus \dots \oplus S_{itj}^E) \quad (7)$$

$$= (l_{itj}, m_{itj}, u_{itj}),$$

where $l_{itj} = (1/E) \sum_{h=1}^E l_{itj}^h$, $m_{itj} = (1/E) \sum_{h=1}^E m_{itj}^h$, $u_{itj} = (1/E) \sum_{h=1}^E u_{itj}^h$.

Case II (for the quantitative subcriteria). We use the following method [35, 36] to deal with the fuzzy ratings of all alternatives versus all quantitative subcriteria.

- When the appropriateness rating of alternative can be estimated effectively in values, the triangular fuzzy numbers can be used directly. For example, if the transport cost per month is about US Dollars 0.65 million, it can be subjectively expressed as (0.63, 0.65, and 0.68) or (0.61, 0.65, and 0.67).
- If there are historical data, for example, let t_1, t_2, \dots, t_v represent the transport cost of past v periods, the fuzzy rating of the transport cost can be used the geometric mean method to express the following:

$$\left(\min_i \{t_i\}, \left(\prod_{i=1}^v t_i \right)^{1/v}, \max_i \{t_i\} \right). \quad (8)$$

For example, four historical data of the transport cost of alternative A_1 are 0.63, 0.71, 0.54, and 0.69, then based on (8) mentioned above, the evaluation value can be transformed into triangular fuzzy number as (0.54, 0.639, 0.71).

3.4. Calculation of Fuzzy Ideal and Anti-Ideal Solutions. In this paper, the ideal and anti-ideal concepts [37] are used to employ in the proposed fuzzy MCDM algorithm. The logic of ideal and anti-ideal solutions is based on the concept of relative closeness in compliance with the shorter (longer) the distance of alternative i to ideal (anti-ideal), the higher the priority can be ranked.

Firstly, to ensure compatibility between fuzzy ratings of qualitatively positive criteria (or subcriteria) and negative criteria (or subcriteria), the average fuzzy superiority values must be converted to dimensionless indices. The fuzzy ideal values with minimum values in negative subcriteria or maximum values in positive subcriteria should have the maximum rating. Based on the principle stated above, let $S_{itj} = (l_{itj}, m_{itj}, u_{itj})$ ($i = 1, 2, \dots, m; t = 1, 2, \dots, k; j = 1, 2, \dots, p_t$) be the average fuzzy rating value of i th alternative under subcriterion SC_{tj} . Let $\delta_{tj} = \max_i\{u_{itj}\}$, $\varepsilon_{tj} = \min_i\{l_{itj}\}$, then the normalized average fuzzy superiority value λ_{itj} of alternative A_i for subcriterion SC_{tj} can be defined as follows.

- (1) For the positive subcriterion SC_{tj} (the subcriteria that have positive contribution to the objective, i.e., benefit subcriterion):

$$\lambda_{itj} = (x_{itj}, y_{itj}, z_{itj}) = \left(\frac{l_{itj}}{\delta_{tj}}, \frac{m_{itj}}{\delta_{tj}}, \frac{u_{itj}}{\delta_{tj}} \right). \quad (9)$$

- (2) For the negative subcriterion SC_{tj} (the subcriteria that have negative contribution to the objective, i.e., cost subcriterion):

$$\lambda_{itj} = (x_{itj}, y_{itj}, z_{itj}) = \left(\frac{\varepsilon_{tj}}{u_{itj}}, \frac{\varepsilon_{tj}}{m_{itj}}, \frac{\varepsilon_{tj}}{l_{itj}} \right). \quad (10)$$

Then, by using the GMIR method mentioned in Section 2.3, the GMIR value can be express as $G(\lambda_{itj})$. The fuzzy ideal value \tilde{I}_{tj}^+ and fuzzy anti-ideal value \tilde{I}_{tj}^- of each subcriterion above the alternatives layer can be judged and determined by comparing with these GMIR values $G(\lambda_{itj})$. Then,

- (1) if $G(\lambda_{xtj}) = \max_i G(\lambda_{itj})$, then the fuzzy ideal value is

$$\tilde{I}_{tj}^+ = \lambda_{xtj}. \quad (11)$$

- (2) if $G(\lambda_{ytj}) = \min_i G(\lambda_{itj})$, then the fuzzy anti-ideal value is

$$\tilde{I}_{tj}^- = \lambda_{ytj}. \quad (12)$$

Finally, we integrate the fuzzy ideal/anti-ideal values into the fuzzy ideal/anti-ideal solutions. Define the fuzzy ideal solution \tilde{I}^+ and fuzzy anti-ideal solution \tilde{I}^- as follows:

$$\begin{aligned} \tilde{I}^+ &= (\tilde{I}_{11}^+, \tilde{I}_{12}^+, \dots, \tilde{I}_{t1}^+, \dots, \tilde{I}_{tp_t}^+, \dots, \\ &\quad \tilde{I}_{k1}^+, \dots, \tilde{I}_{kp_k}^+), \\ \tilde{I}^- &= (\tilde{I}_{11}^-, \tilde{I}_{12}^-, \dots, \tilde{I}_{t1}^-, \dots, \tilde{I}_{tp_t}^-, \dots, \\ &\quad \tilde{I}_{k1}^-, \dots, \tilde{I}_{kp_k}^-). \end{aligned} \quad (13)$$

TABLE 1: The fuzzy weights of all criteria and subcriteria.

Criteria/subcriteria	DMs	LVs	Fuzzy weights
C_1	A	M	(0.633, 0.833, 0.9)
	B	VH	
	C	VH	
C_2	A	M	(0.5, 0.7, 0.9)
	B	H	
	C	H	
C_3	A	H	(0.3, 0.5, 0.7)
	B	L	
	C	M	
C_4	A	H	(0.5, 0.7, 0.9)
	B	M	
	C	H	
C_5	A	VH	(0.633, 0.833, 0.9)
	B	VH	
	C	M	
C_6	A	M	(0.467, 0.667, 0.8)
	B	M	
	C	VH	
C_{11}	A	H	(0.733, 0.933, 1)
	B	VH	
	C	VH	
C_{12}	A	L	(0.1, 0.3, 0.5)
	B	L	
	C	M	
C_{13}	A	VH	(0.667, 0.867, 1)
	B	H	
	C	H	
C_{14}	A	M	(0.467, 0.667, 0.8)
	B	M	
	C	VH	
C_{15}	A	H	(0.733, 0.933, 1)
	B	VH	
	C	VH	
C_{21}	A	M	(0.4, 0.6, 0.8)
	B	M	
	C	H	
C_{22}	A	L	(0.2, 0.4, 0.6)
	B	H	
	C	L	
C_{23}	A	M	(0.567, 0.767, 0.9)
	B	VH	
	C	H	
C_{24}	A	H	(0.5, 0.7, 0.9)
	B	M	
	C	H	
C_{25}	A	H	(0.733, 0.933, 1)
	B	VH	
	C	VH	
C_{31}	A	VH	(0.633, 0.833, 0.9)
	B	VH	
	C	M	

TABLE 1: Continued.

Criteria/subcriteria	DMs	LVs	Fuzzy weights
C ₃₂	A	VH	(0.633, 0.833, 0.9)
	B	M	
	C	VH	
C ₃₃	A	M	(0.5, 0.7, 0.9)
	B	H	
	C	H	
C ₃₄	A	L	(0.2, 0.4, 0.6)
	B	H	
	C	L	
C ₃₅	A	M	(0.567, 0.767, 0.9)
	B	VH	
	C	H	
C ₄₁	A	H	(0.6, 0.8, 1)
	B	H	
	C	H	
C ₄₂	A	M	(0.567, 0.767, 0.9)
	B	H	
	C	VH	
C ₄₃	A	VH	(0.467, 0.667, 0.8)
	B	M	
	C	M	
C ₄₄	A	VH	(0.667, 0.867, 1)
	B	H	
	C	H	
C ₅₁	A	H	(0.567, 0.767, 0.9)
	B	VH	
	C	M	
C ₅₂	A	H	(0.3, 0.5, 0.7)
	B	L	
	C	M	
C ₅₃	A	VH	(0.633, 0.833, 0.9)
	B	M	
	C	VH	
C ₅₄	A	VH	(0.633, 0.833, 0.9)
	B	VH	
	C	M	
C ₅₅	A	H	(0.6, 0.8, 1)
	B	H	
	C	H	
C ₆₁	A	H	(0.5, 0.7, 0.9)
	B	M	
	C	H	
C ₆₂	A	H	(0.733, 0.933, 1)
	B	VH	
	C	VH	
C ₆₃	A	H	(0.5, 0.7, 0.9)
	B	M	
	C	H	
C ₆₄	A	H	(0.733, 0.933, 1)
	B	VH	
	C	VH	

TABLE 1: Continued.

Criteria/subcriteria	DMs	LVs	Fuzzy weights
C ₆₅	A	VH	(0.667, 0.867, 1)
	B	H	
	C	H	

3.5. *Computation of the Distance of Different Alternatives versus the Fuzzy Ideal/Anti-Ideal Solutions.* As mentioned in Section 3.2, let W_t and W_{tj} , $t = 1, 2, \dots, k$; $j = 1, 2, \dots, p_t$, be the average fuzzy weights of criteria C_t and subcriteria SC_{tj} , respectively. Then the normalized integration weights of the subcriteria SC_{tj} can be obtained by using the GMIR method in Section 2.3, denoted by the following:

$$\Gamma_{tj}^* = \frac{G(W_t)}{\sum_{t=1}^k G(W_t)} \times \frac{G(W_{tj})}{\sum_{j=1}^{p_t} G(W_{tj})}, \quad 0 \leq \Gamma_{tj}^* \leq 1, \quad \sum \Gamma_{tj}^* = 1. \tag{14}$$

Then, compute the distance of different alternatives versus \tilde{I}^+ and \tilde{I}^- which were denoted by D_i^+ and D_i^- , respectively. Define the following:

$$D_i^+ = \sqrt{\sum_{t=1}^k \sum_{j=1}^{p_t} [(\Gamma_{tj}^*)^2 \times (\text{MD}(\tilde{I}_{tj}^+, \lambda_{itj}))^2]}, \quad i = 1, 2, \dots, m, \tag{15}$$

$$D_i^- = \sqrt{\sum_{t=1}^k \sum_{j=1}^{p_t} [(\Gamma_{tj}^*)^2 \times (\text{MD}(\tilde{I}_{tj}^-, \lambda_{itj}))^2]}, \quad i = 1, 2, \dots, m,$$

where $\text{MD}(\cdot)$ can be obtained by using (4) mentioned in Section 2.4.

3.6. *Calculation of the Relative Approximation Value of Different Alternatives Versus Ideal Solution and Ranking the Alternatives.* The relative approximation value (i.e., the relative closeness) of different alternatives A_i versus fuzzy ideal solution \tilde{I}^+ can be calculated, which can be denoted as follows:

$$\text{RC}_i^* = \frac{D_i^-}{D_i^+ + D_i^-}, \quad i = 1, 2, \dots, m. \tag{16}$$

It is obvious that $0 \leq \text{RC}_i^* \leq 1$, $i = 1, 2, \dots, m$. Suppose alternative A_i is an ideal solution (i.e., $D_i^+ = 0$), then $\text{RC}_i^* = 1$. Otherwise, if A_i is an anti-ideal solution (i.e., $D_i^- = 0$), then $\text{RC}_i^* = 0$. The nearer the value RC_i^* to 1, the closer alternative A_i comes near the ideal solution. That is, the maximum value of RC_i^* , then the all alternatives can be ranked. Finally, the best alternative can be selected.

4. The Numerical Example

In this section, a numerical example of selecting hub location of transshipment port for a container carrier is illustrated to

TABLE 2: The original fuzzy ratings of three alternatives versus twenty qualitative/positive subcriteria.

Subcriteria	Alternative DM	Linguistic values			Fuzzy ratings		
		X	Y	Z	X	Y	Z
C ₁₁	A	P	G	P	(0.1, 0.233, 0.433)	(0.733, 0.933, 1)	(0, 0.133, 0.333)
	B	VP	VG	VP			
	C	F	VG	P			
C ₁₂	A	VP	VG	VP	(0.2, 0.267, 0.467)	(0.733, 0.933, 1)	(0, 0, 0.2)
	B	G	G	VP			
	C	VP	VG	VP			
C ₁₃	A	P	P	P	(0.2, 0.333, 0.533)	(0.2, 0.333, 0.533)	(0.2, 0.333, 0.533)
	B	G	G	G			
	C	VP	VP	VP			
C ₁₄	A	F	G	P	(0.567, 0.767, 0.9)	(0.667, 0.867, 1)	(0.267, 0.467, 0.6)
	B	G	G	P			
	C	VG	VG	VG			
C ₃₄	A	G	G	G	(0.467, 0.667, 0.8)	(0.467, 0.667, 0.8)	(0.467, 0.667, 0.8)
	B	VG	VG	VG			
	C	P	P	P			
C ₃₅	A	G	VG	VP	(0.4, 0.6, 0.8)	(0.733, 0.933, 1)	(0, 0.067, 0.267)
	B	G	G	VP			
	C	P	VG	P			
C ₄₁	A	F	F	F	(0.3, 0.5, 0.7)	(0.3, 0.5, 0.7)	(0.2, 0.333, 0.533)
	B	F	F	VP			
	C	F	F	F			
C ₄₂	A	P	P	P	(0.1, 0.233, 0.433)	(0.367, 0.567, 0.7)	(0, 0.133, 0.333)
	B	F	F	P			
	C	VP	VG	VP			
C ₄₃	A	G	G	VP	(0.567, 0.767, 0.9)	(0.667, 0.867, 1)	(0.367, 0.5, 0.633)
	B	F	G	F			
	C	VG	VG	VG			
C ₄₄	A	F	G	F	(0.567, 0.767, 0.9)	(0.733, 0.933, 1)	(0.1, 0.233, 0.433)
	B	VG	VG	VP			
	C	G	VG	P			
C ₅₁	A	VP	VG	P	(0.1, 0.167, 0.367)	(0.367, 0.5, 0.633)	(0, 0.133, 0.333)
	B	F	F	P			
	C	VP	VP	VP			
C ₅₂	A	F	F	VP	(0.1, 0.233, 0.433)	(0.3, 0.5, 0.7)	(0, 0.067, 0.267)
	B	VP	F	VP			
	C	P	F	P			
C ₅₃	A	G	G	G	(0.467, 0.667, 0.8)	(0.467, 0.667, 0.8)	(0.567, 0.767, 0.9)
	B	VG	VG	F			
	C	P	P	VG			
C ₅₄	A	P	G	P	(0, 0.133, 0.333)	(0.467, 0.667, 0.8)	(0, 0.133, 0.333)
	B	VP	VG	VP			
	C	P	P	P			
C ₅₅	A	G	P	G	(0.467, 0.667, 0.8)	(0, 0.133, 0.333)	(0.467, 0.667, 0.8)
	B	VG	VP	VG			
	C	P	P	P			
C ₆₁	A	P	G	G	(0, 0.133, 0.333)	(0.467, 0.667, 0.8)	(0.467, 0.667, 0.8)
	B	VP	VG	VG			
	C	P	P	P			
C ₆₂	A	F	G	F	(0.2, 0.333, 0.533)	(0.467, 0.667, 0.8)	(0.2, 0.333, 0.533)
	B	VP	VG	VP			
	C	F	P	F			

TABLE 2: Continued.

Subcriteria	Alternative DM	Linguistic values			Fuzzy ratings		
		X	Y	Z	X	Y	Z
C ₆₃	A	G	F	P			
	B	VG	VP	VP	(0.467, 0.667, 0.8)	(0.2, 0.333, 0.533)	(0, 0.133, 0.333)
	C	P	F	P			
C ₆₄	A	G	F	G			
	B	VG	VP	VG	(0.467, 0.667, 0.8)	(0.2, 0.333, 0.533)	(0.467, 0.667, 0.8)
	C	P	F	P			
C ₆₅	A	G	P	G			
	B	VG	VP	VG	(0.467, 0.667, 0.8)	(0, 0.133, 0.333)	(0.467, 0.667, 0.8)
	C	P	P	P			

TABLE 3: The original fuzzy superiority of three alternatives versus six quantitative/negative subcriteria.

Subcriteria	Month	Original data			Fuzzy ratings		
		X	Y	Z	X	Y	Z
C ₁₅	January	3.2	3.5	3.4			
	February	2.1	3.3	3.3	(2.9, 3.07, 3.2)	(3.1, 3.27, 3.5)	(3.2, 3.0, 3.4)
	March	2.9	3.2	3.3			
	April	3.1	3.1	3.2			
C ₂₁	January	189	182	173			
	February	182	181	179	(178, 181.9, 189)	(181, 182.5, 184)	(173, 179.9, 190)
	March	179	183	178			
	April	178	184	190			
C ₂₂	January	75	74	72			
	February	72	75	74	(72, 75.95, 79)	(74, 75.49, 78)	(71, 72.98, 75)
	March	78	75	75			
	April	79	78	71			
C ₂₃	January	1235	1211	1285			
	February	1235	1211	1285	(1235, 1235, 1235)	(1211, 1211, 1211)	(1285, 1285, 1285)
	March	1235	1211	1285			
	April	1235	1211	1285			
C ₂₄	January	23	23	21			
	February	24	23	20	(23, 24.47, 26)	(23, 23.74, 25)	(20, 21.47, 23)
	March	25	24	23			
	April	26	25	22			
C ₂₅	January	11	9	10			
	February	12	8	11	(8, 9.87, 12)	(7, 8.21, 9)	(9, 9.98, 11)
	March	9	9	10			
	April	8	7	9			

demonstrate the computational process of the proposed fuzzy MCDM model as follows.

Step 1. Assume that a container carrier needs to select a hub location of transshipment port. Three candidate locations X, Y, and Z are chosen after a preliminary screening for further evaluation. A committee of three DMs (i.e., A, B, and C) is formed to evaluate the best location of transshipment ports among three candidates. Six criteria and twenty-nine subcriteria are suggested in the Section 3.1. It is noted that six quantitative subcriteria (i.e., C₁₅, C₂₁, C₂₂, C₂₃, C₂₄, and C₂₅) are negative, whereas three quantitative ones (i.e., C₃₁,

C₃₂, and C₃₃) are positive. The other twenty subcriteria are qualitative and positive.

Step 2. Three DMs use the LVs (mentioned in the Section 2.2) of weighting sets to evaluate the importance weights. Then, according to (5) and (6), the results of the importance weights are shown in Table 1.

Step 3. Evaluate the fuzzy ratings of three alternatives versus all subcriteria. By using the method presented in Section 3.3, the original preference ratings of twenty qualitative/positive subcriteria, the superiority of six quantitative/negative ones

TABLE 4: The original fuzzy superiority of three alternatives versus three quantitative/positive subcriteria.

Subcriteria	Month	Original data			Fuzzy ratings		
		X	Y	Z	X	Y	Z
C ₃₁	January	476	460	509	(476, 499.3, 515)	(430, 456, 470)	(508, 523.3, 545)
	February	495	470	545			
	March	512	465	508			
	April	515	432	532			
C ₃₂	January	871	715	890	(865, 876.4, 892)	(715, 748.2, 763)	(890, 897.5, 912)
	February	892	763	912			
	March	878	755	893			
	April	865	761	895			
C ₃₃	January	325	285	215	(312, 322.2, 327)	(265, 276.6, 285)	(215, 234.9, 251)
	February	312	276	234			
	March	325	281	251			
	April	327	265	241			

TABLE 5: The NFR and GMIR values of three alternatives versus all subcriteria.

Subcriteria	X		Y		Z	
	NFR	GMIR	NFR	GMIR	NFR	GMIR
C ₁₁	(0.1, 0.233, 0.433)	0.244	(0.733, 0.933, 1)	0.911	(0, 0.133, 0.333)	0.144
C ₁₂	(0.2, 0.267, 0.467)	0.289	(0.733, 0.933, 1)	0.911	(0, 0, 0.2)	0.033
C ₁₃	(0.375, 0.625, 1)	0.646	(0.375, 0.625, 1)	0.646	(0.375, 0.625, 1)	0.646
C ₁₄	(0.567, 0.767, 0.9)	0.756	(0.667, 0.867, 1)	0.856	(0.267, 0.467, 0.6)	0.456
C ₁₅	(0.906, 0.945, 1)	0.948	(0.829, 0.887, 0.935)	0.885	(0.853, 0.967, 0.906)	0.938
C ₂₁	(0.915, 0.951, 0.972)	0.949	(0.940, 0.948, 0.956)	0.948	(0.911, 0.962, 1)	0.960
C ₂₂	(0.899, 0.935, 0.986)	0.938	(0.910, 0.941, 0.959)	0.939	(0.947, 0.973, 1)	0.973
C ₂₃	(0.981, 0.981, 0.981)	0.981	(1, 1, 1)	1	(0.942, 0.942, 0.942)	0.942
C ₂₄	(0.769, 0.817, 0.870)	0.818	(0.80, 0.842, 0.870)	0.840	(0.870, 0.932, 1)	0.933
C ₂₅	(0.583, 0.709, 0.875)	0.716	(0.778, 0.853, 1)	0.865	(0.636, 0.701, 0.778)	0.703
C ₃₁	(0.873, 0.916, 0.945)	0.914	(0.789, 0.837, 0.862)	0.833	(0.932, 0.960, 1)	0.962
C ₃₂	(0.948, 0.961, 0.978)	0.962	(0.784, 0.820, 0.837)	0.817	(0.976, 0.984, 1)	0.985
C ₃₃	(0.954, 0.985, 1)	0.982	(0.810, 0.846, 0.872)	0.844	(0.657, 0.718, 0.768)	0.716
C ₃₄	(0.584, 0.834, 1)	0.820	(0.584, 0.834, 1)	0.820	(0.584, 0.834, 1)	0.820
C ₃₅	(0.4, 0.6, 0.8)	0.60	(0.733, 0.933, 1)	0.911	(0, 0.067, 0.267)	0.089
C ₄₁	(0.429, 0.714, 1)	0.714	(0.429, 0.714, 1)	0.714	(0.286, 0.476, 0.761)	0.492
C ₄₂	(0.143, 0.333, 0.619)	0.349	(0.524, 0.810, 1)	0.794	(0, 0.190, 0.476)	0.206
C ₄₃	(0.567, 0.767, 0.9)	0.756	(0.667, 0.867, 1)	0.856	(0.367, 0.5, 0.633)	0.50
C ₄₄	(0.567, 0.767, 0.9)	0.756	(0.733, 0.933, 1)	0.911	(0.1, 0.233, 0.433)	0.244
C ₅₁	(0.158, 0.264, 0.580)	0.299	(0.580, 0.790, 1)	0.790	(0, 0.210, 0.526)	0.228
C ₅₂	(0.143, 0.333, 0.619)	0.349	(0.429, 0.714, 1)	0.714	(0, 0.096, 0.381)	0.128
C ₅₃	(0.519, 0.741, 0.889)	0.729	(0.519, 0.741, 0.889)	0.883	(0.630, 0.852, 1)	0.840
C ₅₄	(0, 0.166, 0.416)	0.180	(0.584, 0.834, 1)	0.820	(0, 0.166, 0.416)	0.180
C ₅₅	(0.584, 0.834, 1)	0.820	(0, 0.166, 0.416)	0.180	(0.584, 0.834, 1)	0.820
C ₆₁	(0, 0.166, 0.416)	0.180	(0.584, 0.834, 1)	0.820	(0.584, 0.834, 1)	0.820
C ₆₂	(0.25, 0.416, 0.666)	0.430	(0.584, 0.834, 1)	0.820	(0.25, 0.416, 0.666)	0.430
C ₆₃	(0.584, 0.834, 1)	0.820	(0.25, 0.416, 0.666)	0.430	(0, 0.166, 0.416)	0.180
C ₆₄	(0.584, 0.834, 1)	0.820	(0.25, 0.416, 0.666)	0.430	(0.584, 0.834, 1)	0.820
C ₆₅	(0.584, 0.834, 1)	0.820	(0, 0.166, 0.416)	0.180	(0.584, 0.834, 1)	0.820

TABLE 6: The fuzzy ideal and anti-ideal values of all subcriteria.

	Fuzzy ideal values	Fuzzy anti-ideal values
C_{11}	(0.733, 0.933, 1)	(0, 0.133, 0.333)
C_{12}	(0.733, 0.933, 1)	(0, 0, 0.2)
C_{13}	(0.375, 0.625, 1)	(0.375, 0.625, 1)
C_{14}	(0.667, 0.867, 1)	(0.267, 0.467, 0.6)
C_{15}	(0.906, 0.945, 1)	(0.829, 0.887, 0.935)
C_{21}	(0.911, 0.962, 1)	(0.940, 0.948, 0.956)
C_{22}	(0.947, 0.973, 1)	(0.899, 0.935, 0.986)
C_{23}	(1, 1, 1)	(0.942, 0.942, 0.942)
C_{24}	(0.870, 0.932, 1)	(0.769, 0.817, 0.870)
C_{25}	(0.778, 0.853, 1)	(0.636, 0.701, 0.778)
C_{31}	(0.932, 0.960, 1)	(0.789, 0.837, 0.862)
C_{32}	(0.976, 0.984, 1)	(0.784, 0.820, 0.837)
C_{33}	(0.954, 0.985, 1)	(0.657, 0.718, 0.768)
C_{34}	(0.584, 0.834, 1)	(0.584, 0.834, 1)
C_{35}	(0.733, 0.933, 1)	(0, 0.067, 0.267)
C_{41}	(0.429, 0.714, 1)	(0.286, 0.476, 0.761)
C_{42}	(0.524, 0.810, 1)	(0, 0.190, 0.476)
C_{43}	(0.667, 0.867, 1)	(0.367, 0.5, 0.633)
C_{44}	(0.733, 0.933, 1)	(0.1, 0.233, 0.433)
C_{51}	(0.580, 0.790, 1)	(0, 0.210, 0.526)
C_{52}	(0.429, 0.714, 1)	(0, 0.096, 0.381)
C_{53}	(0.519, 0.741, 0.889)	(0.519, 0.741, 0.889)
C_{54}	(0.584, 0.834, 1)	(0, 0.166, 0.416)
C_{55}	(0.584, 0.834, 1)	(0, 0.166, 0.416)
C_{61}	(0.584, 0.834, 1)	(0, 0.166, 0.416)
C_{62}	(0.584, 0.834, 1)	(0.25, 0.416, 0.666)
C_{63}	(0.584, 0.834, 1)	(0, 0.166, 0.416)
C_{64}	(0.584, 0.834, 1)	(0.25, 0.416, 0.666)
C_{65}	(0.584, 0.834, 1)	(0, 0.166, 0.416)

and of three quantitative/positive ones can be obtained, as shown in Tables 2, 3, and 4, respectively.

Step 4. Calculate the fuzzy ideal solution and anti-ideal solution. At first, the original fuzzy ratings and superiority of all subcriteria must be normalized by using the method presented in Section 3.4. The normalized fuzzy rating (NFR) values above the three alternatives and the GMIR values can be obtained. The results can be shown in Table 5.

Then, according to Table 5, the fuzzy ideal value (\tilde{I}_{ij}^+) and fuzzy anti-ideal value (\tilde{I}_{ij}^-) can be obtained by comparing with the GMIR values. The results can be shown in Table 6.

Hence, we can obtain the fuzzy ideal solution (\tilde{I}^+) and fuzzy anti-ideal solution (\tilde{I}^-); that is,

$$\tilde{I}^+ = [(0.733, 0.933, 1), (0.733, 0.933, 1), (0.375, 0.625, 1), \dots, \dots, (0.584, 0.834, 1), (0.584, 0.834, 1), (0.584, 0.834, 1)], \text{ and}$$

$$\tilde{I}^- = [(0, 0.133, 0.333), (0, 0, 0.2), (0.375, 0.625, 1), \dots, \dots, (0, 0.166, 0.416), (0.25, 0.416, 0.666), (0, 0.166, 0.416)].$$

TABLE 7: Distance of three alternatives versus fuzzy ideal and anti-ideal solutions.

Candidates	D_i^+	D_i^-
X	0.003276528	0.002853978
Y	0.001331147	0.006064825
Z	0.00613908	0.001789796

Step 5. Compute the distance of three alternatives versus fuzzy ideal/anti-ideal solutions. Using (14) and (15) proposed in Section 3.5, the results can be shown in Table 7.

Step 6. Calculate the relative closeness value of three alternatives and ranking. Using (16) proposed in Section 3.6, the RCs of three alternatives are $RC_X^* = 0.4655$, $RC_Y^* = 0.820$, and $RC_Z^* = 0.2257$.

The ranking order of RC_i^* for three alternatives is Y, X, and Z, respectively. The best location of transshipment port is obviously Y. Therefore, the committee shall recommend that transshipment port Y be the most appropriate location for the container carriers based on the proposed fuzzy MCDM model.

5. Conclusion

Because the role of container logistics centre as home bases for merchandise transportation has become increasingly important. The container carriers need to select a suitable centre location of transshipment port to meet the requirements of container shipping logistics. The evaluation process of location selection problem of transshipment port involves a multiplicity of complex considerations and poses an MCDM situation. Moreover, some evaluation criteria faced an ambiguous and uncertain nature. Hence, the evaluation of location selection of transshipment port is confronted with a fuzzy decision-making environment. In the light of this, the main purpose of this paper is to develop a hybrid fuzzy MCDM model to evaluate the problem of location selection of transshipment port for the container carriers.

To effectively select best location of transshipment port, a hybrid fuzzy MCDM model is proposed. We develop a hierarchical structure of selecting location of transshipment port with six criteria and twenty-nine subcriteria. The fuzzy weights of all criteria and subcriteria are evaluated. The performance values of quantitative and qualitative subcriteria are discussed to evaluate the fuzzy ratings. Then, the concepts of ideal and anti-ideal solutions are employed in the proposed fuzzy MCDM model. Moreover, Zadeh's linguistic values, Chen and Hsieh's GMIR method, and Hsieh and Chen's modified geometrical distance approach are applied to develop the fuzzy MCDM model. Finally, a step-by-step example is illustrated to study the computational process of the fuzzy MCDM model. In addition, the proposed approach has successfully accomplished our goal. Future study can apply this hybrid fuzzy MCDM model to evaluate the best location selection of transshipment port for container carriers.

In general, the merits of this hybrid fuzzy MCDM model are listed as follows: (1) the quantitative and qualitative subcriteria as well as positive and negative ones are considered in

this approach; (2) the GMIR method and the modified distance method can improve the quality of this fuzzy ideal and anti-ideal algorithm process; and (3) the proposed model not only releases the limitation of crisp values, but also facilitates its implementation as a computer-based decision support system for prioritizing the best location selection in a fuzzy environment.

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