

A novel approach to the priority vector deriving from intuitionistic fuzzy preference relations [☆]

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Abstract

In the typical analytic hierarchy process (AHP), consistency analysis of matrix is very important, so it is very important to analyze the consistency of intuitionistic fuzzy matrix, when decision makers construct matrix by intuitionistic fuzzy number. In this paper, a new approximation-consistency of intuitionistic fuzzy preference relation is proposed. Then several properties related to the consistency of the new definition are given. In considering the randomness of pairwise comparison of alternatives, a new method for calculating intuitionistic fuzzy matrix weight vector is proposed. A new algorithm of solving the decision making problem with intuitionistic fuzzy matrix is proposed. Finally, two numerical examples are calculated by the proposed new definition, and the calculated results are compared with the known ones.

Keywords: Intuitionistic fuzzy matrix; Interval weight; Approximate consistency; Permutation of alternatives

1. Introduction

In 1965, Zadeh[1] proposed fuzzy set theory. The fuzzy set is an extension of the classical Cantor set. At present, the research on fuzzy set theory has been

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very mature. Scholars at home and abroad consider how to expand the theory
 5 of fuzzy set, and various expansion forms emerge, such as interval fuzzy number,
 intuitionistic fuzzy number, trapezoidal fuzzy number, etc. Among them,
 intuitionistic fuzzy set theory is the most active. Intuitionistic fuzzy sets were
 proposed by Atanassov [2]. A series of definitions and algorithms proposed by
 Atanassov [2] laid the foundation of intuitionistic fuzzy set theory. The disadvantage
 10 of fuzzy set theory is that it has only one membership function. When
 the decision is disapproved for scheme, it cannot describe. For example, a voting
 model with yes, no, and no votes. The introduction of intuitionistic fuzzy set
 solves this problem well. The membership function, non-membership function
 and the third parameter of intuitionistic fuzzy set can express the three attitudes
 15 of support, opposition and neutrality. Since intuitionistic fuzzy set theory
 was proposed, some scholars have combined it with Saaty's [3] analytic hierarchy
 process. A large number of scholars at home and abroad have studied the intuitionistic
 fuzzy AHP theory. In 2007, Xu [4] proposed the intuitionistic fuzzy
 matrix and solved the decision-making problem of intuitionistic fuzzy. In analytic
 20 ic hierarchy process, it is very important to obtain the weight of matrix. Xu [5]
 constructed a linear programming model to obtain an accurate weight. Liao [6]
 obtained the precise weights by introducing relaxation variables. Considering
 that intuitionistic fuzzy set are mathematically equivalent to interval fuzzy set,
 Gong [7] obtained the priority of interval values by constructing the minimum
 25 variance model, and Xu [8] obtained the priority weight of interval value of intuitionistic
 fuzzy preference relation by error analysis. From the characteristics
 of intuitionistic fuzzy set itself, Xu [9] obtained the intuitionistic fuzzy priority
 weight through normalized sequential summation. By constructing linear
 programming model, Wang [10] obtained intuitionistic fuzzy weights. Liao [11]
 30 obtained the intuitionistic fuzzy priority weights by constructing a fractional
 programming model. In short, there are three main forms to obtain the weights
 of intuitionistic fuzzy preference relation. The first is the exact weight, which is
 the traditional real weight. The second is the interval weights, which indicates
 the membership range of the importance of the scheme. The third is intuitionis-

35 tic fuzzy weights, which represent the membership and non-membership of the
 importance of the scheme by intuitionistic fuzzy number. In analytic hierar-
 chy process (AHP), the consistency analysis of matrix is also important. The
 consistency of matrix determines whether the final weight is credible or not.
 Therefore, it is very important to study the consistency of intuitionistic fuzzy
 40 matrix. At home and abroad, the additive consistency, multiplicative consis-
 tency, satisfied consistency and weak transfer of intuitionistic fuzzy matrix are
 mainly studied. The research on additive consistency and multiplicative con-
 sistency of intuitionistic fuzzy matrix is the most active. Xu[5] transforms the
 intuitionistic fuzzy matrix into the interval additive reciprocal matrix by using
 45 the transformation formula between the intuitionistic fuzzy preference relation
 and the interval fuzzy preference relation. Xu[5] think the original intuition-
 istic fuzzy matrix with additive consistency or multiplicative consistency, if the
 interval additive reciprocal matrix with additive consistency or multiplicative
 consistency. Gong [7] proposed the additive consistency of intuitionistic fuzzy
 50 matrix by using interval weight vector. According to the definition of additive
 consistency of additive reciprocal matrix, Wang [10] proposed additive consis-
 tency only considering membership degree. At the same time, Wang[10] introduced
 the concept of standardized intuitionistic fuzzy weight vector and defined addi-
 tive consistency. According to the transformation relation between intuitionistic
 55 fuzzy preference relation and interval fuzzy preference relation, Gong [12] pro-
 posed the multiplicative consistency. Xu [13] gives a piecewise multiplicative
 consistency. Liao[11] defined multiplicative consistency by introducing the con-
 cept of standardized intuitionistic fuzzy weight vector. There are few studies
 on the satisfied consistency and weak consistency of intuitionistic fuzzy matrix.
 60 Xu[14] constructed the completely multiplicative consistent matrix, and then
 constructed the consistency index by using the distance between the general
 intuitionistic fuzzy matrix and the completely multiplicative consistent matrix.
 When the consistency index reached the requirements of the decision maker,
 Xu[14] believed that the intuitionistic fuzzy matrix was satisfied consistency.
 65 Xu[15] defined the weak transitivity of intuitionistic fuzzy matrix according to

the weak transitivity of additive reciprocal matrix. In general, there are three kinds of research on consistency of intuitionistic fuzzy matrix. Some people [10] [13] [11] turn intuitionistic fuzzy matrix into interval additive reciprocal matrix, according to the consistency of interval additive reciprocal matrix, consistency of intuitionistic fuzzy matrix is given. Some [7] [12] directly generalize the consistency of additive reciprocal matrix to the intuitionistic fuzzy matrix. Some [4] construct the consistency matrix based on the relationship between the intuitionistic fuzzy matrix and its weight vector.

According to the search of literatures at home and abroad, we find that the consistency definition of intuitionistic fuzzy matrix is mostly given from the traditional additive reciprocal matrix or interval additive reciprocal matrix, but rarely from the characteristics of intuitionistic fuzzy matrix itself. Scholars have given various definitions of consistency without considering the randomness of scheme arrangement. For example, the consistency index of the intuitionistic fuzzy matrix proposed by Xu [14] changes with the change of the order of alternative schemes. When we construct a consistency indicator, we should require that it should not change under any arrangement. In this article, through the transformation formula of between intuitionistic fuzzy preference relation and interval additive preference relation, we transform the intuitionistic fuzzy matrix into the interval additive reciprocal matrix, and put forward the approximate consistency of intuitionistic fuzzy matrix considering the arrangement of schemes, and give a new method to obtain the interval weight vector of intuitionistic fuzzy matrix. A new algorithm for solving decision-making problems with intuitionistic fuzzy matrix has been given. Two examples have been carried out to illustrate the proposed definition, algorithm and some comparisons have been made.

The structure of this paper is shown as follows. In Section 2, we analyze the definition of consistent of intuitionistic fuzzy matrix. In Section 3, a new concept of approximation-consistency of intuitionistic fuzzy matrix is proposed. In Section 4, the method of obtaining the interval weight vector of intuitionistic fuzzy matrix is given. In Section 5, we offer a new algorithm to the decision-

making problem with intuitionistic fuzzy matrix. Two examples are carried out to illustrate the new definition and algorithm. The main conclusions are shown in Section 6.

100 2. Preliminaries

In 1965, Zadeh[1] proposed the fuzzy set theory, and in 1986, Atanassov[2] extended the fuzzy set and proposed the intuitionistic fuzzy set. Firstly, we give the concept of fuzzy set and intuitionistic fuzzy set:

Definition 1. [1] Let X is a non-empty set, F is constituted by μ_F whose value is $[0,1]$, where $\mu_F : X \rightarrow [0, 1]$, we call F as fuzzy set on X .
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Definition 2. [2] Let X is a non-empty set, we call

$A = \{ \langle x, \mu_A(x), v_A(x) \rangle \mid x \in X \}$ is the intuitionistic fuzzy set, where $0 \leq \mu_A(x) \leq 1, 0 \leq v_A(x) \leq 1, 0 \leq \mu_A(x) + v_A(x) \leq 1$.

among them, $\mu_A(x)$ and $v_A(x)$ are the membership and non-membership of the element x , and $\pi_A(x) = 1 - \mu_A(x) - v_A(x)$ represents the hesitation x belonging to A .
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Definition 3. [15] Let $\alpha = (\mu_\alpha, \nu_\alpha)$ an intuitionistic fuzzy number, where $0 \leq \mu_\alpha \leq 1, 0 \leq \nu_\alpha \leq 1, 0 \leq \mu_\alpha + \nu_\alpha \leq 1$. The score function and accurate function are defined as

$$S(\alpha) = \mu_\alpha - \nu_\alpha, H(\alpha) = \mu_\alpha + \nu_\alpha. \quad (1)$$

Based on the above functions, Xu[15] proposed the following way to compare the sizes of intuitionistic fuzzy number $a_i = (\mu_i, \nu_i)$ and $a_j = (\mu_j, \nu_j)$:

Step 1. Calculate the S values of the IFVs $a_i = (\mu_i, \nu_i)$ and $a_j = (\mu_j, \nu_j)$ using
115 1.

Step 2. If $S(\alpha_i) < S(\alpha_j)$, then $\alpha_i < \alpha_j$, then go to Step 3.

Step 3. Calculate the accuracy degrees of these IFVs using the accuracy function $H(\alpha) = \mu_\alpha + \nu_\alpha$, and then rank the IFVs according to the following principles:
(1) If $H(\alpha_i) > H(\alpha_j)$, then $\alpha_i > \alpha_j$;

120 (2) If $H(\alpha_i) = H(\alpha_j)$, then $\alpha_i = \alpha_j$.

Beliakov[?] pointed out that the above ordering scheme is not closed to the scalar product of intuitionistic fuzzy numbers. In other words, $\alpha_i < \alpha_j$ does not necessarily mean $\lambda\alpha_i < \lambda\alpha_j$, where λ is a scalar. Based on the similarity function and the exact function, Zhang[?] proposed a full order scheme to
125 compare any two intuitionistic fuzzy numbers.

Definition 4. [?] Let $\alpha = (\mu_\alpha, \nu_\alpha)$ an intuitionistic fuzzy number, where $0 \leq \mu_\alpha \leq 1, 0 \leq \nu_\alpha \leq 1, 0 \leq \mu_\alpha + \nu_\alpha \leq 1$. The similarity function and accurate function are defined as

$$L(\alpha) = \frac{1 - \nu_\alpha}{1 + \pi_\alpha}, H(\alpha) = \mu_\alpha + \nu_\alpha. \quad (2)$$

Based on the above functions, Zhang[?] proposed the following way to compare the sizes of intuitionistic fuzzy number $a_i = (\mu_i, \nu_i)$ and $a_j = (\mu_j, \nu_j)$:

Step 1. Calculate the L values of the IFVs $a_i = (\mu_i, \nu_i)$ and $a_j = (\mu_j, \nu_j)$ using 2.

130 Step 2. If $L(\alpha_i) < L(\alpha_j)$, then $\alpha_i < \alpha_j$, then go to Step 3.

Step 3. Calculate the accuracy degrees of these IFVs using the accuracy function $H(\alpha) = \mu_\alpha + \nu_\alpha$, and then rank the IFVs according to the following principles:

(1) If $H(\alpha_i) > H(\alpha_j)$, then $\alpha_i > \alpha_j$;

(2) If $H(\alpha_i) = H(\alpha_j)$, then $\alpha_i = \alpha_j$.

135 The algorithm for intuitionistic fuzzy number is given below:

Definition 5. [15] If $a_{ij} = (\mu_{ij}, \nu_{ij})$ and $a_{kl} = (\mu_{kl}, \nu_{kl})$ are two intuitionistic fuzzy number, then, we have the follow algorithm :

$$\bar{a}_{ij} = (\nu_{ij}, \mu_{ij}). \quad (3)$$

$$a_{ij} + a_{kl} = (\mu_{ij} + \mu_{kl} - \mu_{ij} \cdot \mu_{kl}, \nu_{ij} \cdot \nu_{kl}). \quad (4)$$

$$a_{ij} \cdot a_{kl} = (\mu_{ij} \cdot \mu_{kl}, \nu_{ij} + \nu_{kl} - \nu_{ij} \cdot \nu_{kl}). \quad (5)$$

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$$\lambda a_{ij} = \left(1 - (1 - \mu_{ij})^\lambda, \nu_{ij}^\lambda\right), \lambda > 0. \quad (6)$$

$$a_{ij}^\lambda = \left(\mu_{ij}^\lambda, 1 - (1 - v_{ij})^\lambda \right), \lambda > 0. \quad (7)$$

In what follow, for the set of alternatives $X = \{x_1, x_2, \dots, x_n\}$, the definition of the additive reciprocal matrix $B = (b_{ij})_{n \times n}$ and its consistency are given:

Definition 6. [16] $B = (b_{ij})_{n \times n}$ is called as an additive reciprocal preference
 145 relation, if $b_{ij} + b_{ji} = 1$, $0 \leq b_{ij} \leq 1$, and $b_{ii} = 0.5$ for all $i, j = 1, 2, \dots, n$.

Then by considering the additive transitivity, we has the following definition of additive consistency:

Definition 7. [16] Additive reciprocal preference relation $B = (b_{ij})_{n \times n}$ is additively consistent, if $b_{ij} = b_{ik} - b_{jk} + 0.5$, for all $i, j, k \in \{1, 2, \dots, n\}$.

150 In addition, the multiplicative consistency of additive reciprocal matrix $B = (b_{ij})_{n \times n}$ is defined as follows:

Definition 8. [16] Additive reciprocal matrix $B = (b_{ij})_{n \times n}$ is multiplicative consistency, if

$$b_{ij} \cdot b_{jk} \cdot b_{ki} = b_{ik} \cdot b_{kj} \cdot b_{ji}, \quad (8)$$

for all $i, j, k \in \{1, 2, \dots, n\}$.

The interval additive reciprocal matrix is a generalization of the additive reciprocal matrix. We know that an interval additive reciprocal matrix can be
 155 used to express the opinions of decision makers, the corresponding definition is given as follows:

Definition 9. [17] An interval additive reciprocal matrix $R = (r_{ij})_{n \times n}$ is defined as

$$R = (r_{ij})_{n \times n} = \begin{bmatrix} [0.5, 0.5] & [r_{12}^-, r_{12}^+] & \cdots & [r_{1n}^-, r_{1n}^+] \\ [r_{21}^-, r_{21}^+] & [0.5, 0.5] & \cdots & [r_{2n}^-, r_{2n}^+] \\ \vdots & \vdots & \ddots & \vdots \\ [r_{n1}^-, r_{n1}^+] & [r_{n2}^-, r_{n2}^+] & \cdots & [0.5, 0.5] \end{bmatrix}, \quad (9)$$

where $r_{ij}^- \leq r_{ij}^+$, $r_{ii}^- = r_{ii}^+ = 0.5$, $r_{ij}^- + r_{ji}^+ = 1$ and $r_{ij}^+ + r_{ji}^- = 1$ for $i, j = 1, 2, \dots, n$.

In what follow, we introduce the intuitionistic fuzzy matrix. A finite set of objects $X = \{x_1, x_2, \dots, x_n\}$, $a_{ij} = (\mu_{ij}, \nu_{ij})$ is an intuitionistic fuzzy number, where μ_{ij} is the preference intensity of the alternatives x_i to x_j , and ν_{ij} is the preference intensity of the alternatives x_j to x_i , where

$$\mu_{ij}, \nu_{ij} \in [0, 1], \mu_{ij} + \nu_{ij} \leq 1. \quad (10)$$

In addition, $\pi_{ij} = 1 - \mu_{ij} - \nu_{ij}$ is a hesitancy degree for decision maker, all preference values a_{ij} constitute decision matrix $A = (a_{ij})_{n \times n}$.

We use intuitionistic fuzzy matrix $A = (a_{ij})_{n \times n}$ to express the opinions of decision makers, the corresponding definition is given as follows:

Definition 10. [4] An intuitionistic fuzzy matrix $A = (a_{ij})_{n \times n}$ is defined as

$$A = (a_{ij})_{n \times n} = \begin{pmatrix} (0.5, 0.5) & (\mu_{12}, \nu_{12}) & \cdots & (\mu_{1n}, \nu_{1n}) \\ (\mu_{21}, \nu_{21}) & (0.5, 0.5) & \cdots & (\mu_{2n}, \nu_{2n}) \\ \vdots & \vdots & \ddots & \vdots \\ (\mu_{n1}, \nu_{n1}) & (\mu_{n2}, \nu_{n2}) & \cdots & (0.5, 0.5) \end{pmatrix}, \quad (11)$$

where

$$\mu_{ij}, \nu_{ij} \in [0, 1], \mu_{ij} + \nu_{ij} \leq 1, \mu_{ij} = \nu_{ji}, \mu_{ji} = \nu_{ij}, \mu_{ii} = \nu_{ii} = 0.5. \quad (12)$$

We know that $\mu_{ik} \leq 1 - \nu_{ik}$ ($i, k = 1, 2, \dots, n$) is always true for any intuitionistic fuzzy number, so we can transform the intuitionistic fuzzy matrix

$$A = (a_{ik})_{n \times n} = \begin{pmatrix} (0.5, 0.5) & (\mu_{12}, \nu_{12}) & \cdots & (\mu_{1n}, \nu_{1n}) \\ (\mu_{21}, \nu_{21}) & (0.5, 0.5) & \cdots & (\mu_{2n}, \nu_{2n}) \\ \vdots & \vdots & \ddots & \vdots \\ (\mu_{n1}, \nu_{n1}) & (\mu_{n2}, \nu_{n2}) & \cdots & (0.5, 0.5) \end{pmatrix}, \quad (13)$$

into the interval additive reciprocal matrix

$$R = (r_{ik})_{n \times n} = \begin{bmatrix} [0.5, 0.5] & [r_{12}^-, r_{12}^+] & \cdots & [r_{1n}^-, r_{1n}^+] \\ [r_{21}^-, r_{21}^+] & [0.5, 0.5] & \cdots & [r_{2n}^-, r_{2n}^+] \\ \vdots & \vdots & \ddots & \vdots \\ [r_{n1}^-, r_{n1}^+] & [r_{n2}^-, r_{n2}^+] & \cdots & [0.5, 0.5] \end{bmatrix}, \quad (14)$$

where

$$r_{ik}^- = \mu_{ik}, r_{ik}^+ = 1 - v_{ik}, \quad (15)$$

Then we arrive at the following theorem:

Theorem 1. *Let $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) the intuitionistic fuzzy preference relation, for interval additive reciprocal matrix $R = (r_{ik})_{n \times n}$ by (15). If and only if $\mu_{ik} + \mu_{kt} + \mu_{ti} = \mu_{tk} + \mu_{ki} + \mu_{it}$, $\nu_{ik} + \nu_{kt} + \nu_{ti} = \nu_{tk} + \nu_{ki} + \nu_{it}$, we have $r_{ik}^- + r_{kt}^- + r_{ti}^- = r_{tk}^- + r_{ki}^- + r_{it}^-$, $r_{ik}^+ + r_{kt}^+ + r_{ti}^+ = r_{tk}^+ + r_{ki}^+ + r_{it}^+$, $i, t, k = 1, 2, \dots, n$.*

Proof. When $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) is the intuitionistic fuzzy preference relation, we can obtain that interval additive reciprocal matrix $R = (r_{ik})_{n \times n}$ by (15). It is noted from Definition (10) that

$$r_{ik}^- = \mu_{ik}, r_{ik}^+ = 1 - v_{ik}.$$

If

$$r_{ik}^- + r_{kt}^- + r_{ti}^- = r_{tk}^- + r_{ki}^- + r_{it}^-, r_{ik}^+ + r_{kt}^+ + r_{ti}^+ = r_{tk}^+ + r_{ki}^+ + r_{it}^+,$$

We have

$$\mu_{ik} + \mu_{kt} + \mu_{ti} = r_{ik}^- + r_{kt}^- + r_{ti}^- = r_{tk}^- + r_{ki}^- + r_{it}^- = \mu_{tk} + \mu_{ki} + \mu_{it},$$

$$\nu_{ik} + \nu_{kt} + \nu_{ti} = 1 - r_{ik}^+ + 1 - r_{kt}^+ + 1 - r_{ti}^+ = 1 - r_{tk}^+ + 1 - r_{ki}^+ + 1 - r_{it}^+ = \nu_{tk} + \nu_{ki} + \nu_{it}.$$

On the contrary, if

$$\mu_{ik} + \mu_{kt} + \mu_{ti} = \mu_{tk} + \mu_{ki} + \mu_{it}, \nu_{ik} + \nu_{kt} + \nu_{ti} = \nu_{tk} + \nu_{ki} + \nu_{it}.$$

We have

$$r_{ik}^- + r_{kt}^- + r_{ti}^- = \mu_{ik} + \mu_{kt} + \mu_{ti} = \mu_{tk} + \mu_{ki} + \mu_{it} = r_{tk}^- + r_{ki}^- + r_{it}^-,$$

$$r_{ik}^+ + r_{kt}^+ + r_{ti}^+ = 1 - \nu_{ik} + 1 - \nu_{kt} + 1 - \nu_{ti} = 1 - \nu_{tk} + 1 - \nu_{ki} + 1 - \nu_{it} = r_{tk}^+ + r_{ki}^+ + r_{it}^+.$$

In the process of transforming interval additive reciprocal judgment matrix into intuitionistic fuzzy judgment matrix, the formula is very effective if some good

properties can be preserved. For example, The advantages and disadvantages of the scheme and the weak transitivity are preserved. Probability is a way of comparing the sizes of interval numbers. So let's give you a definition of what is possible:

Definition 11. [17] Let $\tilde{w}_i = [w_i^-, w_i^+]$ and $\tilde{w}_j = [w_j^-, w_j^+]$ are the interval number.

$$p(\tilde{w}_i \geq \tilde{w}_j) = \max \left\{ 1 - \max \left\{ \frac{w_j^+ - w_i^-}{w_i^+ - w_i^- + w_j^+ - w_j^-}, 0 \right\}, 0 \right\} \quad (16)$$

is the probability of $\tilde{w}_i \geq \tilde{w}_j$

Possible degrees $p(\tilde{w}_i \geq \tilde{w}_j)$ satisfy the following properties:

$$0 \leq p(\tilde{w}_i \geq \tilde{w}_j) \leq 1. \quad (17)$$

$$p(\tilde{w}_i \geq \tilde{w}_j) + p(\tilde{w}_j \geq \tilde{w}_i) = 1. \quad (18)$$

$$p(\tilde{w}_i \geq \tilde{w}_j) = 1 \Leftrightarrow w_j^+ \leq w_i^-. \quad (19)$$

$$p(\tilde{w}_i \geq \tilde{w}_j) = 0 \Leftrightarrow w_i^+ \leq w_j^-. \quad (20)$$

$$p(\tilde{w}_i \geq \tilde{w}_j) \geq 0.5, p(w_j \geq w_k) \geq 0.5 \Rightarrow p(w_i \geq w_k) \geq 0.5 \quad (21)$$

$$p(\tilde{w}_i \geq \tilde{w}_j) > 0.5 \Leftrightarrow w_i^+ + w_i^- > w_j^+ + w_j^-. \quad (22)$$

$$p(\tilde{w}_i \geq \tilde{w}_j) = 0.5 \Leftrightarrow w_i^+ + w_i^- = w_j^+ + w_j^-. \quad (23)$$

$P(\tilde{w}_i \geq \tilde{w}_j) > 0.5$ means that w_i is better than w_j , $p(\tilde{w}_i \geq \tilde{w}_j) < 0.5$ means that w_i is worse than w_j , $p(\tilde{w}_i \geq \tilde{w}_j) = 0.5$ means that there is no difference between w_i and w_j .

By means of probability, we can describe the advantages and disadvantages of pairwise comparison of schemes and have the following theorems:

Theorem 2. Let interval additive reciprocal matrix $R = (r_{ik})_{n \times n}$, for the intuitionistic fuzzy preference relation $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) is constructed by (15).

- (1) If $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) > 0.5$, then $S_{ij} - S_{ji} > 0$.
- (2) If $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) = 0.5$, then $S_{ij} - S_{ji} = 0$.
- (3) If $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) < 0.5$, then $S_{ij} - S_{ji} < 0$.

Proof. Under the condition that $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) > 0.5$, we obtain $u_{ij}^- + u_{ij}^+ > u_{jt}^- + u_{ji}^+$. Since $r_{ij}^- = \mu_{ij}$, $r_{ij}^+ = 1 - v_{ij}$, $u_{ij}^- + u_{ij}^+ > u_{jt}^- + u_{ji}^+$ is equal to $\mu_{ij} - v_{ij} > \mu_{ji} - v_{ji}$. It means that $S_{ij} - S_{ji} > 0$. In the same way, under the condition that $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) = 0.5$ and $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) < 0.5$, we have $S_{ij} - S_{ji} = 0$ and $S_{ij} - S_{ji} < 0$.

Theorem 3. Let interval additive reciprocal matrix $R = (r_{ik})_{n \times n}$, for the intuitionistic fuzzy preference relation $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) is constructed by (15).

- (1) If $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) > 0.5$, then $L_{ij} - L_{ji} > 0$.
- (2) If $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) = 0.5$, then $L_{ij} - L_{ji} = 0$.
- (3) If $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) < 0.5$, then $L_{ij} - L_{ji} < 0$.

Proof. Since $L_{ij} > L_{ji}$ is equal to

$$\frac{1 - v_{ij}}{2 - \mu_{ij} - v_{ij}} > \frac{1 - v_{ji}}{2 - \mu_{ji} - v_{ji}},$$

\Downarrow

$$(1 - v_{ij})(2 - \mu_{ji} - v_{ji}) > (1 - v_{ji})(2 - \mu_{ij} - v_{ij}),$$

\Downarrow

$$\mu_{ij} - v_{ij} > \mu_{ji} - v_{ji}.$$

Under the condition that $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) > 0.5$, we obtain $u_{ij}^- + u_{ij}^+ > u_{jt}^- + u_{ji}^+$. It means that $L_{ij} - L_{ji} > 0$. In the same way, under the condition that $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) = 0.5$ and $p([u_{ij}^-, u_{ij}^+] \geq [u_{jt}^-, u_{ji}^+]) < 0.5$, we have $L_{ij} - L_{ji} = 0$ and $L_{ij} - L_{ji} < 0$.

215 **Theorem 4.** *Let interval additive reciprocal matrix $R = (r_{ik})_{n \times n}$, for the intuitionistic fuzzy preference relation $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) is constructed by (15). When the interval additive reciprocal matrix is converted into fuzzy complementary judgment matrix, the weak transitivity remains by (1).*

Proof. Since $R = (r_{ik})_{n \times n}$ satisfies weak transfer, under the condition

$$p([u_{ij}^-, u_{ij}^+] \geq [u_{ji}^-, u_{ji}^+]) \geq 0.5, p([u_{jk}^-, u_{jk}^+] \geq [u_{kj}^-, u_{kj}^+]) \geq 0.5,$$

we obtain

$$p([u_{ij}^-, u_{ij}^+] \geq [u_{ji}^-, u_{ji}^+]) \geq 0.5.$$

According to the theorem 2, we have $S_{ik} \geq 0.5$, $S_{kj} \geq 0.5$, $S_{ik} \geq 0.5$. So the
220 weak transitivity remains (1).

Theorem 5. *Let interval additive reciprocal matrix $R = (r_{ik})_{n \times n}$, for the intuitionistic fuzzy preference relation $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) is constructed by (15). When the interval additive reciprocal matrix is converted into fuzzy complementary judgment matrix, the weak transitivity remains by (2).*

Proof. Since $R = (r_{ik})_{n \times n}$ satisfies weak transfer, under the condition

$$p([u_{ij}^-, u_{ij}^+] \geq [u_{ji}^-, u_{ji}^+]) \geq 0.5, p([u_{jk}^-, u_{jk}^+] \geq [u_{kj}^-, u_{kj}^+]) \geq 0.5,$$

we obtain

$$p([u_{ij}^-, u_{ij}^+] \geq [u_{ji}^-, u_{ji}^+]) \geq 0.5.$$

225 According to the theorem 3, we have $L_{ik} \geq 0.5$, $L_{kj} \geq 0.5$, $L_{ik} \geq 0.5$. So the weak transitivity remains (2).

The study of consistency is very important in the decision-making problem, according to the consistency of additive reciprocal matrix or interval additive reciprocal matrix, scholars put forward the consistency of intuitionistic fuzzy matrix, the
230 relevant definitions are as follows:

Definition 12. [5] Let $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) the intuitionistic fuzzy preference relation, we call $A = (a_{ik})_{n \times n}$ additive consistency,

if $\mu_{ik} \leq 0.5 (\omega_i - \omega_k + 1) \leq 1 - v_{ik}$, $i = 1, 2, \dots, n-1, k = i+1, \dots, n$, where $\omega = (\omega_1, \omega_2, \dots, \omega_n)^T$ is an interval vector.

In addition, Gong[7] believed that the condition for the additive consistency of the interval additive reciprocal matrix $R = (r_{ik})_{n \times n}$ is the existence of the interval weight vector

$$\hat{\omega} = (\hat{\omega}_1, \hat{\omega}_2, \dots, \hat{\omega}_n)^T = ([\omega_1^l, \omega_1^u], [\omega_2^l, \omega_2^u], \dots, [\omega_n^l, \omega_n^u])^T,$$

where

$$r_{ik} = (0.5 + 0.2 \log_3 \omega_i^l / \omega_k^u, 0.5 + 0.2 \log_3 \omega_i^u / \omega_k^l, i, k = 1, 2, \dots, n).$$

235 Based on between the intuitionistic fuzzy preference relation $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})$ and interval fuzzy preference relation $R = (r_{ik})_{n \times n}$ conversion formula. According to the constraint relationship between the interval weight vector and the additive consistency of $A = (a_{ik})_{n \times n}$, Gong [7] proposed the following additive consistency:

Definition 13. [7] Let $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) the intuitionistic fuzzy preference relation, we call $A = (a_{ik})_{n \times n}$ additive consistency, if $\mu_{ik} = 0.5 + 0.2 \log_3 \omega_i^l / \omega_k^u$, $v_{ik} = 0.5 + 0.2 \log_3 \omega_k^l / \omega_i^u$, $i, k = 1, 2, \dots, n$, where

$$\tilde{\omega} = (\tilde{\omega}_1, \tilde{\omega}_2, \dots, \tilde{\omega}_n)^T = ((\omega_1^\mu, \omega_1^v), (\omega_2^\mu, \omega_2^v), \dots, (\omega_n^\mu, \omega_n^v))^T,$$

240 is an interval weight vector.

Both definition 12 and 13 are obtained by converting intuitionistic fuzzy preference relation to interval additive reciprocal preference relation. In addition, some scholars give the definition of consistency from the characteristics of intuitionistic fuzzy number. Wang[10] by considering the membership and non-
245 membership of intuitionistic fuzzy number to define the additive consistency of intuitionistic fuzzy preference relation:

Definition 14. [10] Let $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) the intuitionistic fuzzy preference relation, we call $A = (a_{ik})_{n \times n}$ additive consistency. If

$$\mu_{ik} + \mu_{kt} + \mu_{ti} = \mu_{tk} + \mu_{ki} + \mu_{it}, \quad i, t, k = 1, 2, \dots, n.$$

In addition to the additive consistency, the additive reciprocal matrix also has the multiplicative consistency. By extending the multiplicative consistency of additive reciprocal matrix, Xu[5] has proposed the multiplicative consistency of the intuitionistic fuzzy matrix:

Definition 15. [5] Let $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) the intuitionistic fuzzy preference relation, we call $A = (a_{ik})_{n \times n}$ multiplicative consistency. If

$$\mu_{ik} \leq \frac{\omega_i}{\omega_i + \omega_k} \leq 1 - v_{ik}, \quad i = 1, 2, \dots, m-1, \quad k = i+1, \dots, n,$$

where $\omega = (\omega_1, \omega_2, \dots, \omega_n)^T$ is a set of interval weight vector.

According to the transformation rule between intuitionistic fuzzy preference relation $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ and interval fuzzy preference relation $R = (r_{ik})_{n \times n}$, Gong[12] put forward multiplicative consistency of intuitionistic fuzzy preference relation:

Definition 16. [12] Let $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$)

the intuitionistic fuzzy preference relation, we call $A = (a_{ik})_{n \times n}$ multiplicative consistency, if

$$\mu_{ik} = \frac{\omega_i^l}{\omega_i^l + \omega_k^u}, \quad v_{ik} = \frac{\omega_k^l}{\omega_k^l + \omega_i^u}, \quad i, k = 1, 2, \dots, n,$$

where

$$\tilde{\omega} = (\tilde{\omega}_1, \tilde{\omega}_2, \dots, \tilde{\omega}_n)^T = ((\omega_1^\mu, \omega_1^v), (\omega_2^\mu, \omega_2^v), \dots, (\omega_n^\mu, \omega_n^v))^T,$$

is a set of interval weight vector.

In addition, Liao[11] directly defined the multiplicative consistency of intuitionistic fuzzy preference relation by membership degree and non-membership degree of intuitionistic fuzzy number:

Definition 17. [11] Let $A = (a_{ik})_{n \times n} = (\mu_{ik}, v_{ik})_{n \times n}$ ($i, k = 1, 2, \dots, n$) the intuitionistic fuzzy preference relation, we call $A = (a_{ik})_{n \times n}$ multiplicative consistency, if $\mu_{ik} \cdot \mu_{kt} \cdot \mu_{ti} = v_{ik} \cdot v_{kt} \cdot v_{ti}$, $i, t, k = 1, 2, \dots, n$

Consistency is a very ideal case. In practice, we consider more of acceptable consistency. In the literature[14], the construction of completely multiplicative consistency matrix and acceptable multiplicative consistency are proposed, the definition of completely multiplicative consistency matrix is given as follows:

Definition 18. [14] In a fixed set of $X = \{x_1, x_2, \dots, x_n\}$, intuitionistic fuzzy matrix $A = (a_{ij})_{n \times n} = (\mu_{ij}, v_{ij})$, $i, j \in I_n$. We construct completely multiplicative consistency intuitionistic fuzzy matrix $\bar{A} = (\bar{a}_{ij})_{n \times n}$, where

$$\bar{\mu}_{ij} = \begin{cases} \frac{j-i-1 \sqrt{\prod_{t=i+1}^{j-1} \mu_{it} \mu_{tj}}}{j-i-1 \sqrt{\prod_{t=i+1}^{j-1} \mu_{it} \mu_{tj}} + j-i-1 \sqrt{\prod_{t=t+1}^{j-1} (1-\mu_{it})(1-\mu_{tj})}}, j > i+1 \\ \mu_{ij}, j = i+1 \\ 0.5, j = i \\ \bar{\nu}_{ji}, j < i \end{cases},$$

$$\bar{\nu}_{ij} = \begin{cases} \frac{j-i-1 \sqrt{\prod_{t=i+1}^{j-1} \nu_{it} \nu_{tj}}}{j-i-1 \sqrt{\prod_{t=i+1}^{j-1} \nu_{it} \nu_{tj}} + j-i-1 \sqrt{\prod_{t=t+1}^{j-1} (1-\nu_{it})(1-\nu_{tj})}}, j > i+1 \\ \nu_{ij}, j = i+1 \\ 0.5, j = i \\ \bar{\mu}_{ji}, j < i \end{cases}.$$

By analyzing definition 18, we can find that different permutation σ will construct different completely multiplicative consistency intuitionistic fuzzy matrix, which is illustrated by the following example:

Example 1.[14] Consider the intuitionistic fuzzy matrix $R_{\sigma(123)}$ as follow:

$$R_{\sigma(123)} = \begin{pmatrix} & A_1 & A_2 & A_3 \\ A_1 & (0.5, 0.5) & (0.7, 0.2) & (0.65, 0.2) \\ A_2 & (0.2, 0.7) & (0.5, 0.5) & (0.55, 0.25) \\ A_3 & (0.2, 0.65) & (0.25, 0.55) & (0.5, 0.5) \end{pmatrix}, \quad (24)$$

we have $\bar{R}_{\sigma(123)}$ by definition 18

$$\bar{R}_{\sigma(123)} = \begin{pmatrix} & A_1 & A_2 & A_3 \\ A_1 & (0.5, 0.5) & (0.7, 0.2) & (0.7404, 0.0769) \\ A_2 & (0.2, 0.7) & (0.5, 0.5) & (0.55, 0.25) \\ A_3 & (0.0769, 0.7404) & (0.25, 0.55) & (0.5, 0.5) \end{pmatrix}, \quad (25)$$

Switching the positions of x_2 and x_3 , we have

$$R_{\sigma(132)} = \begin{pmatrix} & A_1 & A_3 & A_2 \\ A_1 & (0.5, 0.5) & (0.65, 0.2) & (0.7, 0.2) \\ A_3 & (0.2, 0.65) & (0.5, 0.5) & (0.25, 0.55) \\ A_2 & (0.2, 0.7) & (0.55, 0.25) & (0.5, 0.5) \end{pmatrix}, \quad (26)$$

You can also get $\bar{R}_{\sigma(132)}$

$$\bar{R}_{\sigma(132)} = \begin{pmatrix} & A_1 & A_3 & A_2 \\ A_1 & (0.5, 0.5) & (0.65, 0.2) & (0.3824, 0.2340) \\ A_2 & (0.2, 0.65) & (0.5, 0.5) & (0.25, 0.55) \\ A_3 & (0.2340, 0.3824) & (0.55, 0.25) & (0.5, 0.5) \end{pmatrix}, \quad (27)$$

Through observation, we can find that different permutations σ can construct different completely multiplicative consistency intuitionistic fuzzy matrices. In what follow, approximate consistency of intuitionistic fuzzy matrix is put forward, and present a new method to obtain the interval weight vector of intuitionistic fuzzy matrix.

3. approximation-consistency of intuitionistic fuzzy matrixs

It is assumed that $X = \{x_1, x_2, \dots, x_n\}$ is a set of alternatives. The interval preference degree r_{ij} is produced by comparing the alternative x_i to the alternative x_j . τ is a one by one mapping from $I = \{1, 2, \dots, n\}$ to $I = \{1, 2, \dots, n\}$ defined as

$$\tau : k \rightarrow i_k, \forall k, i_k \in \{1, 2, \dots, n\}.$$

285 The interval number $r_{ij}^\sigma = [r_{\sigma(i)\sigma(j)}^-, r_{\sigma(i)\sigma(j)}^+]$ ($\forall i, j \in I$) is the preference intensity of the alternatives $x_{\sigma(i)}$ to $x_{\sigma(j)}$. Therefore, a permutation σ corresponds to an interval additive reciprocal matrix R^σ . The definition of the interval additive reciprocal matrix with a permutation σ can be given as follows:

Definition 19. [18] Let σ is a permutation of $\{1, 2, \dots, n\}$. R^σ is an interval additive reciprocal matrix with a permutation σ :

$$R^\sigma = (r_{ij}^\sigma)_{n \times n} = \begin{bmatrix} x_{\sigma(1)} & x_{\sigma(2)} & \cdots & x_{\sigma(n)} \\ x_{\sigma(1)} & [0.5, 0.5] & [r_{\sigma(1)\sigma(2)}^-, r_{\sigma(1)\sigma(2)}^+] & \cdots & [r_{\sigma(1)\sigma(n)}^-, r_{\sigma(1)\sigma(n)}^+] \\ x_{\sigma(2)} & [r_{\sigma(2)\sigma(1)}^-, r_{\sigma(2)\sigma(1)}^+] & [0.5, 0.5] & \cdots & [r_{\sigma(2)\sigma(n)}^-, r_{\sigma(2)\sigma(n)}^+] \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x_{\sigma(n)} & [r_{\sigma(n)\sigma(1)}^-, r_{\sigma(n)\sigma(1)}^+] & [r_{\sigma(n)\sigma(2)}^-, r_{\sigma(n)\sigma(2)}^+] & \cdots & [0.5, 0.5] \end{bmatrix}.$$

Liu[18] defined the additive approximation-consistency of interval additive reciprocal matrix $R^\sigma = (r_{ij}^\sigma)_{n \times n}$. First of all, Liu[18] defined the additive reciprocal matrix $P^\sigma = (p_{ij}^\sigma)_{n \times n}$ and $Q^\sigma = (q_{ij}^\sigma)_{n \times n}$, where

$$p_{ij}^\sigma = \begin{cases} r_{\sigma(i)\sigma(j)}^-, & i < j, \\ 0.5, & i = j, \\ r_{\sigma(i)\sigma(j)}^+, & i > j, \end{cases} \quad , \quad q_{ij}^\sigma = \begin{cases} r_{\sigma(i)\sigma(j)}^+, & i < j, \\ 0.5, & i = j, \\ r_{\sigma(i)\sigma(j)}^-, & i > j. \end{cases} \quad (28)$$

290 It is noted that the definition of approximation-consistency of interval additive reciprocal matrices have been proposed in [18]. That is:

Definition 20. [18] It is supposed that R is an interval additive preference relation. R^σ is a matrix derived from R by applying a permutation σ . We call R^σ is additive approximation-consistency, if there is a permutation σ such that P^σ and Q^σ determined by (28) are all additively consistent,

Definition 21. [18] It is supposed that R is an interval additive preference relation. R^σ is a matrix derived from R by applying a permutation σ . We call R^σ is multiplicative approximation-consistency, if there is a permutation σ such that P^σ and Q^σ determined by (28) are all multiplicative consistency.

300 In what follows, through the transformation formula of between intuitionistic fuzzy preference relation and interval additive preference relation, we propose the concept of approximation-consistency of intuitionistic fuzzy matrix with permutations σ :

Definition 22.

305 $A = (a_{ij})_{n \times n}$ is an intuitionistic fuzzy matrix, interval additive reciprocal matrix $R = (r_{ij})_{n \times n}$ is constructed by (15). Assume that σ is a permutation of $\{1, 2, \dots, n\}$. R^σ is an interval additive reciprocal matrix with a permutation σ as follow:

$$R^\sigma = (r_{ij}^\sigma)_{n \times n} = \begin{bmatrix} x_{\sigma(1)} & x_{\sigma(2)} & \cdots & x_{\sigma(n)} \\ x_{\sigma(1)} & [0.5, 0.5] & [r_{\sigma(1)\sigma(2)}^-, r_{\sigma(1)\sigma(2)}^+] & \cdots & [r_{\sigma(1)\sigma(n)}^-, r_{\sigma(1)\sigma(n)}^+] \\ x_{\sigma(2)} & [r_{\sigma(2)\sigma(1)}^-, r_{\sigma(2)\sigma(1)}^+] & [0.5, 0.5] & \cdots & [r_{\sigma(2)\sigma(n)}^-, r_{\sigma(2)\sigma(n)}^+] \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x_{\sigma(n)} & [r_{\sigma(n)\sigma(1)}^-, r_{\sigma(n)\sigma(1)}^+] & [r_{\sigma(n)\sigma(2)}^-, r_{\sigma(n)\sigma(2)}^+] & \cdots & [0.5, 0.5] \end{bmatrix}.$$

where

$$r_{\sigma(ij)}^- = \mu_{\sigma(ij)}, \quad r_{\sigma(ij)}^+ = 1 - v_{\sigma(ij)}, \quad r_{\sigma(ij)}^- + r_{\sigma(ji)}^+ = r_{\sigma(ij)}^+ + r_{\sigma(ji)}^- = 1. \quad (29)$$

Letting $C_\sigma = (c_{ij}^\sigma)_{n \times n}$ and $D_\sigma = (d_{ij}^\sigma)_{n \times n}$, where

$$c_{ij}^\sigma = \begin{cases} r_{\sigma(i)\sigma(j)}^+, & i < j \\ 0.5, & i = j \\ r_{\sigma(i)\sigma(j)}^-, & i > j \end{cases}, \quad d_{ij}^\sigma = \begin{cases} r_{\sigma(i)\sigma(j)}^-, & i < j \\ 0.5, & i = j \\ r_{\sigma(i)\sigma(j)}^+, & i > j \end{cases}. \quad (30)$$

310 **Definition 23.** $A = (a_{ij})_{n \times n}$ is an intuitionistic fuzzy matrix, $R = (r_{ij})_{n \times n}$ is constructed by (15). Assume that σ is a permutation of $\{1, 2, \dots, n\}$. R^σ is an interval additive reciprocal matrix with a permutation σ , it is said that $A = (a_{ij})_{n \times n}$ is multiplicative approximation-consistency, if $C_\sigma = (c_{ij}^\sigma)_{n \times n}$ and $D_\sigma = (d_{ij}^\sigma)_{n \times n}$ are all multiplicative consistency through (19).

315 **Definition 24.**

$A = (a_{ij})_{n \times n}$ is an intuitionistic fuzzy matrix, $R = (r_{ij})_{n \times n}$ is constructed by (15). Assume that σ is a permutation of $\{1, 2, \dots, n\}$. R^σ is an interval additive reciprocal matrix with a permutation σ , it is said that $A = (a_{ij})_{n \times n}$ is additive approximation-consistency, if $C_\sigma = (c_{ij}^\sigma)_{n \times n}$ and $D_\sigma = (d_{ij}^\sigma)_{n \times n}$ are all additive consistent through (19).
320

In what follow, we propose acceptable approximate consistency of intuitionistic fuzzy matrix:

Definition 25. We call intuitionistic fuzzy matrix $A = (a_{ij})_{n \times n}$ is a acceptable approximate multiplicative consistency. If there is a permutation σ of $(1, 2, \dots, n)$, the distance between the intuitionistic fuzzy matrix $A = (a_{ij})_{n \times n}$ and the completely multiplicative consistency intuitionistic fuzzy matrix $\bar{A} = (\bar{a}_{ij})_{n \times n}$ satisfy that
325

$$d(A, \bar{A}) < \tau, \quad (31)$$

the τ is consistency check index.

According to the distance formula of Hamming[19], $d(A, \bar{A})$ can be expressed by the Hamming distance formula as follows:

$$d_h(A, \bar{A}) = \frac{1}{2(n-1)(n-2)} \sum_{i=1}^n \sum_{j=1}^n (|\mu_{ij} - \bar{\mu}_{ij}| + |v_{ij} - \bar{v}_{ij}| + |\pi_{ij} - \bar{\pi}_{ij}|).$$

As shown in the above analysis, we have defined the approximation- consistency of intuitionistic fuzzy matrix. Next, for the sake of applications in decision making problems, we introduce the weight vector of intuitionistic fuzzy number.
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4. Methods of obtaining interval weights

There have been a lot of researches on the weight of intuitionistic fuzzy number, mainly including precise weight, interval value weight and intuitionistic fuzzy weight. In what follow, we define the interval weight vector of alternatives by considering the permutations of alternatives:
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It is supposed that σ is any permutation of $\{1, 2, \dots, n\}$ with $\sigma(i) = k$ ($i, k \in \{1, 2, \dots, n\}$). where $\omega_k(C^\sigma)$ and $\omega_k(D^\sigma)$ stand for the weights of the alternative k derived from C^σ and D^σ , respectively. We define the interval weight vector of the alternative k as

$$\omega_k^\sigma = [\min\{\omega_k(C^\sigma), \omega_k(D^\sigma)\}, \max\{\omega_k(C^\sigma), \omega_k(D^\sigma)\}]. \quad (32)$$

Moreover, by considering the randomness of the permutation σ , the expected value of ω_k^σ is calculated as the final weight of the alternative k . That is, one has

$$\omega_k = \frac{1}{n!} \sum_{\sigma} \omega_k^\sigma.$$

Then the final wight vector is given as

$$\omega = (\omega_1, \omega_2, \dots, \omega_n). \quad (33)$$

In the next section, we illustrate the new method by carrying out numerical examples and offer some comparisons.

5. Algorithm and illustrative examples

340 Based on the proposed definition and methods, it is requisite to consider their applications in decision making problems. Now we offer the algorithm to solve the decision making problem with intuitionistic fuzzy matrix:

Step 1: Consider a decision-making problem with a set of alternatives $X = \{x_1, x_2, \dots, x_n\}$. The decision maker provides an intuitionistic fuzzy matrix A
 345 , $R_\sigma = (r_{\sigma(i)\sigma(j)})_{n \times n}$ with a permutation σ of $\{1, 2, \dots, n\}$ is constructed by (29).

Step 2: A series of R^{σ_i} are obtained with $i = 1, 2, \dots, n!/2$. Then we achieve additive reciprocal matrices C^{σ_i} and D^{σ_i} by using (30).

Step 3: Additive approximation-consistency of R^{σ_i} is checked by testing the
 350 additive consistency of C^{σ_i} and D^{σ_i} . If there is a permutation σ such that C^σ and D^σ determined by using (30) are all additive additive consistency, one can

go directly to the next step. Otherwise, any one of them can be adjusted to that with additive consistency by the consistency improving methods [20]

Step 4: Based on (33), the interval weights of alternatives are given.

355 *Step 5:* By applying the possibility degree formula in [19], the possibility degree matrix is given.

Step 6: According to the possibility degree matrix, a method such as that in [21] is used to generate the ranking of alternatives.

Step 7: End.

360 **Example 1:** [8][11][12][22] Suppose that a decision maker provides his preference information over a collection of alternatives x_1, x_2, x_3 with the following IFPR:

$$A = \begin{pmatrix} (0.5, 0.5) & (0.2, 0.6) & (0.6, 0.4) \\ (0.6, 0.2) & (0.5, 0.5) & (0.7, 0.1) \\ (0.4, 0.6) & (0.1, 0.7) & (0.5, 0.5) \end{pmatrix}.$$

We have the following result from (1):

$$R = \begin{bmatrix} [0.5, 0.5] & [0.2, 0.4] & [0.6, 0.6] \\ [0.6, 0.8] & [0.5, 0.5] & [0.7, 0.9] \\ [0.4, 0.4] & [0.1, 0.3] & [0.5, 0.5] \end{bmatrix}.$$

First, we check the additive approximation-consistency of the interval additive reciprocal matrix. Applying any permutation σ of alternatives to R , it is found that C^σ and D^σ are not additively consistent, meaning that R has not additive approximation-consistency.

$$C^\sigma = \begin{bmatrix} 0.5 & 0.4 & 0.6 \\ 0.6 & 0.5 & 0.9 \\ 0.4 & 0.1 & 0.5 \end{bmatrix},$$

$$D^\sigma = \begin{bmatrix} 0.5 & 0.2 & 0.6 \\ 0.8 & 0.5 & 0.7 \\ 0.4 & 0.3 & 0.5 \end{bmatrix}.$$

Second, we choose randomly a permutation of alternatives such as $\sigma = (1, 2, 3)$ to adjust R to that with additively consistent. Here the improving method in [19] is applied to adjust C^σ and D^σ to C'_σ and D'_σ

$$C'_\sigma = \begin{bmatrix} 0.5 & 0.3333 & 0.6667 \\ 0.6667 & 0.5 & 0.8333 \\ 0.333 & 0.1667 & 0.5 \end{bmatrix},$$

$$D'_\sigma = \begin{bmatrix} 0.5 & 0.2667 & 0.5333 \\ 0.7333 & 0.5 & 0.7667 \\ 0.4667 & 0.2333 & 0.5 \end{bmatrix}.$$

Third, according to (33), the interval weight vector can be computed as $w1 = [0.4889, 0.5722]$, $w2 = [0.4332, 0.5056]$, $w3 = [0.4555, 0.5444]$.

Finally, the matrix of possibility degrees is determined by using the method in [21] as

$$P = \begin{bmatrix} 0.5 & 0.9769 & 0.7920 \\ 0.0231 & 0.5 & 0.195 \\ 0.2080 & 0.8050 & 0.5 \end{bmatrix},$$

the ranking of the alternative is given as $x_1 > x_2 > x_3$. The obtained result is in agreement with that in [11]. As compared to those in [11], here the permutations of alternatives are considered and the weights are given by using the expected value.

Example 2: [6] Suppose that a decision maker provides his preference information over a collection of alternatives x_1, x_2, x_3 with the following IFPR:

$$A = \begin{pmatrix} (0.5, 0.5) & (0.8, 0.1) & (0.6, 0.4) \\ (0.1, 0.8) & (0.5, 0.5) & (0.7, 0.1) \\ (0.4, 0.6) & (0.1, 0.7) & (0.5, 0.5) \end{pmatrix}.$$

We have the following result from (1):

$$R = \begin{bmatrix} [0.5, 0.5] & [0.8, 0.9] & [0.6, 0.6] \\ [0.1, 0.2] & [0.5, 0.5] & [0.7, 0.9] \\ [0.4, 0.4] & [0.1, 0.3] & [0.5, 0.5] \end{bmatrix}.$$

First, we check the additive approximation-consistency of the interval additive reciprocal matrix. Applying any permutation σ of alternatives to R , it is found that C^σ and D^σ are not additively consistent, meaning that R has not additive approximation-consistency.

$$C^\sigma = \begin{bmatrix} 0.5 & 0.9 & 0.6 \\ 0.1 & 0.5 & 0.9 \\ 0.4 & 0.1 & 0.5 \end{bmatrix},$$

$$D^\sigma = \begin{bmatrix} 0.5 & 0.8 & 0.6 \\ 0.2 & 0.5 & 0.7 \\ 0.4 & 0.3 & 0.5 \end{bmatrix}.$$

Second, we choose randomly a permutation of alternatives such as $\sigma = (1, 2, 3)$ to adjust R to that with additively consistent. Here the improving method in [19] is applied to adjust C^σ and D^σ to C'_σ and D'_σ

$$C'_\sigma = \begin{bmatrix} 0.5 & 0.6667 & 0.8333 \\ 0.3333 & 0.5 & 0.6667 \\ 0.1667 & 0.3333 & 0.5 \end{bmatrix},$$

$$D'_\sigma = \begin{bmatrix} 0.5 & 0.6667 & 0.7333 \\ 0.3333 & 0.5 & 0.5667 \\ 0.2667 & 0.4333 & 0.5 \end{bmatrix}.$$

Third, according to (33), the interval weight vector can be computed as $w1 = [0.4721, 0.5333]$, $w2 = [0.4721, 0.5278]$, $w3 = [0.4666, 0.5333]$.

Finally, the matrix of possibility degrees is determined by using the method in [21] as

$$P = \begin{bmatrix} 0.5 & 0.5449 & 0.5412 \\ 0.4506 & 0.5 & 0.4959 \\ 0.4588 & 0.5041 & 0.5 \end{bmatrix},$$

the ranking of the alternative is given as $x_1 > x_3 > x_2$. The best plan is the same with that in [6].

6. Conclusions

In real decision-making problems, in order to solve the complexity and uncertainty of problems, scholars put forward intuitionistic fuzzy number, which solved the practical problems well. In this paper, we review the consistency definition of intuitionistic fuzzy matrix and point out the shortcomings of the definition. In order to solve the shortage of traditional definition, we put forward the concept of approximate consistency of intuitionistic fuzzy matrix. Through the transformation formula of intuitionistic fuzzy preference relation and interval additive preference relation, we transform the intuitionistic fuzzy matrix into the interval additive reciprocal matrix, and put forward the approximate consistency of intuitionistic fuzzy matrix considering the arrangement of schemes, and give a new method to obtain the interval weight vector of intuitionistic fuzzy matrix. A new algorithm for solving decision-making problems with intuitionistic fuzzy matrix has been given. Two examples have been carried out to illustrate the proposed definition and algorithm and some comparisons have been made.

In the future, the developed definition and methods will be extended to address group decision making problems with intuitionistic fuzzy matrix.

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