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Grey Language Hesitant Fuzzy Group Decision Making Method Based on Kernel and Grey Scale

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Abstract: Based on grey language multi-attribute group decision making, a kernel and grey scale scoring function is put forward according to the definition of grey language and the meaning of the kernel and grey scale. The function introduces grey scale into the decision-making method to avoid information distortion. This method is applied to the grey language hesitant fuzzy group decision making, and the grey correlation degree is used to sort the schemes. The effectiveness and practicability of the decision-making method are further verified by the industry chain sustainable development ability evaluation example of a circular economy. Moreover, its simplicity and feasibility are verified by comparing it with the traditional grey language decision-making method and the grey language hesitant fuzzy weighted arithmetic averaging (*GLHWAA*) operator integration method after determining the index weight based on the grey correlation.

Keywords: kernel and grey scale; grey language; hesitant fuzzy; grey correlation

1. Introduction

The deepening interdependence and mutual influence of the world economy and the application of the new generation of Internet information technology have significantly influenced the decisions of the state and enterprises. The complexity, uncertainty and collectivization of decision-making information in politics, the economy and military have become the new normal of decision making.

A complex decision-making environment and the limited cognitive ability of decision makers require overcoming the constraints of traditional sets. In 1965, the American cybernetics expert Zadeh [1] first proposed the concept of the fuzzy set, which expresses the idea of uncertainty through membership degree and has been widely used in various fields, such as management decisions, medical diagnosis and engineering control. The Bulgarian scholar Atanassov [2] argued that the membership degree of traditional fuzzy sets can only be the numbers between 0 and 1, which cannot dig deeper into the fuzzy degree in actual decision making. Based on the above consideration, Atanassov studied three dimensions of membership degree, non-membership degree and hesitation degree, and gave the relational expression, which was further developed into an intuitive fuzzy set. Since Zadeh proposed fuzzy sets, many scholars have studied their extension forms, such as the intuitionistic fuzzy set (IFS) [3], the type 2 fuzzy set [4,5], the fuzzy multiple set, and [6,7] the hesitant fuzzy set. This includes group decision making based on information integration and the measurement of hesitant fuzzy sets. Based on Archimedes T-module and S-module, Xia Meimei gave the rules of hesitant fuzzy operation and the information integration operator of hesitant fuzzy decision; defined the distance, similarity, correlation, entropy and cross-entropy of hesitant fuzzy sets; and discussed their properties. Moreover, their correlation was studied in order to apply this to group decision-making problems [8]. These studies have laid a foundation for the study of hesitant fuzzy group decision making based on the kernel and grey level.

Increasingly complex decision information requires experts to make group decisions in more cases. Experts need to face the complex environment with both complete and incomplete information, both definite information and uncertain information. The complex group decision-making environment has also attracted the attention of domestic scholars. In 1982, Deng Julong, a Chinese scholar, published the first paper about the grey system entitled “The Control Problems of Grey Systems in Systems and Control Letters” [9]. In the same year, Deng Julong published the first Chinese grey system paper “Grey Control System” in the Journal of the Huazhong Institute of Technology [10]. These two innovative classical papers have aroused wide attention and successfully initiated an emerging trans-disciplinary science—grey system theory [11].

The grey multi-attribute decision-making method was studied by [12–15]. Youchen et al. studied the inverse order problem of the grey decision model [16]. Zhang Na et al. established the multi-stage grey situation group decision model based on the Orness measure constraint [17]. Liu Sifeng et al. explored the two-stage grey comprehensive measurement decision model and the improvement of the triangular whitening weight function [18]. Luo Dang and Liu Sifeng studied the grey relational decision method of the grey relational decision method and an incomplete information system [19,20]. Dang Yaoguo et al. studied the grey relational decision model based on a dynamic multi-index [21]. Jiang Shiquan et al. put forward a grey relational decision model based on area [22]. Xie Ming et al. studied the grey relational analysis method of multi-attribute group decision making based on the analytic hierarchy process (AHP) and its application [23]. Chen Xiaoxin and Liu Sifeng put forward a grey relational decision method with partial weight information and a preference to scheme [24]. The rapid development of the grey relational group decision-making method provides the research foundation for the fuzzy grey relational decision making in this paper.

The differences of knowledge level, judgment experience and preference of decision experts will inevitably lead to the deviation of information interpretation, which further causes uncertain decision-making problems accompanied by fuzzy and grey characteristics, namely, the grey fuzzy multi-attribute group decision-making problem. The intuitionistic fuzzy decision making method was studied by [25–29] on the basis of grey correlation combined with evidence reasoning and the expert system uncertainty factor. Hu Lifang et al. organically combined grey with fuzzy, and put forward the grey fuzzy multi-attribute decision making method under the frame of the closed world [30]. Li Peng et al. put forward the combination method of grey relation and Dempster-Shafer evidence theory, which was applied to multiple-criteria decision making (MCDM) based on the intuitionistic fuzzy set (IFS) [31].

Some scholars from Nanjing University of Aeronautics and Astronautics combined grey system with fuzzy theory, and put forward relevant decision methods. For example, Li Peng, Liu Sifeng and Zhu Jianjun studied the combination of GR, DS and MYCIN, which was applied to IFS decision-making [32,33]. Liu Yong and Liu Sifeng studied the dynamic multi-attribute grey relational decision making method based on interval intuitionistic fuzzy [26]. In [34], the grey correlation method and non-linear programming model were used to obtain the optimal weight and study the relevant decision methods. The study of grey system and fuzzy combination promotes the fusion and promotion of the uncertainty decision method.

Research in the field of grey language multi-attribute decision-making is relatively scarce. Wu Jianwen studied the grey language multi-criteria decision method and its application [35]. Wang Jiang et al. put forward a multi-criteria decision method based on interval grey uncertain language [36]. Liu Peide et al. proposed a multi-attribute group decision-making method based on a geometric weighted integration operator of interval grey language variables [37]. Wang Hehua studied the information aggregation model of language evaluation based on grey decision and its application [38]. Li Qingsheng et al. [39–41] put forward the grey hesitant fuzzy set and extended it to the grey language form, namely, the grey language hesitant fuzzy set, which can conveniently use the special decision method of a grey system and effectively solve the problem of hesitant fuzzy decision making. How to take the advantages of a grey system to make grey language hesitant fuzzy decision

expand the application scope of a grey system is significant for enriching and perfecting grey system theory and the organic combination of uncertain decision making. This paper proposes a kernel and grey scale-scoring function of the grey language hesitant fuzzy set which uses the grey correlation degree to make decisions.

2. Preliminaries

2.1. Grey Language Set

The concepts of language value and grey scale are generally combined in some practical decision-making processes, so that the decision information is more in line with the decision habit, which is more like the description of kernel and grey scale based on a grey system. Wu Jianwen and Wang Jianjiang put forward the grey language set by studying the multi-criteria decision and grey system of language, and discussed the problem of multi-attribute decision-making based on grey language. For example, a college counselor gives the evaluation results of “excellent”, “good”, “fair”, “qualified” and “failing”. If a college student is rated as “good”, the membership degree of the comprehensive evaluation value of “good” is 1. However, the language evaluation value of membership degree of 1 is often impractical, as it requires counselors to grasp comprehensive information about the student, including learning, interpersonal relations, comprehensive ability in the family, society, classroom and dormitory, etc. Obviously, this is impossible in reality. The decision maker’s “limited rationality” principle proposed by Simon is closer to reality. If the counselor can give the membership degree and incomplete degree of information (grey scale), it will be very helpful for the scientific evaluation of college students. Therefore, evaluation values in the following form can be obtained: $A = \{x, H_i, \mu_i, v_i | x \in X\} = \{x, \text{good}, 0.8, 0.4\}$, which represents the membership degree to which the college student is evaluated as “good” at 0.8, and the information uncertainty at 0.4.

2.1.1. Definition of Grey Language Set

Definition 1. Set the discourse domain X , and language phase set $H = \{H_0, H_1, H_2, \dots, H_{T-1}, H_T\}$. For $\forall x \in X$, there is a corresponding language value $H_i (H_i \in H)$. If the membership degree of x to H_i is the grey number in $[0, 1]$, its point grey scale is $v(x)$, $A = \{x, H_i, \mu_i, v_i | x \in X\}$ is a grey language set (GLS) on X . $\mu_{H_i}(x)$ is the membership degree, which can be understood as the close degree of x to H_i [38]. Different from intuitionistic fuzzy sets, $v(x)$ represents the extent to which the decision maker’s information is incomplete, namely, grey scale, instead of non-membership degree. Although there is no necessary correlation between grey scale $v(x)$ and membership degree $\mu_{H_i}(x)$, the smaller the value of $|\mu_i + v_i - 1|$, the better the reliability.

In general, the grey language set A is abbreviated as $A = \{H(x), \mu(x), v(x) | x \in X\}$. If $a = \{H(x), \mu(x), v(x)\}$ is GLS, any specific grey language element a in A can be abbreviated as GLE. If the values of membership degree and grey scale in a is $\mu(x) = 1, v(x) = 0$, a does not have incomplete information degree, and retreats into the general language number.

2.1.2. Algorithm of Grey Language Number

(1) Set $a_i = \{H_i, \mu_i, v_i\}$ and $a_j = \{H_j, \mu_j, v_j\}$ as arbitrary GLE, according to the following rules [35].

$$a_i \oplus a_j = (H_i \oplus H_j, \frac{i \cdot \mu_i + j \cdot \mu_j}{i+j}, \max(v_i, v_j)), \text{ when } H_i \text{ and } H_j \text{ are not equal to } H_0 \text{ simultaneously.}$$

$a_i \oplus a_j = (H_i \oplus H_j, \max(\mu_i, \mu_j), \max(v_i, v_j))$, when H_i and H_j are equal to H_0 . In the expression and calculation process of grey language “ \oplus ” in this paper, if there is no special explanation, it is operated as H_i and H_j are not equal to H_0 simultaneously.

$$(2) r a_i = \{r H_i, \mu_i, v_i\} \text{ [35].}$$

$$(3) a_i \otimes a_j = (H_i \otimes H_j, \mu_i \cdot \mu_j, v_i + v_j - v_i \cdot v_j) \text{ [35].}$$

$$(4) a_i^\lambda = (H_i^\lambda, \mu_i^\lambda, 1 - (1 - v_i)^\lambda) \text{ [35].}$$

2.1.3. Properties of Grey Language Number Algorithm

Assume that $a_i = \{H_i, \mu_i, v_i\}$, $a_j = \{H_j, \mu_j, v_j\}$ and $a_k = \{H_k, \mu_k, v_k\}$ are any three grey language numbers, and the arithmetic operation has the following properties:

- (1) $a_i \oplus a_j = a_j \oplus a_i$.
- (2) $(a_i \oplus a_j)a_k = a_i \oplus (a_j \oplus a_k)$.
- (3) $a_i \otimes a_j = a_j \otimes a_i$.
- (4) $(a_i \otimes a_j)a_k = a_i \otimes (a_j \otimes a_k)$.
- (5) $\lambda a_i \oplus \lambda a_j = \lambda(a_i \oplus a_j), \lambda > 0$.
- (6) $\lambda_1 a_i \oplus \lambda_2 a_i = (\lambda_1 + \lambda_2)a_i, \lambda_1 > 0, \lambda_2 > 0$.
- (7) $(a_i \otimes a_j)^\lambda = a_i^\lambda \otimes a_j^\lambda, \lambda \geq 0$.
- (8) $a_i^{\lambda_1} \otimes a_i^{\lambda_2} = a_i^{(\lambda_1 + \lambda_2)}, \lambda_1 \geq 0, \lambda_2 \geq 0$.

2.1.4. Size of Grey Language Numbers

Definition 2. Assume $a_i = \{H_i, \mu_i, v_i\}$ and $a_j = \{H_j, \mu_j, v_j\}$ are any two grey language numbers [38].

If $(1 - v_i) \cdot \mu_i \cdot IH_i < (1 - v_j) \cdot \mu_j \cdot IH_j$, a_i is smaller than a_j , denoted as $a_i < a_j$;

If $(1 - v_i) \cdot \mu_i \cdot IH_i > (1 - v_j) \cdot \mu_j \cdot IH_j$, a_i is greater than a_j , denoted as $a_i > a_j$;

If $(1 - v_i) \cdot \mu_i \cdot IH_i = (1 - v_j) \cdot \mu_j \cdot IH_j$, and $H_i = H_j, \mu_i = \mu_j, v_i = v_j$, a_i is strongly equal to a_j , denoted as $a_i = a_j$;

If $(1 - v_i) \cdot \mu_i \cdot IH_i = (1 - v_j) \cdot \mu_j \cdot IH_j$, and at least one of $H_i = H_j, \mu_i = \mu_j, v_i = v_j$, is tenable, a_i is weakly equal to a_j , denoted as $a_i \approx a_j$. Where, I is subscript operator.

2.2. Hesitant Fuzzy Set

When the attribute of a given scheme belongs to the membership degree of a certain set, it is not necessarily a random interval nor does it obey a certain distribution. The performance will be more prominent in group decision making. The membership degree given by different groups may be uncoordinated, and sometimes it is not necessary to reach consistent possible values. To this end, Torra [42,43] proposed the hesitant fuzzy set which can better meet the needs of group decision making.

2.2.1. Definitions Related to Hesitant Fuzzy Sets

Definition 3. Assume X is a given set, and hesitant fuzzy set is the projection from $X \rightarrow [0,1]$, abbreviated as $A = \{\langle x, h(x) \rangle | x \in X\}$. Where, $h(x)$ is the set of several hesitant fuzzy numbers (HFN) in $[0,1]$. HFN is the membership degree value of $x \in X$ belonging to set A . Obviously, there is $0 \leq \text{HFN} \leq 1$ [42,43].

The hesitant fuzzy element contains the possible hesitant fuzzy number in the set. The length of data in each hesitant fuzzy element is likely to vary. The following method is used to determine the size of the hesitant fuzzy element.

Definition 4. Any hesitation fuzzy element h , $s(h) = 1/l \sum_{\gamma \in h} \gamma$ is score function, where l is the number of elements in h . There are two hesitant fuzzy elements h_1, h_2 . If $s(h_1) > s(h_2)$, $h_1 > h_2$. If $s(h_1) = s(h_2)$, $h_1 \approx h_2$.

Torra believed that membership degree at both ends of hesitant fuzzy can be understood as an intuitionistic fuzzy number.

Definition 5. The envelope of any hesitant fuzzy number h can be converted to intuitionistic fuzzy number through the following formula: $A_{env}(h) = (h^-, (1 - h^+))$, where $h^- = \min \gamma, h^+ = \max \gamma \quad \gamma \in h$ [42,43].

Based on the connotation of hesitant fuzzy envelope relation [8] studied the envelop relation set operation.

- (1) $A_{env}(h^c) = (A_{env}(h))^c$.

- (2) $A_{env}(h_1 \cup h_2) = A_{env}(h_1) \cup A_{env}(h_2).$
- (3) $A_{env}(h_1 \cap h_2) = A_{env}(h_1) \cap A_{env}(h_2).$
- (4) $A_{env}(h^\lambda) = (A_{env}(h))^\lambda.$
- (5) $A_{env}(\lambda h) = \lambda A_{env}(h).$
- (6) $A_{env}(h_1 \cdot h_2) = A_{env}(h_1) \cdot A_{env}(h_2).$
- (7) $A_{env}(h_1 + h_2) = A_{env}(h_1) + A_{env}(h_2).$

2.2.2. Algorithms of Hesitant Fuzzy Set

Assume any three hesitant fuzzy numbers $h, h_1, h_2,$ and Torra [42,43] studied their algorithms:

- (1) $h^c = \cup_{\gamma \in h} \{1 - \gamma\}.$
- (2) $h_1 \cup h_2 = \cup_{\gamma_1 \in h_1, \gamma_2 \in h_2} \max\{\gamma_1, \gamma_2\}.$
- (3) $h_1 \cap h_2 = \cup_{\gamma_1 \in h_1, \gamma_2 \in h_2} \min\{\gamma_1, \gamma_2\}.$

On the basis of studying the operation of Archimedes T-module and S-module [8] studied the algorithm of hesitant fuzzy operation. There are arbitrary hesitant fuzzy elements $h, h_1, h_2.$

- (1) $h^\lambda = \cup_{\gamma \in h} \{k^{-1}(\lambda k(\gamma))\}.$
- (2) $\lambda h = \cup_{\gamma \in h} \{l^{-1}(\lambda l(\gamma))\}.$
- (3) $h_1 \cdot h_2 = \cup_{\gamma_1 \in h_1, \gamma_2 \in h_2} \{k^{-1}(k(\gamma_1) + k(\gamma_2))\}.$
- (4) $h_1 + h_2 = \cup_{\gamma_1 \in h_1, \gamma_2 \in h_2} \{l^{-1}(l(\gamma_1) + l(\gamma_2))\}.$

where $l(t) = k(1 - t),$ and $k : [0, 1] \rightarrow [0, +\infty)$ is a strictly decreasing function.

The hesitant fuzzy operation rule can be defined as below:

Definition 6. There are three hesitant fuzzy numbers $h, h_1, h_2.$

- (1) $(h^c)^\lambda = (\lambda h)^c.$
- (2) $\lambda(h^c) = (h^\lambda)^c.$
- (3) $h_1^c \cdot h_2^c = (h_1 + h_2)^c.$
- (4) $h_1^c + h_2^c = (h_1 \cdot h_2)^c.$

where, $h^c = \cup_{\gamma \in h} \{1 - \gamma\}$ [8].

Definition 7. Assume $h_i (i = 1, 2, \dots, n)$ as a set of hesitant fuzzy numbers, $\omega = (\omega_1, \omega_2, \dots, \omega_n)^T$ satisfy, $\sum_1^n \omega_i = 1, \omega_i \geq 0, (i = 1, 2, \dots, n),$, $ATS - HFWA(h_1, h_2, \dots, h_n) = \sum_{i=1}^n \omega_i \cdot h_i.$

$ATS - HFWA$ is a hesitant fuzzy weighted arithmetic averaging operator based on Archimedes T-Module and S-Module.

$$ATS - HFWG(h_1, h_2, \dots, h_n) = \prod_{i=1}^n h_i^{\omega_i}.$$

$ATS - HFWG$ is a hesitant fuzzy weighted geometric averaging operator based on Archimedes T-Module and S-Module [8].

2.3. Kernel and Grey Scale

2.3.1. Definitions of Kernel and Grey Scale

The study of grey number operation in grey system theory has attracted attention for a long period of time, but no satisfactory result has been achieved.

Definition 8. Let the grey number $\otimes \in [\underline{a}, \bar{a}], \underline{a} < \bar{a}$ in the absence of grey number value distribution information [44]:

If \otimes is a continuous grey number, $\hat{\otimes} = \frac{1}{2}(\underline{a} + \bar{a})$ is the kernel of grey number $\otimes.$

If \otimes is discrete grey number, $a_i \in [a, \bar{a}]$, ($i = 1, 2, \dots, n$) is the possible value of grey number. Its average value is the kernel of grey number, namely, $\hat{\otimes} = \frac{1}{n} \sum_{i=1}^n a_i$ is the kernel of grey number \otimes .

Definition 9. If \otimes is a random grey number, and the mathematical expectation $E(\otimes)$ can be calculated, $E(\otimes)$ is the kernel of random \otimes and can expressed as $\hat{\otimes} = E(\otimes)$ [44].

Definition 10. Set the background or all of the grey number \otimes is Ω , $\mu(\otimes)$ as the measure of interval grey number \otimes taking the number field. $g^o(\otimes) = \mu(\otimes) / \mu(\Omega)$ is the grey level of grey number \otimes , which can be understood as the ratio of all possible membership degree to background, abbreviated as g^o [44].

The properties of $\otimes \in \Omega$ and the measure show that the grey scale satisfies the following properties: grey scale satisfies the criterion, namely, $0 \leq g^o(\otimes) \leq 1$.

- When $g^o = 0$, the uncertainty is 0, which is the white number for complete certainty;
- When $g^o = 1$, the uncertainty is 1, which is the black number for complete uncertainty;
- When $0 < g^o(\otimes) < 1$, it is an uncertain grey number.

The grey scale reflects the uncertainty degree of the things described by the grey number. The white number is a completely certain number, and the black number is completely unknown. When g^o is closer to 0, the uncertainty of the grey number is smaller. When g^o is closer 1, the uncertainty of the value range is greater.

Definition 11. $\hat{\otimes}$ is the kernel of grey number \otimes , and g^o is the grey scale. $\hat{\otimes}_{g^o}$ is the simplified form of grey numbers. $\hat{\otimes}_{g^o}$ includes all information of value, and has one-to-one correspondence [44].

On the one hand, $\hat{\otimes}_{g^o}$ can be calculated according to the definition of the kernel and grey scale. On the other hand, the range of \otimes can also be calculated according to $\hat{\otimes}_{g^o}$ [45] proposed the algorithm based on $\hat{\otimes}$ and g^o and the grey scale non-decreasing axiom.

2.3.2. Algorithms of Kernel and Grey Scale

Axiom 1. Grey scale non-decreasing axiom: grey scale is equal to the maximum grey level of all grey numbers involved in the operation, and it will not decrease [44].

According to Axiom 1, the simplified form $\tilde{\otimes}_k(g_k^o)$ based on the grey number can obtain the grey data algebraic algorithm [45].

- (1) $\tilde{\otimes}_k(g_k^o) + \tilde{\otimes}_m(g_m^o) = (\tilde{\otimes}_k + \tilde{\otimes}_m)(g_k^o \vee g_m^o)$.
- (2) $\tilde{\otimes}_k(g_k^o) - \tilde{\otimes}_m(g_m^o) = (\tilde{\otimes}_k - \tilde{\otimes}_m)(g_k^o \vee g_m^o)$.
- (3) $\tilde{\otimes}_k(g_k^o) \times \tilde{\otimes}_m(g_m^o) = (\tilde{\otimes}_k \times \tilde{\otimes}_m)(g_k^o \vee g_m^o)$.
- (4) $\tilde{\otimes}_k(g_k^o) \div \tilde{\otimes}_m(g_m^o) = (\tilde{\otimes}_k \div \tilde{\otimes}_m)(g_k^o \vee g_m^o) (\tilde{\otimes}_m \neq 0)$.
- (5) $m \times \tilde{\otimes}_k(g_k^o) = (m \times \tilde{\otimes}_k)(g_k^o)$.
- (6) $(\tilde{\otimes}_k(g_k^o))^r = (\tilde{\otimes}_k)^r(g_k^o)$.

Grey heterogeneous data may be included by the interval grey number, discrete grey number, real number or other grey information. All the different data structures and grey information characteristics belong to the same “grey” category, have a common attribute “kernel” and “grey scale” (real is a special grey number whose grey scale is “0”). So the algebraic algorithm between grey heterogeneous data can be studied through the “kernel” and “grey scale”.

Definition 12. Assume $F(\otimes)$ is a grey heterogeneous data set. For arbitrary \otimes_i and \otimes_j , $\otimes_i + \otimes_j$, $\otimes_i - \otimes_j$, $\otimes_i \times \otimes_j$ and $\otimes_i \div \otimes_j (\otimes_j \neq 0)$ belong to $F(\otimes)$, and is grey heterogeneous data field [46].

Assume $F(\otimes)$ as the grey heterogeneous data set. \otimes_i, \otimes_j and $\otimes_k \in F(\otimes)$. The kernel and grey scale are $\tilde{\otimes}_i(g_i^o), \tilde{\otimes}_j(g_j^o)$ and $\tilde{\otimes}_k(g_k^o)$. Grey heterogeneous data algebraic algorithms have the following properties [45]:

- (1) $\tilde{\otimes}_i(g_i^o) + \tilde{\otimes}_j(g_j^o) = \tilde{\otimes}_j(g_j^o) + \tilde{\otimes}_i(g_i^o)$.
- (2) $(\tilde{\otimes}_i(g_i^o) + \tilde{\otimes}_j(g_j^o)) + \tilde{\otimes}_k(g_k^o) = \tilde{\otimes}_i(g_i^o) + (\tilde{\otimes}_j(g_j^o) + \tilde{\otimes}_k(g_k^o))$.
- (3) $\exists \forall 0 \in F(\otimes)$, Make $\tilde{\otimes}_i(g_i^o) + 0 = \tilde{\otimes}_i(g_i^o)$.
- (4) For $\forall \otimes_s \in F(\otimes)$, $-\otimes_s \in F(\otimes)$ and make $-\otimes_s + \otimes_s = 0$.
- (5) $\tilde{\otimes}_i(g_i^o) \times (\tilde{\otimes}_j(g_j^o) \times \tilde{\otimes}_k(g_k^o)) = (\tilde{\otimes}_i(g_i^o) \times \tilde{\otimes}_j(g_j^o)) \times \tilde{\otimes}_k(g_k^o)$.
- (6) There is unit element $1 \in F(\otimes)$ which makes $1 \times \tilde{\otimes}_i(g_i^o) = \tilde{\otimes}_i(g_i^o) \times 1 = \tilde{\otimes}_i(g_i^o)$.
- (7) $(\tilde{\otimes}_i(g_i^o) + \tilde{\otimes}_j(g_j^o)) \times \tilde{\otimes}_k = \tilde{\otimes}_i(g_i^o) \times \tilde{\otimes}_k(g_k^o) + \tilde{\otimes}_j(g_j^o) \times \tilde{\otimes}_k(g_k^o)$.
- (8) $\tilde{\otimes}_i(g_i^o) \times (\tilde{\otimes}_j(g_j^o) + \tilde{\otimes}_k(g_k^o)) = \tilde{\otimes}_i(g_i^o) \times \tilde{\otimes}_j(g_j^o) + \tilde{\otimes}_i(g_i^o) \times \tilde{\otimes}_k(g_k^o)$.

3. Grey Language Hesitant Fuzzy Information Aggregation and Kernel and Grey Scale Scoring Function

3.1. Information Aggregation of Grey Language Hesitant Fuzzy Decision

Since the evaluation criteria are given in the form of grey language, the number l of decision experts is the length of the data, which is denoted as $r_{ij} = \{(H_{\sigma 1}, \mu_1, v_1), (H_{\sigma 2}, \mu_2, v_2), \dots, (H_{\sigma l}, \mu_l, v_l)\}$, where σk is the IH value of the k th item. The data are aggregated by the algorithm of the grey language fuzzy number to obtain z_{ij} .

$$z_{ij} = (H_{\frac{1}{l}(\sum_{k=1}^l \sigma k)}, \frac{\sigma 1 \cdot \mu_1 + \sigma 2 \cdot \mu_2 + \dots + \sigma l \cdot \mu_l}{\sum_{k=1}^l \sigma k}, \max(v_1, v_2, \dots, v_l)) \tag{1}$$

Definition 13. Assume $A = \{a_k | k = 1, 2, 3, \dots, l\}$ as the grey language set, $a_k = \{H_k, \mu_k, v_k\}$ and $g(a_1, a_2, a_3, \dots, a_l) = \sum_{k=1}^l w_k a_k$.

Where, w_k is the weight vector corresponding to the grey language array a_k . $w_k \in [0, 1]$ and $\sum_{k=1}^l w_k = 1$, g is grey language weighted average operator, abbreviated as the *GLHFWAA* operator.

$$GLHFWAA(a_1, a_2, \dots, a_l) = \sum_{k=1}^l w_k a_k = \sum_{k=1}^l (w_k H_k, \mu_k, v_k) \tag{2}$$

If $w = (\frac{1}{l}, \frac{1}{l}, \dots, \frac{1}{l})$, *GLHFWAA* degenerates to grey language averaging operator *GLAA*,

$$GLHFAA(a_1, a_2, \dots, a_l) = \sum_{k=1}^l w_k a_k = \sum_{k=1}^l (\frac{1}{l} H_k, \mu_k, v_k)$$

3.2. Grey Language Hesitant Fuzzy Score Function

Definition 14. Assume $a_i = \{H_i, \mu_i, v_i\}$ as any grey language number, and $S(a_k)$ is the score function of a_k .

$$S(a_k) = (1 - v_k) \cdot \mu_k \cdot IH_k \tag{3}$$

Assume $glh \in GLH$, $GLH = \{(H_1, \mu_1, v_1), (H_2, \mu_2, v_2), \dots, (H_l, \mu_l, v_l)\}$ are grey language decision data given by l experts.

$$S(GLH) = \frac{1}{l} \sum_{k=1}^l S(a_k) \quad k = 1, 2 \dots l \tag{4}$$

$S(GLH)$ is the score function of GLH .

Example. Let $GLH = \{(3, 0.6, 0.2), (4, 0.6, 0.3), \dots, (4, 0.7, 0.2)\}$, $S(GLH) = \frac{1}{l} \sum_{k=1}^l S(a_k) = \frac{1}{3} \times [(1 - 0.2) \times 0.6 \times 3 + (1 - 0.3) \times 0.6 \times 4 + (1 - 0.2) \times 0.7 \times 4] = 1.7867$.

The above definition transforms uncertain grey language information into a score function and becomes a definite value, which is theoretically inadequate. In this paper, it is considered that the score function should be a grey number [45] believed that the kernel $\hat{\otimes}$ of the grey number is an important representative of the value of the grey scale \otimes . Kernel is used in all kinds of synthesis operations, and has an important value. Also, the grey algebra operation properties of $\hat{\otimes}_{g^0}$ are discussed, and the relevant algorithm is proved. According to the gradation function of grey language and the definition of the kernel and grey scale, the grey language scoring function based on the kernel and grey scale is studied in this paper.

Definition 15. Assume $glh \in GLH$, $GLH = \{(H_1, \mu_1, v_1), (H_2, \mu_2, v_2), \dots, (H_l, \mu_l, v_l)\}$

$$S(\hat{\otimes}_{g^0}) = \left(\frac{1}{l} \sum_{k=1}^l \hat{\otimes}_k, \max(g_k^0)\right) = \left(\frac{1}{l} \sum_{k=1}^l \mu_i \cdot IH_k, \max(v_k)\right) \quad k = 1, 2 \dots l \tag{5}$$

$S(\hat{\otimes}_{g^0})$ is the score function of the grey language based on the kernel and grey scale.

4. Kernel and Grey Scale Group Decision Making Method Based on Grey Language Hesitant Fuzzy Information

4.1. Model Construction

Assume that there are m feasible schemes A_1, A_2, \dots, A_m and n evaluation criteria of a hesitant fuzzy group decision making problem based on grey language. $w = (w_1, w_2, \dots, w_n)^T$ is weight vector corresponding to the evaluation criteria, where w_i represents the weight of P_i . Assume the ordered language evaluation set $H = (H_0, H_1, \dots, H_r)$. According to the evaluation set, l experts evaluate various scheme indexes. In the grey language hesitant fuzzy set, \otimes belongs to the discrete grey numbers.

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

4.2. Decision-Making Steps

Step 1. In this step, the decision matrix $P_{m \times n} = (p_{ij})_{m \times n}$ is normalized as $R_{m \times n} = (r_{ij})_{m \times n}$.

It is necessary to eliminate the influence of different physical quantities before making a decision and to normalize different types of index values. Generally speaking, the cost index value should be converted to the benefit index. In special cases, the benefit index value also needs to be converted to the cost index value.

Cost index value (H_i, μ, v) is converted to benefit index value.

$$r_{ij} = (H_{T-i}, \mu, v) \tag{6}$$

Benefit index value (H_i, μ, v) is converted to cost index value:

$$r_{ij} = (H_{T-i}, \mu, v) \tag{7}$$

Step 2. Data supplement. Torra studied the hesitancy faced by experts in group decision making and proposed a hesitant fuzzy set suitable for group decision making.

Add the maximum membership scale to the hesitant fuzzy element when the decision maker has risk preference.

Add the minimum membership scale to the hesitant fuzzy element when the decision maker is risk-averse and pessimistic.

Step 3. Transform grey language data into a score function of kernel and grey scale through the Formula (4).

Step 4. Calculate grey correlation degree and grey scale.

Definition 16. $\hat{\otimes}$ is the kernel part of score function of grey language element based on kernel and grey scale. The grey correlation coefficient $\xi_i(j)$ and grey correlation degree $\gamma(j)$ between the empirical judgment value and the reference weight value of each evaluation index are obtained by the following formula:

Make $D = |\hat{\otimes}_i(j) - \hat{\otimes}_0(j)|$, and

$$\xi_i(j) = \frac{\min_i \min_j D + \rho \max_i \max_j D}{D + \rho \max_i \max_j D} \tag{8}$$

$$\gamma(i) = \frac{1}{n} \sum_{j=1}^n \xi_i(j), \quad 1 \leq i \leq m \tag{9}$$

where, ρ is the resolution ratio, $\rho \in (0, 1)$, generally, $\rho = 0.5$. The correlation degree of each sequence directly reflects the degree to which each evaluation is close to the optimal ideal scheme.

Definition 17. Assume $g^0(j_{\max})$ as the maximum grey scale of kernel value in j column and $g^0(i_{\max})$ as the maximum grey scale of kernel value in i row.

$$g^0(i) = \max(g^0(j_{\max}), g^0(i_{\max})) \quad 1 \leq j \leq n \tag{10}$$

$$\gamma_{g^0}(i) = (\gamma_{(g^0)}(1), \gamma_{(g^0)}(2), \dots, \gamma_{(g^0)}(m))^T \text{ Where, } 1 \leq i \leq m \tag{11}$$

Step 5. The schemes are sorted according to the grey correlation degree and grey scale. The following principles should be followed:

- (1) When $\gamma(i) > \gamma(m)$, $i \succ m$;
- (2) When $\gamma(i) = \gamma(m)$, and $g^0(i) > g^0(m)$, $i \prec m$;
- (3) When $\gamma(i) = \gamma(m)$, and $g^0(i) = g^0(m)$, $i \approx m$.

5. Example Analysis

As an economic development mode harmonious with the environment, the circular economy has become an important way to promote the comprehensive development ability of an industrial chain. Many industrial chains have carried out ecological transformations under the guidance of circular economy idea. The sustainable development of an industrial chain based on the circular economy is a complicated evaluation process, with uncertainty and multiple objectives. Evaluation of the sustainable development ability of an industrial chain based on the circular economy aims to study the optimization function of the circular economy on the design and reconstruction of the industrial chain, and to analyze the comprehensive development level and potential of the industrial chain from the perspective of the circular economy. The evaluation system of the sustainable development ability of an industrial chain based on the characteristics of the circular economy can provide scientific and effective means for the operation, management and improvement of the industrial chain. Therefore, it is necessary to establish an industrial chain sustainable development capacity evaluation index system covering

four aspects of economic development: (P_1), resource utilization (P_2), environmental protection (P_3) and green management (P_4), so as to scientifically and objectively reflect the sustainable development capacity of an industrial chain [46]. Four industries A_1, A_2, A_3, A_4 are assumed to be compared in the sustainable development of the circular economy industry chain in a certain area. Three experts evaluate and select $H = \left\{ \begin{array}{l} H_0 = \text{very poor}, H_1 = \text{poor}, H_2 = \text{relatively poor}, H_3 = \text{general} \\ H_4 = \text{good}, H_5 = \text{relatively good}, H_6 = \text{very good} \end{array} \right\}$ as the language evaluation term set. The evaluation value is represented by grey language value, forming the grey language hesitant fuzzy decision matrix in Table 1.

Table 1. The values of the grey language hesitant fuzzy decision matrix.

Industries	Attribute (P_1)	Attribute (P_2)
A_1	$\{(H_3, 0.6, 0.3), (H_4, 0.6, 0.4), (H_4, 0.7, 0.4)\}$	$\{(H_3, 0.7, 0.4), (H_3, 0.8, 0.4), (H_3, 0.9, 0.3)\}$
A_2	$\{(H_3, 0.7, 0.4), (H_3, 0.8, 0.4), (H_4, 0.9, 0.1)\}$	$\{(H_3, 0.9, 0.1), (H_4, 0.7, 0.4), (H_4, 0.8, 0.2)\}$
A_3	$\{(H_4, 0.7, 0.2), (H_5, 0.6, 0.2), (H_5, 0.8, 0.1)\}$	$\{(H_4, 0.7, 0.2), (H_5, 0.6, 0.2), (H_6, 0.9, 0.2)\}$
A_4	$\{(H_3, 0.6, 0.2), (H_4, 0.8, 0.4), (H_5, 0.9, 0.3)\}$	$\{(H_3, 0.7, 0.4), (H_4, 0.6, 0.4), (H_5, 0.7, 0.4)\}$
Industries	Attribute (P_3)	Attribute (P_4)
A_1	$\{(H_2, 0.8, 0.5), (H_3, 0.6, 0.4), (H_4, 0.7, 0.4)\}$	$\{(H_2, 0.8, 0.3), (H_2, 0.6, 0.4), (H_2, 0.7, 0.4)\}$
A_2	$\{(H_3, 0.6, 0.2), (H_3, 0.7, 0.4), (H_3, 0.8, 0.3)\}$	$\{(H_3, 0.7, 0.3), (H_3, 0.9, 0.4), (H_4, 0.8, 0.3)\}$
A_3	$\{(H_4, 0.8, 0.3), (H_4, 0.8, 0.2), (H_5, 0.7, 0.2)\}$	$\{(H_4, 0.9, 0.3), (H_5, 0.8, 0.2), (H_6, 0.8, 0.1)\}$
A_4	$\{(H_3, 0.8, 0.1), (H_3, 0.6, 0.2), (H_4, 0.8, 0.1)\}$	$\{(H_3, 0.9, 0.2), (H_4, 0.9, 0.3), (H_5, 0.8, 0.2)\}$

5.1. Method 1

Step 1. Normalize decision making matrix $P_{m \times n} = (p_{ij})_{m \times n}$ as $R_{m \times n} = (r_{ij})_{m \times n}$. All indexes in this example are benefit indexes, so they are omitted.

Step 2. Supplement data. Since there are three experts in this example, none of whom has abstained or given the same score, it is not necessary to supplement the data, and this step is omitted.

Step 3. Transform grey language data into the score function of the kernel and grey scale through Formula (5), and Table 2 is obtained.

Table 2. $\hat{\otimes}_{go}$ form of grey hesitant fuzzy decision matrix.

Industries	Attribute (P_1)	Attribute (P_2)
A_1	2.3333 _(0.40)	2.4000 _(0.40)
A_2	2.7000 _(0.40)	2.9000 _(0.40)
A_3	3.2667 _(0.20)	3.7333 _(0.20)
A_4	3.1667 _(0.40)	2.6667 _(0.40)
Industries	Attribute (P_3)	Attribute (P_4)
A_1	2.0667 _(0.50)	1.4000 _(0.40)
A_2	2.1000 _(0.30)	2.6667 _(0.40)
A_3	3.3000 _(0.30)	4.1333 _(0.30)
A_4	2.4667 _(0.20)	3.4333 _(0.30)

Step 4. Calculate grey correlation degree and grey scale.

$$\xi_i(j) = \begin{bmatrix} 0.5942 & 0.5062 & 0.5256 & 0.3333 \\ 0.7069 & 0.6212 & 0.5325 & 0.4824 \\ 1.0000 & 1.0000 & 1.0000 & 1.0000 \\ 0.9318 & 0.5616 & 0.6212 & 0.6613 \end{bmatrix}$$

$$\gamma(i) = (0.4898, 0.5857, 1.0000, 0.6940)^T$$

$$\gamma_{g^o}(i) = (0.4898_{(0.50)}, 0.5857_{(0.40)}, 1.0000_{(0.30)}, 0.6940_{(0.40)})^T$$

Step 5. Rank the schemes according to the grey correlation and grey scale of each industry. Obviously, $A_3 \succ A_4 \succ A_2 \succ A_1$.

5.2. Comparison Method 2

The following steps can be obtained according to the traditional decision making method in grey language.

Step 1. Normalize decision making matrix $P_{m \times n} = (p_{ij})_{m \times n}$ as $R_{m \times n} = (r_{ij})_{m \times n}$. All indexes in this example are benefit indexes, so they are omitted.

Step 2. Aggregate data r_{ij} through Formula (1) to obtain z_{ij} , as shown in Table 3.

Table 3. z_{ij} of grey language decision matrix.

Industries	Attribute (P_1)	Attribute (P_2)
A_1	$\{(H_{3.6667}, 0.6364, 0.4)\}$	$\{(H_3, 0.8, 0.4)\}$
A_2	$\{(H_{3.3333}, 0.89, 0.4)\}$	$\{(H_{3.6666}, 0.7909, 0.4)\}$
A_3	$\{(H_{4.6667}, 0.7, 0.2)\}$	$\{(H_5, 0.7615, 0.2)\}$
A_4	$\{(H_4, 0.7917, 0.4)\}$	$\{(H_4, 0.6667, 0.4)\}$
Industries	Attribute (P_3)	Attribute (P_4)
A_1	$\{(H_3, 0.6888, 0.5)\}$	$\{(H_2, 0.7, 0.4)\}$
A_2	$\{(H_3, 0.7, 0.4)\}$	$\{(H_{3.6667}, 0.8, 0.4)\}$
A_3	$\{(H_5, 0.8267, 0.3)\}$	$\{(H_2, 0.7, 0.3)\}$
A_4	$\{(H_{3.3333}, 0.74, 0.2)\}$	$\{(H_4, 0.8583, 0.3)\}$

Step 3. Calculate the score function $S(a_{ij})$ of z_{ij} through Formula (2), as shown in Table 4.

Table 4. $S(a_{ij})$ decision matrix of grey language hesitation fuzzy.

Industries	Attribute (P_1)	Attribute (P_2)	Attribute (P_3)	Attribute (P_4)
A_1	1.4000	1.4400	1.0333	0.8400
A_2	1.7800	1.7400	1.2600	1.7600
A_3	2.6133	3.0462	2.8933	0.9800
A_4	1.9000	1.6000	1.9733	2.4033

Step 4. Calculate grey correlation coefficient based on $S(a_i)$ to obtain index weight $w(j)$.

$$\tilde{\xi}_i(j) = \begin{bmatrix} 0.4339 & 0.3667 & 0.3333 & 0.3730 \\ 0.5274 & 0.4159 & 0.3628 & 0.5911 \\ 1.0000 & 1.0000 & 1.0000 & 0.3952 \\ 0.5659 & 0.3914 & 0.5027 & 1.0000 \end{bmatrix}$$

$$w = (0.2729, 0.2348, 0.2375, 0.2548)^T$$

Step 5. Calculate the comprehensive criterion values of each scheme through the GLHFWAA operator.

$$Z_1 = 0.2729 \times (H_{3.6667}, 0.6364, 0.4) + 0.2348 \times (H_3, 0.8, 0.4) + 0.2375 \times (H_3, 0.6889, 0.5) + 0.2548 \times (H_2, 0.7, 0.4) = (H_{2.9272}, 0.6996, 0.5)$$

Similarly,

$$Z_2 = (H_{3.4174}, 0.8008, 0.4), Z_3 = (H_{4.1446}, 0.7537, 0.3), Z_4 = (H_{3.8417}, 0.7681, 0.4).$$

Step 6. Compare $Z_i (i = 1, 2, 3, 4)$ and sort the schemes.

$$IH_{2.9272} \times 0.6697 \times (1 - 0.5) < IH_{3.4174} \times 0.8008 \times (1 - 0.4) < IH_{3.8417} \times 0.7681 \times (1 - 0.4) < IH_{4.1446} \times 0.7535 \times (1 - 0.3)$$

$Z_1 < Z_2 < Z_4 < Z_3$. Therefore, industry A_3 is the optimal, followed by industry A_4 and industry A_1 is the worst.

5.3. Comparison Method 3

Step 5 can also be used to continue the decision making after decision step 4 in method 2.

Step 5. Select the grey language $S(a_{ij})$ decision matrix and grey correlation weight $w(j)$ by using the multiple linear weighting method. The comprehensive criterion value of each scheme is calculated through the following formula:

$$Z_i = \sum_{j=1}^n S(a_{ij}) \times w(j)$$

$$= \begin{bmatrix} 1.4000 & 1.4400 & 1.0333 & 0.8400 \\ 1.7800 & 1.7400 & 1.2600 & 1.7600 \\ 2.6133 & 3.0462 & 2.8933 & 0.9800 \\ 1.9000 & 1.6000 & 1.9733 & 2.4033 \end{bmatrix} \times (0.2729, 0.2348, 0.2375, 0.2548)^T$$

$$= [1.1796 \ 1.6420 \ 2.3653 \ 1.9752]^T$$

$Z_1 < Z_2 < Z_4 < Z_3$, which is consistent with the sorting obtained through methods 1 and 2. By comparison, the grey scoring function used in method 1 is more reasonable and simple. Method 3 is simpler than method 2. On the whole, it can be concluded that method 1 \succ method 3 \succ method 2.

6. Conclusions

This paper studied the grey language decision making method, and defined the scoring function of grey language based on the kernel and grey scale which is more suitable for uncertain decision making. Moreover, a group decision-making method based on the grey correlation degree of the kernel and grey scale was proposed.

In comparison, the method is more reasonable than traditional methods by introducing the kernel and grey scale to avoid adding subjective data, thereby reducing information distortion; in addition, the grey relational degree decision method does not need to calculate the index weight, and hence to some extent this method is more convenient. Of course, it is hard to avoid the lack of depth only through a comparison of three different decision methods of the same example, so we will conduct systematic comparative research on data simulation in future research so as to improve the credibility of the comparison.

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References

1. Zadeh, L.A. Fuzzy sets. *Inf. Control* **1965**, *8*, 338–353. [[CrossRef](#)]
2. Nassov, K. Intuitionistic fuzzy sets. *Fuzzy Sets Syst.* **1986**, *20*, 87–96.

3. Atanassov, K.T. Intuitionistic fuzzy sets. *Fuzzy Sets Syst.* **1995**, *75*, 401–402. [[CrossRef](#)]
4. Dubois, D.; Prade, H. *Fuzzy Sets and Systems: Theory and Applications*; Academic Press: New York, NY, USA, 1980.
5. Yamoto, S. Remarks on basics of fuzzy sets and fuzzy multisets. *Fuzzy Sets Syst.* **2005**, *156*, 427–431.
6. Yager, R.R. On the theory of bags. *Int. J. Gen. Syst.* **1986**, *13*, 23–37. [[CrossRef](#)]
7. Miyamoto, S. Multisets and fuzzy multisets. In *Soft Computing and Human-Centered Machines*; Springer: Berlin, Germany, 2000.
8. Xia, M. *Research on Fuzzy Decision Information Integration and Its Measure*; Southeast China University: Nanjing, China, 2012.
9. Deng, J.L. Control problems of grey systems. *Syst. Control Lett.* **1982**, *1*, 288–294.
10. Deng, J. Grey control system. *J. Huazhong Univ. Technol.* **1982**, *3*, 9–18.
11. Liu, S. Advances in grey system research (2004–2014). *J. Nanjing Univ. Aeronaut. Astronaut.* **2015**, *47*, 1–18.
12. Luo, D. *Study of Analysis Method of Grey Decision Problem*; Nanjing University of Aeronautics and Astronautics: Nanjing, China, 2005.
13. Chen, X. *Study of Grey Multi-Attribute Decision-Making Method*; Nanjing University of Aeronautics and Astronautics: Nanjing, China, 2009.
14. Song, J. *Research on Grey Decision Method and Its Application*; Nanjing University of Aeronautics and Astronautics: Nanjing, China, 2010.
15. Hu, L.; Guan, X.; He, Y. A new approach for grey multi-attribute decision making. *Control Decision* **2012**, *27*, 895–898, 908.
16. You, C.; Wu, L. Research on inverse order problem of Grey decision Model. *Stat. Decision* **2013**, *22*, 44–46.
17. Zhang, N.; Fang, Z.; Zhu, J. Multi-Stage grey situation group decision-making model based on Orness. *Control Decision* **2015**, *7*, 1227–1232.
18. Liu, S.; Fang, Z.; Yang, Y. Two stages decision model with grey synthetic measure and a betterment of triangular whitenization weight function. *Control Decision* **2014**, *7*, 1232–1238.
19. Luo, D.; Liu, S. Study on the method for grey incidence decision-making. *Chin. J. Manag. Sci.* **2005**, *1*, 102–107.
20. Luo, D.; Liu, S. Grey incidence decision-making with incomplete information. *J. Appl. Sci.* **2005**, *4*, 408–412.
21. Dang, Y.; Liu, S.; Liu, B. Study on grey incidence decision model of the dynamic multiple-attribute. *Eng. Sci.* **2005**, *2*, 69–72.
22. Jiang, S.; Liu, S.; Liu, Z.; Fang, Z. Grey incidence decision making model based on area. *Control Decision* **2015**, *4*, 685–690.
23. Xie, M.; Xiao, X. Grey Relational Analysis of Multi-Attribute Group Decision Making Based on AHP and its Application. In Proceedings of the 16th National Academic Conference on Grey System, China Higher Science and Technology Center, Nanjing, China, 18–20 November 2007.
24. Chen, X.; Liu, S. Grey incidence decision-making method with partial weight information but with preference information on alternatives. *Syst. Eng. Electron.* **2007**, *11*, 1868–1871.
25. Li, P.; Liu, S. Interval-Valued intuitionistic fuzzy numbers decision-making method based on grey incidence analysis and d-s theory of evidence. *Acta Auomat. Sin.* **2011**, *8*, 993–998.
26. Liu, Y.; Jeffrey, F.; Liu, S.; Zhao, H.; Jian, L. Dynamic multiple attribute grey incidence decision making method based on interval valued intuitionistic fuzzy number. *Control Decision* **2013**, *9*, 1303–1308.
27. Li, P.; Zhu, J. Intuitionistic fuzzy cluster models based on case-based reasoning and grey incidence analysis. *Syst. Eng.-Theor. Pract.* **2015**, *9*, 2348–2353.
28. Li, P. *Research on Intuitionistic Fuzzy Information Decision-Making Method Based on Grey Relation and Evidence Reasoning*; Nanjing University of Aeronautics and Astronautics: Nanjing, China, 2012.
29. Li, P.; Liu, S.; Fang, Z. Interval-Valued intuitionistic fuzzy numbers decision-making method based on grey incidence analysis and MYCIN certainty factor. *Control Decision* **2012**, *7*, 1009–1014.
30. Hu, L.; Wang, C.; Zhu, J.; He, Y. Approach for grey fuzzy MADA in closed world. *Control Decision* **2014**, *2*, 246–250.
31. Li, P.; Liu, S.; Fang, Z. Intuitionistic fuzzy numbers decision-making method based on grey incidence analysis and MYCIN certainty factor. *Control Decision* **2011**, *26*, 1680–1684.
32. Li, P.; Liu, S.; Zhu, J. Intuitionistic fuzzy stochastic multi-criteria decision-making methods based on MYCIN certainty factor and prospect theory. *Syst. Eng.-Theor. Pract.* **2013**, *33*, 1509–1515.

33. Li, P.; Zhu, J.; Liu, S. Intuitionistic fuzzy decision-making methods based on case-based reasoning. *Chin. Manag. Manag. Sci.* **2015**, *23*, 113–118.
34. Wei, G.W. Grey relational analysis method for intuitionistic fuzzy multiple attribute decision making. *Expert Syst. Appl.* **2011**, *38*, 11671–11677. [[CrossRef](#)]
35. Wu, J. *Study on Grey Language Multi-Criteria Decision-Making Method and Its Application*; Central South University: Changsha, China, 2009.
36. Wang, J.; Wu, J. Multi-Criteria decision-making approach based on the interval grey uncertain linguistic. *Chin. J. Manag. Sci.* **2010**, *3*, 107–111.
37. Liu, P.; Zhang, X. Multi-Attribute group decision making method based on interval grey linguistic variables weighted geometric aggregation operator. *Control Decision* **2011**, *5*, 743–747.
38. Wang, H. *Research on the Model and Application of Linguistic Evaluation Information Aggregation Based on Grey Decision Making*; Nanjing University of Aeronautics & Astronautics: Nanjing, China, 2013.
39. Liu, S.; Li, Q.; Zhao, N. Grey correlation decision making method of grey hesitant fuzzy sets based on kernel and grey degree. *J. Nanjing Univ. Aeronaut. Astronaut.* **2016**, *48*, 683–688.
40. Li, Q.; Liu, S. Grey hesitant fuzzy sets and its decision making based on grey relation and TOPSIS. *J. Jiangsu Univ. Sci. Technol.* **2015**, *29*, 597–601, 606.
41. Li, Q.; Liu, S. Grey hesitant fuzzy set and its decision making based on grey relation and vikor method. *Math. Pract. Theor.* **2015**, *45*, 119–126.
42. Torra, V. Hesitant fuzzy sets. *Int. J. Intell. Syst.* **2010**, *25*, 529–539. [[CrossRef](#)]
43. Torra, V.; Narukawa, Y. On Hesitant Fuzzy Sets and Decision. In Proceedings of the 18th IEEE International Conference on Fuzzy Systems, Jeju Island, Korea, 20–24 August 2009.
44. Liu, S.; Fang, Z.; Xie, N. Algorithm rules of interval grey numbers based on the “Kernel” and the degree of greyness of grey numbers. *Syst. Eng. Electron.* **2010**, *32*, 313–316.
45. Zeng, B. Algebra algorithm rules of grey isomerism data based on kernel and grey degree and its application. *Stat. Inf. Forum* **2013**, *29*, 18–22.
46. Liu, L. Evaluation of the sustainability of the industrial chain based on circular economy. *J. Shandong Jianzhu Univ.* **2012**, *27*, 88–91, 100.



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