

Received 9 December 2023, accepted 14 January 2024, date of publication 18 January 2024, date of current version 30 January 2024. Digital Object Identifier 10.1109/ACCESS.2024.3355737

RESEARCH ARTICLE

An ExpTODIM-MACONT Based Multiple-Attribute Group Decision-Making Technique for Smart Classroom Teaching Evaluation of Basic English Under Interval-Valued Pythagorean Fuzzy Circumstances

YAMIN LIANG

School of Foreign Languages, Xianyang Normal University, Xianyang, Shaanxi 712000, China e-mail: liangyamin0102@126.com

This work was supported in part by the Shaanxi Provincial Scientific Research Project under Grant 2022HZ1894, in part by the Xianyang Normal University' Talent Project under Grant XSYQL201703, and in part by the Xianyang Normal University' Teaching Reform Project under Grant 2023YB54.

ABSTRACT With the rapid development of information technology, the university teaching model has also begun to break free from constraints from two dimensions of time and space, becoming a new teaching model - smart classroom. Supported by the theory of smart classrooms, the reform of basic English curriculum teaching in China has received much attention. Currently, the reform of basic English curriculum teaching mainly faces three problems: incomplete resource allocation, limited development conditions, and poor practicality. Based on this, utilizing the theory of smart classrooms, we can solve the problem of incomplete resource allocation by building an "educational community", innovate teaching models, enhance students' English autonomous learning and practical application abilities, and achieve strategic analysis of basic English curriculum teaching reform. The smart classroom teaching evaluation of basic English is a multiple-attribute group decision-making (MAGDM) problem. Recently, the Exponential TODIM (an acronym in Portuguese of interactive and multicriteria decision making) technique and Mixed Aggregation by Comprehensive Normalization Technique (MACONT) has been used to cope with MAGDM issues. The interval-valued Pythagorean fuzzy sets (IVPFSs) are used as a tool for characterizing uncertain information during the smart classroom teaching evaluation of basic English. In this manuscript, the intervalvalued Pythagorean fuzzy Exponential TODIM- MACONT (IVPF-ExpTODIM-MACONT) technique is built to solve the MAGDM under IVPFSs. In the end, a numerical case study for smart classroom teaching evaluation of basic English is implemented to validate the proposed technique. The main contributions of this paper are outlined: 1) the ExpTODIM and MACONT technique has been extended to IVPFSs; 2) Information Entropy technique is employed to manage the weight values under IVPFSs; 3) IVPF-ExpTODIM-MACONT technique is founded to implement the MAGDM under IVPFSs; 4) numerical case study for smart classroom teaching evaluation of basic English and some comparative analysis is supplied to verify the IVPF-ExpTODIM-MACONT technique.

INDEX TERMS Multiple attribute group decision making (MAGDM), interval-valued Pythagorean fuzzy sets (IVPFSs), Exponential TODIM (ExpTODIM) technique, MACONT technique, smart classroom teaching evaluation.

I. INTRODUCTION

The associate editor coordinating the review of this manuscript and approving it for publication was Agostino Forestiero¹⁰.

The construction of smart classroom has become the main development direction of basic English teaching in the era of "Internet plus", which requires a relatively long

process to actively innovate classroom teaching forms [1], [2]. In the teaching of basic English courses, we generally emphasize the importance of putting practice first. The socalled practice is not only about the country vigorously practicing and finding suitable teaching forms, but also about universities and teachers constantly practicing, innovating teaching models, stimulating the motivation of teachers and students to learn [3], [4], [5]. It is also about cultivating students' awareness of self-directed and continuous learning. This internal motivation is the key to truly promoting the reform of basic English course education. To establish an autonomous learning model, it is necessary to work together from multiple directions, utilize the form of smart classrooms, and create more discussion opportunities for students during the teaching process. For example, large class teaching and small class discussions can be adopted to centralize learning time, allowing students to find the meaning and fun of basic English courses in the process of autonomous learning and communication. Only based on fun can students better stimulate their learning enthusiasm [6], [7], [8]. At the same time, in a learning atmosphere of mutual communication and complementarity, the speed of knowledge dissemination can be accelerated, and the problem of insufficient effectiveness of basic English courses can be solved [9], [10]. By setting up "thinking points", the teaching content can also be centralized, allowing students to improve learning efficiency and abilities during the process of self-reflection. This can cultivate students' thinking leaps, better open up their thinking and knowledge, and cultivate their autonomous learning ability. The importance of basic English curriculum education is already self-evident, and it is of great significance for the cultivation of versatile talents and the future talent planning of the country [11]. Especially in the context of the rise of smart classrooms, the reform of basic English curriculum education has encountered unprecedented development opportunities. We should start the reform with targeted measures, promote the smooth progress of the reform, construct reform strategies, and accelerate the further implementation of basic English curriculum education. Firstly, in response to the difficulty in implementing the educational philosophy of basic English courses, it can be solved by establishing basic curriculum units [12], [13]. A successful example of the implementation of the concept of basic course units is Fudan University. Some University is one of the most successful universities in China in implementing English course education. It is by adding a course group between courses and modules that the structure of university English courses is adjusted more meticulously, making the connection between courses closer, and then turning it into a systematic course structure. Secondly, in response to issues such as the lack of concentration of teaching staff and difficulty in mobilizing teachers' enthusiasm, an "educational community" can be established to solve them. Teachers who teach basic English courses are generally professional teachers from various universities and colleges, and there are not many issues with their knowledge [14], [15]. However, under the traditional teacher selection system, professional teachers are generally trapped in the misconception of scientific research achievements and promotion. The inconsistent recognition of English education concepts makes it difficult for the teaching team to spontaneously form an educational consensus, Therefore, it is necessary to further regulate and promote teacher teaching behavior through relevant systems and measures [16], [17]. Finally, in response to the widespread lack of effectiveness in curriculum teaching in current basic English curriculum education, the focus of the reform can be to change the teaching mode from the traditional teacher-based teaching mode to the student independent learning mode. No matter how well the curriculum is set, without effective teaching from teachers and students' self-directed learning as the foundation, it will be ineffective in terms of teaching effectiveness [18], [19]. This point is still extremely important in the current reform of basic English curriculum education. By analyzing the problems in the teaching of basic English courses, based on smart classrooms, the information-based support environment consists of digital learning resources and smart teaching platforms. The fundamental task is to support teaching interaction and self-directed learning, develop students' intelligent abilities, promote students' intelligent behavior, construct teaching reform strategies for basic English courses, cultivate students' interest in English learning, and improve the efficiency of students' English learning [20], [21].

MADM, as a commonly used decision-making method [22], [23], [24], is often difficult to evaluate things with accurate values due to limitations in decision-makers' knowledge level and the fuzziness of things themselves [25], [26]. In response to this issue, Zadeh [27] first proposed fuzzy sets in 1965 to characterize the uncertainty of information. Atanassov [28] extended it and proposed intuitionistic fuzzysets (IFS) that include membership, non-membership and hesitancy, with a total of 1. But when the membership degree and non-membership degree are independently implemented by the decision-maker, due to the fuzziness of the decision itself and the varying degrees of detail in describing things, there may be situations where their total is greater than 1. Therefore, Yager [29] proposed the Pythagorean fuzzy sets (PFSs), where the sum of the squares of membership and non-membership is not greater than 1, expanding its spatial range [30], [31], [32]. As the complexity and uncertainty of decision-making continue to increase, decision-makers may find it difficult to accurately give membership or nonmembership degrees with a specific value. In response to this situation, combined with the relevant knowledge of interval numbers [33], [34], the concept of interval valued PFSs (IVPFSs) [35] has been proposed, which can represent membership or non-membership in the form of intervals. In the field of MADM, Haktanir and Kahraman [36] measured and prioritized product design requirements and customer needs based on IVPFS, and combines quality function deployment (QFD) to conduct evaluation decisions in the development and application process of solar photovoltaic power

generation. Tang and Yang [37] implemented the concept of interval valued Pythagorean fuzzy preference relationship (IVPFPR) for the selection of shared electric bicycle recycling suppliers, and proposes a Pythagorean fuzzy decisionmaking method that considers trust level. Fu et al. [38] proposed a product ranking method that combines feature opinion IVPFS to effectively analyze and extract relevant opinions from online reviews, introducing IVPFS to represent interval values; References [39], [40], and [41] utilized the relevant measures of membership and non-membership in IVPFS to conduct decision analysis and research on risk assessment of autonomous vehicles, sustainable supplier selection, and high-tech project portfolio evaluation.

The smart classroom teaching evaluation of basic English is classical MAGDM. Recently, the Exponential TODIM (ExpTODIM) technique [42], [43] and MACONT technique [44], [45] has been used to cope with MAGDM issues. The IVPFSs [35] are used as a technique for characterizing uncertain information during the smart classroom teaching evaluation of basic English. Until now, there is not related research works to implement the ExpTODIM technique [42] and MACONT technique [44], [45], [46] under IVPFSs [35]. Thus, the main aim of this paper is to investigate the ExpTODIM-MACONT technique under IVPFSs and construct the corresponding MAGDM technique for smart classroom teaching evaluation of basic English. In this paper, the interval-valued Pythagorean fuzzy Exponential TODIM-MACONT (IVPF-ExpTODIM-MACONT) technique is built to solve the MAGDM under IVPFSs. The information Entropy is used to obtain the weight values based on the score value and accuracy value under IVPFSs. At last, a numerical case study for smart classroom teaching evaluation of basic English is implemented to validate the proposed technique. The main research aim and motivation of this paper is constructed: (1) the Exponential ExpTODIM-MACONT technique is extended to the IVPFSs; (2) the intervalvalued Pythagorean fuzzy Exponential TODIM-MACONT (IVPF-ExpTODIM-MACONT) technique is built to solve the MAGDM under IVPFSs; (3) the information Entropy is used to obtain the weight values based on the score value and accuracy value under IVPFSs; (4) Finally, a numerical case study for smart classroom teaching evaluation of basic English is implemented to validate the proposed technique.

The framework of this paper is implemented below. In Section II, the IVPFSs is implemented. In Section III, IVPF-ExpTODIM-MACONT technique is implemented under IVPFSs with entropy. Section IV implements an illustrative case for smart classroom teaching evaluation of basic English and some comparative analysis. Some remarks are implemented in Section V.

II. PRELIMINARIES

The PFSs and IVPFSs are implemented.

Definition 1 ([47]): Let Θ be fix set, the PFS is implemented:

$$AT = \{ \langle \theta, (AM_{AT} (\theta), AM_{AT} (\theta)) \rangle | \theta \in \Theta \}$$
(1)

where $AM_{AT}(\theta) \subseteq [0, 1]$, $AN_{AT}(\theta) \subseteq [0, 1]$ denotes the membership non-membership of $\theta \in \Theta$ to AT, and, for $\theta \in \Theta$, it has the implemented condition

$$\left(AM_{AT}\left(\theta\right)\right)^{2} + \left(AN_{AT}\left(\theta\right)\right)^{2} \le 1.$$
 (2)

Definition 2 ([47]): Let $T_1 = (M_1, N_1)$, $AT_2 = (AM_2, AN_2)$, and AT = (AM, AN) be three PFNs, and some operations are implemented.

$$(1) AT_{1} \oplus AT_{2} = \left(\sqrt{(AM_{1})^{2} + (AM_{2})^{2} - (AM_{1})^{2} (AM_{2})^{2}}, AN_{1}AN_{2}\right);$$

$$(2) AT_{1} \otimes AT_{2} = (AM_{1}AM_{2}, \sqrt{(AN_{1})^{2} + (AN_{2})^{2} - (AN_{1})^{2} (AN_{2})^{2}});$$

$$(3) \lambda AT = \left(\sqrt{1 - (1 - AM^{2})^{\lambda}}, AN^{\lambda}\right), \lambda > 0;$$

$$(4) (AT)^{\lambda} = \left(AM^{\lambda}, \sqrt{1 - (1 - AN^{2})^{\lambda}}\right), \lambda > 0.$$

Liang, Zhang and Liu [35] implemented the intervalvalued PFSs (IVPFSs).

Definition 3 ([35]): Let Θ be a fix set. An IVPFS is implemented:

$$XT = \{ \langle \theta, (XM_{VT}(\theta), XN_{VT}(\theta)) \rangle | \theta \in \Theta \}$$
(3)

where $XM_{VT}(\theta) \subseteq [0, 1], XN_{VT}(\theta) \subseteq [0, 1]$ are interval values, $XM_{XT}(\theta) = [XM_{VT}^{L}(\theta), XM_{VT}^{R}(\theta)], XN_{XT}(\theta) = [XN_{XT}^{L}(\theta), XN_{XT}^{R}(\theta)], 0 \leq XM_{XT}^{R}(\theta) + XN_{XT}^{R}(\theta) \leq 1,$ $\forall \theta \in \Theta, XM_{XT}(\theta), XN_{XT}(\theta)$ denote the membership and no-membership.

For convenience, $XT = ([XM^L, XM^R], [XN^L, XN^R])$ is an interval-valued Pythagorean fuzzy number (IVPFN), and $[XM^L, XM^R], [XN^L, XN^R] \subset [0, 1],$ $(XM^R)^2 + (XM^R)^2 \leq 1.$

Definition 4 ([48]): Let $XT_1 = ([XM_1^L, XM_1^R], [XN_1^L], XN_1^R])$, $XT_2 = ([XM_2^L, XM_2^R], [XN_2^L, XN_2^R])$, and $XT = ([XM^L, XM^R], [XN^L, XN^R])$ be three IVPFNs, and some operations are implemented as shown in the equation at the bottom of the next page.

Definition 5 ([48]): Let

$$XT_{1} = \left(\left[XM_{1}^{L}, XM_{1}^{R} \right], \left[XN_{1}^{L}, XN_{1}^{R} \right] \right),$$

$$XT_{2} = \left(\left[XM_{2}^{L}, XM_{2}^{R} \right], \left[XN_{2}^{L}, XN_{2}^{R} \right] \right),$$

$$SV (XT_{1}) = \frac{1}{4} \begin{bmatrix} \left(1 + \left(XM_{1}^{L} \right)^{2} - \left(XN_{1}^{L} \right)^{2} \right) \\ + \left(1 + \left(XM_{1}^{R} \right)^{2} - \left(XN_{1}^{R} \right)^{2} \right) \end{bmatrix}$$

and

$$SV(XT_2) = \frac{1}{4} \begin{bmatrix} \left(1 + \left(XM_2^L \right)^2 - \left(XN_2^L \right)^2 \right) \\ + \left(1 + \left(XM_2^R \right)^2 - \left(XN_2^R \right)^2 \right) \end{bmatrix}$$

is the scores value of

$$XT_1 = \left(\left[XM_1^L, XM_1^R \right], \left[XN_1^L, XN_1^R \right] \right)$$

and

$$XT_{2} = \left(\left[XM_{2}^{L}, XM_{2}^{R} \right], \left[XN_{2}^{L}, XN_{2}^{R} \right] \right),$$

$$AV (XT_{1}) = \frac{1}{2} \left[\left(\left(XM_{1}^{L} \right)^{2} + \left(XN_{1}^{L} \right)^{2} \right) + \left(XM_{1}^{R} \right)^{2} + \left(XN_{1}^{R} \right)^{2} \right]$$

and

$$AV (XT_2) = \frac{1}{2} \begin{bmatrix} \left(\left(XM_2^L \right)^2 + \left(XN_2^L \right)^2 \right) \\ + \left(XM_2^R \right)^2 + \left(XN_2^R \right)^2 \end{bmatrix}$$

is the accuracy value of

$$XT_1 = \left(\left[XM_1^L, XM_1^R \right], \left[XN_1^L, XN_1^R \right] \right)$$

and

$$XT_2 = \left(\left[XM_2^L, XM_2^R \right], \left[XN_2^L, XN_2^R \right] \right),$$

then if

$$SV(XT_1) < SV(XT_2), XT_1 < XT_2;$$

if $SV(XT_1) = SV(XT_2)$, (1) if $AV(XT_1) = AV(XT_2)$, $XT_1 = XT_2$; (2) if $AV(XT_1) < AV(XT_2)$, $XT_1 < XT_2$.

Definition 6 ([35]): Let $XT_1 = ([XM_1^L, XM_1^R], [XN_1^L], XN_1^R])$, $XT_2 = ([XM_2^L, XM_2^R], [XN_2^L, XN_2^R])$, the normalized Hamming distance is:

$$HD(XT_{1}, XT_{2}) = \frac{1}{4} \begin{pmatrix} \left| (XM_{1}^{L})^{2} - (XM_{2}^{L})^{2} \right| + \left| (XM_{1}^{R})^{2} - (XM_{2}^{R})^{2} \right| \\ + \left| (XL_{1}^{L})^{2} - (XL_{2}^{L})^{2} \right| + \left| (XN_{1}^{R})^{2} - (XN_{2}^{R})^{2} \right| \end{pmatrix}$$

$$(4)$$

III. IVPF-ExpTODIM-MACONT TECHNIQUE FOR MAGDM WITH ENTROPY WEIGHT

The IVPF-ExpTODIM-MACONT technique is implemented for MAGDM. Let $XA = \{XA_1, XA_2, \dots, XA_m\}$ be alternatives, and the attributes set $XG = \{XG_1, XG_2, \dots, XG_n\}$ with weight vw, where $xw_j \in [0, 1], \sum_{j=1}^n xw_j = 1$ and a set of invited experts $DE = \{DE_1, DE_2, \dots, DE_q\}$, let expert's weight be $\{dw_1, dw_2, \dots, dw_q\}$ $dw_t \in [0, 1], \sum_{t=1}^q dw_t =$ 1. Then, IVPF-ExpTODIM-MACONT technique is implemented for MAGDM (See Figure 1).

A. STAGE 1: OBTAIN THE DECISION INFORMATION WITH IVPFSs

Step 1. Implement the IVPFN group decision matrix

$$\begin{aligned} XR^{(t)} &= \left[XR_{ij}^{(t)} \right]_{m \times n} \\ &= \left(\left[XM_{ij}^{L(t)}, XM_{ij}^{R(t)} \right], \left[XN_{ij}^{L(t)}, XN_{ij}^{R(t)} \right] \right)_{m \times n} \end{aligned}$$

and calculate the average matrix $XR = [XR_{ij}]_{m \times n} = ([XM_{ij}^L, XM_{ij}^R], [XN_{ij}^L, XN_{ij}^R])_{m \times n}$:

$$XR^{t} = \begin{bmatrix} XR_{ij}^{t} \end{bmatrix}_{m \times n} = \begin{bmatrix} XA_{1} \\ XA_{2} \\ \vdots \\ XA_{m} \end{bmatrix} \begin{bmatrix} XR_{11}^{t} & XR_{12}^{t} \dots & XR_{1n}^{t} \\ XR\phi_{21}^{t} & XR_{22}^{t} \dots & XR_{2n}^{t} \\ \vdots & \vdots & \vdots \\ XR_{m1}^{t} & XR_{m2}^{t} \dots & XR_{mn}^{t} \end{bmatrix}$$
(5)
$$XR = \begin{bmatrix} XR_{ij} \end{bmatrix}_{m \times n} = \begin{bmatrix} XA_{1} \\ XA_{2} \\ \vdots \\ XA_{m} \end{bmatrix} \begin{bmatrix} XR_{11} & XR_{12} \dots & XR_{1n} \\ XR_{21} & XR_{22} \dots & XR_{2n} \\ \vdots & \vdots & \vdots \\ XR_{m1} & XR_{m2} \dots & XR_{mn} \end{bmatrix}$$
(6)

In line with the IVPFWA technique, the $XR = [XR_{ij}]_{m \times n} = \left(\left[XM_{ij}^L, XM_{ij}^R \right], \left[XN_{ij}^L, XN_{ij}^R \right] \right)_{m \times n}$ is

$$(1)XT_{1} \oplus XT_{2} = \left(\begin{bmatrix} \sqrt{(XM_{1}^{L})^{2} + (XM_{2}^{L})^{2} - (XM_{1}^{L})^{2} (XM_{2}^{L})^{2}}, \\ \sqrt{(XM_{1}^{R})^{2} + (XM_{2}^{R})^{2} - (XM_{1}^{R})^{2} (XM_{2}^{R})^{2}} \end{bmatrix}, \begin{bmatrix} XN_{1}^{L}XN_{2}^{L}, XN_{1}^{R}XN_{2}^{R} \end{bmatrix} \right); \\ (2)XT_{1} \otimes XT_{2} = \left(\begin{bmatrix} XM_{1}^{L}XM_{2}^{L}, XM_{1}^{R}XM_{2}^{R} \end{bmatrix}, \begin{bmatrix} \sqrt{(XN_{1}^{L})^{2} + (XN_{2}^{L})^{2} - (XN_{1}^{L})^{2} (XN_{2}^{L})^{2}}, \\ \sqrt{(XN_{1}^{R})^{2} + (XN_{2}^{R})^{2} - (XN_{1}^{R})^{2} (XN_{2}^{R})^{2}} \end{bmatrix} \right); \\ (3) \lambda XT = \left(\begin{bmatrix} \sqrt{1 - (1 - (XM^{L})^{2})^{\lambda}}, \sqrt{1 - (1 - (XM^{R})^{2})^{\lambda}} \end{bmatrix}, \begin{bmatrix} (XN^{L})^{\lambda}, (XN^{R})^{\lambda} \end{bmatrix} \right), \lambda > 0; \\ (4) (XT)^{\lambda} = \left(\begin{bmatrix} (XM^{L})^{\lambda}, (XM^{R})^{\lambda} \end{bmatrix}, \begin{bmatrix} \sqrt{1 - (1 - (XN^{L})^{2})^{\lambda}}, \sqrt{1 - (1 - (XN^{R})^{2})^{\lambda}} \end{bmatrix} \right), \lambda > 0; \\ (5) XT^{c} = \left(\begin{bmatrix} XN^{L}, XN^{R} \end{bmatrix}, \begin{bmatrix} XM^{L}, XM^{R} \end{bmatrix} \right). \end{cases}$$

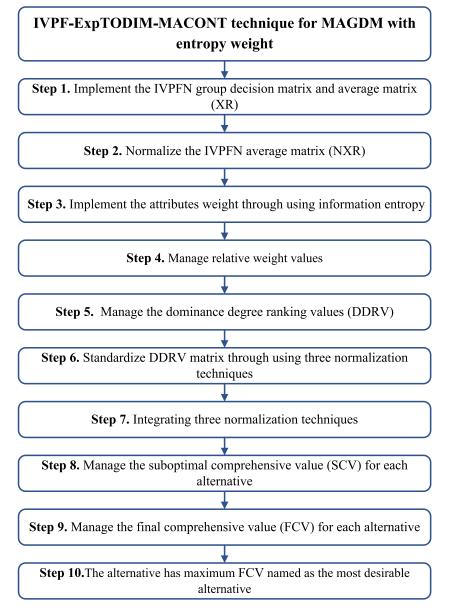


FIGURE 1. IVPF-ExpTODIM-MACONT technique for MAGDM with entropy weight.

implemented:

$$XR_{ij} = dw_1 XR_{ij}^1 \oplus dw_2 XR_{ij}^2 \oplus \dots \oplus dw_q XR_{ij}^q$$

$$= \begin{pmatrix} \left[\sqrt{1 - \prod_{t=1}^q \left(1 - \left(XM_{ij}^{L(t)} \right)^2 \right)^{dw_t}}, \\ \sqrt{1 - \prod_{t=1}^q \left(1 - \left(XM_{ij}^{R(h)} \right)^2 \right)^{dw_t}} \\ \left[\prod_{t=1}^q \left(XN_{ij}^{L(t)} \right)^{dw_t}, \prod_{t=1}^q \left(XN_{ij}^{R(t)} \right)^{dw_t} \\ \right] \end{pmatrix},$$
(7)
$$Step 2. \text{ Normalize the } XR = [XR_{ij}]_{m \times n} \text{ into } NXR = [NXR_{ij}]_{m \times n} = \begin{pmatrix} \left[NXM_{ij}^L, NXM_{ij}^R \\ NXN_{ij}^L, NXN_{ij}^R \\ \end{bmatrix}, \\ \end{pmatrix}$$

For benefit attributes:

$$NXR_{ij} = \left(\left[NXM_{ij}^{L}, NXM_{ij}^{R} \right], \left[NXN_{ij}^{L}, NXN_{ij}^{R} \right] \right)$$
$$= XR_{ij} = \left(\left[XM_{ij}^{L}, XM_{ij}^{R} \right], \left[XN_{ij}^{L}, XN_{ij}^{R} \right] \right)$$
(8)

For cost attributes:

$$NXR_{ij} = \left(\left[NXM_{ij}^L, NXM_{ij}^R \right], \left[NXN_{ij}^L, NXN_{ij}^R \right] \right) \\ = \left(\left[XN_{ij}^L, XN_{ij}^R \right], \left[XM_{ij}^L, XM_{ij}^R \right] \right)$$
(9)

B. STAGE 2: IMPLEMENT THE ATTRIBUTES WEIGHT THROUGH USING INFORMATION ENTROPY

Step 3.Implement the attributes weight through using information entropy

The weight is important for MAGDM [49], [50], [51]. Entropy [52] is a conventional technique to derive weight. Firstly, the normalized IVPF-matrix XNN_{ij} is implemented:

$$XNN_{ij} = \frac{AV\left(\left[NXM_{ij}^{L}, NXM_{ij}^{R}\right], \left[NXN_{ij}^{L}, NXN_{ij}^{R}\right]\right)}{SV\left(\left[NXM_{ij}^{L}, NXM_{ij}^{R}\right], \left[NXN_{ij}^{L}, NXN_{ij}^{R}\right]\right)}{\sum_{i=1}^{m} \frac{AV\left(\left[NXM_{ij}^{L}, NXM_{ij}^{R}\right], \left[NXN_{ij}^{L}, NXN_{ij}^{R}\right]\right)}{SV\left(\left[NXM_{ij}^{L}, NXM_{ij}^{R}\right], \left[NXN_{ij}^{L}, NXN_{ij}^{R}\right]\right)},$$
(10)

Then, the IVPFN Shannon information entropy (IVPFN-SIE) is implemented:

$$IVPFNSIE_{j} = -\frac{1}{\ln m} \sum_{i=1}^{m} XNN_{ij} \ln XNN_{ij}$$
(11)

and $XNN_{ij} \ln XNN_{ij} = 0$ if $XNN_{ij} = 0$.

Then, the weights $xw = (xw_1, xw_2, \cdots, xw_n)$ is implemented:

$$xw_{j} = \frac{1 - IVPFNSIE_{j}}{\sum_{j=1}^{n} (1 - IVPFNSIE_{j})}, \quad j = 1, 2, \cdots, n.$$
(12)

C. STAGE 3: IVPF-ExpTODIM-MACONT TECHNIQUE FOR MAGDM

The IVPF-ExpTODIM-MACONT technique is implemented for MAGDM.

Step 4. Manage relative weight of XG_i as:

$$rxw_j = xw_j \left/ \max_j xw_j, \right. \tag{13}$$

Step 5. Manage the dominance degree ranking values (DDRV).

(1) The $DDRV_j(XA_i, XA_t)$ of XA_i over XA_t for XG_j is managed (14), as shown at the bottom of the page, where π is built in line with. Tversky and Kahneman [53] and $\pi \in [1, 5]$ [42].

r

(2) The $DDRV_i$ (XA_i) ($j = 1, 2, \dots, n$) with respect to XG_i is implemented (15), as shown at the bottom of the page.

(3) The overall DDRV of XA_i over other alternatives for XG_i is managed:

$$DDRV_j(XA_i) = \sum_{t=1}^m DDRV_j(XA_i, XA_t)$$
(16)

The overall DDRV matrix is implemented (17), as shown at the bottom of the next page.

Step 6. Standardize DDRV matrix through using three normalization techniques:

$$= \max_{j=1}^{n} DDRV_{ij}, IVPFNNIS_j = \min_{j=1}^{n} DDRV_{ij}$$
(18)

$$NDDRV_{ij}^{(1)} = \begin{cases} \frac{DDRV_{ij} - IVPFNNIS_j}{IVPFNPIS_j - IVPFNNIS_j}, & for benefit attribute\\ \frac{DDRV_{ij} - IVPFNPIS_j}{IVPFNNIS_j - IVPFNPIS_j}, & for \cos t attribute \end{cases}$$
(19)

 $NDDRV_{ii}^{(2)}$

IVPFNPIS_i

$$= \begin{cases} \frac{e^{DDRV_{ij}}}{\sum\limits_{i=1}^{m} e_{ij}^{DDRV_{ij}}}, \text{ for benefit attribute} \\ \frac{1/e^{DDRV_{ij}}}{\sum\limits_{i=1}^{m} (1/e^{DDRV_{ij}})}, \text{ for benefit attribute} \end{cases}$$
(20)

$$NDDRV_{ij}^{(3)}$$

$$= \begin{cases} \frac{e^{DDRV_{ij}}}{\max_{i} (e^{DDRV_{ij}})}, & \text{for benefit attribute} \\ \frac{\min_{i} (e^{DDRV_{ij}})}{\frac{i}{e^{DDRV_{ij}}}, & \text{for benefit attribute} \end{cases}$$
(21)

$$DDRV_{j}(XA_{i}, XA_{t}) = \begin{cases} \frac{rxw_{j} \times \left(1 - 10^{-\rho HD(NXR_{ij}, NXR_{ij})}\right)}{\sum_{j=1}^{n} rxw_{j}} & \text{if } SF(NXR_{ij}) > SF(NXR_{ij}) \\ 0 & \text{if } SF(NXR_{ij}) = SF(NXR_{ij}) \\ -\frac{1}{\pi} \frac{\sum_{j=1}^{n} rxw_{j} \times \left(1 - 10^{-\rho HD(NXR_{ij}, NXR_{ij})}\right)}{rxw_{j}} & \text{if } SF(NXR_{ij}) < SF(NXR_{ij}) \end{cases}$$
(14)

$$DDRV_{j} (XA_{i}) = \begin{bmatrix} DDRV_{j} (XA_{i}, XA_{t}) \end{bmatrix}_{m \times m} \\ XA_{1} & XA_{2} & \cdots & XA_{m} \\ = \frac{XA_{1}}{XA_{2}} \begin{bmatrix} 0 & DDRV_{j} (XA_{1}, XA_{2}) & \cdots & DDRV_{j} (XA_{1}, XA_{m}) \\ DDRV_{j} (XA_{2}, XA_{1}) & 0 & \cdots & DDRV_{j} (XA_{2}, XA_{m}) \\ \vdots & \vdots & \ddots & \vdots \\ DDRV_{j} (XA_{m}, XA_{1}) & DDRV_{j} (XA_{m}, XA_{2}) & \cdots & 0 \end{bmatrix}$$
(15)

Step 7. Integrating three normalization techniques:

$$NDDRV_{ij} = \begin{pmatrix} \lambda NDDRV_{ij}^{(1)} + \mu NDDRV_{ij}^{(2)} \\ + (1 - \lambda - \mu) NDDRV_{ij}^{(3)} \end{pmatrix}$$
(22)

Among them, $0 \leq \lambda, \mu \leq 1$, the values of these two equilibrium parameters are determined by experts. If the expert pays more attention to a particular option among all alternative options, the λ value is higher; If experts want to highlight the best attributes of alternative solutions, the μ value is higher; If experts want to emphasize the gap between alternative solutions, that is, highlight the best and worst attributes of the alternative solutions, then the λ, μ value of and is smaller.

Step 8. Manage the suboptimal comprehensive value (SCV) for each alternative:

$$SCV_{i}^{(1)} = \delta \frac{XQ_{i}^{(1)}}{\sqrt{\sum_{i=1}^{m} \left(XQ_{i}^{(1)}\right)^{2}}} + (1-\delta) \frac{XQ_{i}^{(2)}}{\sqrt{\sum_{i=1}^{m} \left(XQ_{i}^{(2)}\right)^{2}}}$$
(23)

where,

$$XQ_{i}^{(1)} = \sum_{j=1}^{n} xw_{j} \left(NDDRV_{ij} - \frac{1}{m} \sum_{i=1}^{m} NDDRV_{ij} \right),$$

$$XQ_{i}^{(2)} = \frac{\prod_{x\gamma=1}^{n} \left(\frac{1}{m} \sum_{i=1}^{m} NDDRV_{ij} - NDDRV_{ij} \right)^{xw_{j}}}{\prod_{x\eta=1}^{n} \left(NDDRV_{ij} - \frac{1}{m} \sum_{i=1}^{m} NDDRV_{ij} \right)^{xw_{j}}},$$

$$SCV_{i}^{(2)} = \left(\frac{\theta \max_{j} \left(xw_{j} \left(NDDRV_{ij} - \frac{1}{m} \sum_{i=1}^{m} NDDRV_{ij} - \frac{1}{m} \sum_{i=1}^{m} NDDRV_{ij} \right) \right)}{\left(1 - \theta \right) \min_{j} \left(xw_{j} \left(\frac{NDDRV_{ij} - 1}{m} \sum_{i=1}^{m} NDDRV_{ij} \right) \right) \right)}$$

$$(24)$$

where $x\gamma$ ($x\gamma = 1, 2, \dots, n$) represent the part of attributes that satisfy $\frac{1}{m} \sum_{i=1}^{m} NDDRV_{ij} \ge NDDRV_{ij}$, and $x\eta$ ($x\eta = 1, 2, \dots, n$) represent the part of attributes that

(DDDU)

DDDU

satisfy $NDDRV_{ij} \ge \frac{1}{m} \sum_{i=1}^{m} NDDRV_{ij}$. In addition, δ and $\theta \ (0 \le \delta, \theta \le 1)$ are preference parameters.

Step 9. Manage the final comprehensive value (FCV) for each alternative:

$$FCV_{i} = \frac{1}{2} \left(SCV_{i}^{(1)} + \frac{SCV_{i}^{(2)}}{\sqrt{\sum_{i=1}^{m} \left(SCV_{i}^{(2)}\right)^{2}}} \right)$$
(25)

Step 10. In line with FCV, the largest FCV is the optimal choice.

IV. EMPIRICAL EXAMPLE AND COMPARATIVE ANALYSIS A. EMPIRICAL EXAMPLE FOR SMART CLASSROOM TEACHING EVALUATION OF BASIC ENGLISH

The teaching of basic English courses places great emphasis on the concept of effective teaching, which means that under the guidance of objective teaching laws, teachers spend the least time and energy to achieve the prescribed teaching objectives, achieve the teaching objectives, and meet the needs of students as much as possible. The process of effective teaching by teachers is essentially the process of disseminating the content of basic English courses to students. With the widespread dissemination of the concept of basic English curriculum, it is widely regarded by the academic community as a product of the combination of traditional education methods in China and modern educational concepts in the West. Chinese universities have had the concept of developing "generalist" education for a long time, but in the later stage, as national policies began to learn from the single talent education model of the Soviet Union, the education model mainly focused on cultural cultivation became the main form of university education in China. Over time, the drawbacks of a single talent cultivation model have gradually emerged, and the talents cultivated by universities cannot adapt to social needs. Under the longterm accumulation of social contradictions, after the reform and opening up, China began to realize that offering basic English courses and cultivating diverse talents can meet the development needs of adapting to the national economic system. In the late 20th century, the country began to

$$DDRV = (DDRV_{ij})_{m \times n}$$

$$= \begin{bmatrix} XG_1 & XG_2 & \dots & XG_n \\ XA_1 & \sum_{t=1}^{m} DDRV_1 (XA_1, XA_t) & \sum_{t=1}^{m} DDRV_2 (XA_1, XA_t) & \dots & \sum_{t=1}^{m} DDRV_n (XA_1, XA_t) \\ XA_2 & \sum_{t=1}^{m} DDRV_1 (XA_2, XA_t) & \sum_{t=1}^{m} DDRV_2 (XA_2, XA_t) & \dots & \sum_{t=1}^{m} DDRV_n (XA_2, XA_t) \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ XA_m & \sum_{t=1}^{m} DDRV_1 (XA_m, XA_t) & \sum_{t=1}^{m} DDRV_2 (XA_m, XA_t) & \dots & \sum_{t=1}^{m} DDRV_n (XA_m, XA_t) \end{bmatrix}$$
(17)

IEEEAccess

TABLE 1. Linguistic scale and IVPFNs [54].

_

Linguistic Terms	IVPFNs←
Exceedingly Terrible-XET←	$\langle [0.00, 0.10], [0.85, 0.90] \rangle \in$
Very Terrible-XVT←	$\langle [0.00, 0.10], [0.70, 0.75] \rangle \in$
Terrible-XT←	$\langle [0.30, 0.40], [0.45, 0.50] \rangle < 1$
Medium-XM←	$\langle [0.40, 0.50], [0.35, 0.40] \rangle < 1$
Well-XW [⊲]	$\langle [0.60, 0.70], [0.15, 0.20] \rangle < 1$
Very Well-XVW←	$\langle [0.70, 0.80], [0.05, 0.10] \rangle < 1$
Exceedingly Well-XEW←	$\langle [1.00, 1.00], [0.00, 0.00] \rangle$

TABLE 2. Linguistic scale through XE_1 .

	XG_1	XG_2	XG_3	XG4
XA ₁	XVW	XW	XVT	XM
XA_2	XM	XW	XVT	XVW
XA ₃	XM	XT	XVW	XW
XA_4	XVT	XM	XVW	XT
XA_5	XT	XW	XM	XVW

TABLE 3. Linguistic scale through XE₂.

	XG_1	XG_2	XG ₃	XG_4
XA_1	XW	XM	XVT	XVW
XA_2	XM	XT	XVW	XW
XA ₃	XW	XVW	XVT	XM
XA_4	XVT	XVW	XM	XVT
XA ₅	XVW	XVT	XM	XT

increase the social practice of college students, while solidly cultivating their professional foundation, reducing the variety of professional courses. This measure actively promoted the attention of various sectors of society to the development of basic English courses. However, the development of basic English courses still has a long way to go. At the end of

	XG_1	XG_2	XG_3	XG_4
XA_1	XW	XVW	XT	XM
XA_2	XVT	XVT	XM	XW
XA ₃	XM	XT	XVW	XW
XA_4	XW	XVW	XVT	XT
XA ₅	XM	XW	XT	XM

TABLE 4. Linguistic scale through XE₃.

TABLE 5. The $XR = \left[XR_{ij}\right]_{5\times 4}$.

Alternatives	XG_2	XG_2	
XA_1	([0.1524,0.4531], [0.2678,0.5961])	([0.1647,0.3251], [0.2780,0.3806])	
XA_2	([01732,0.2105], [0.2208,0.3085])	([0.1316,0.3036], [0.1721,0.5098])	
XA ₃	([0.1109,0.4514], [0.3580,0.5031])	([0.2345,0.5016], [0.1867,0.3084])	
XA_4	([0.3105,0.5026], [0.2157,0.6123])	([0.3031,0.6953], [0.1428,0.3125])	
XA_5	([0.2976,03595], [0.1456,0.2297])	([0.2789,0.6547], [0.1106,0.3572])	
Alternatives	XG ₃	XG ₄	
XA_1	([0.1367,0.3478], [0.3652,0.6194])	([0.2845,0.5956], [0.1767,0.3778])	
XA_2	([0.2846,0.5953], [0.1785,0.5909])	([0.3456,0.7519], [0.1484,0.4136])	
XA ₃	([0.3115,0.6947], [0.1215,0.2835])	([0.1439,0.4601], [0.1812,0.3338])	
XA_4	([0.1165,0.7145], [0.1875,0.2654])	([0.1216,0.3014], [0.1886,0.3872])	
	([0.4126,0.6434], [0.2446,0.3246])	([0.2337,0.3326], [0.2875,0.5894])	

the 20th century, the concept of increasing students' cultural literacy education began to emerge in various universities. The concept of English education abroad truly played a practical role in China's university education, and in the later stage, a systematic concept including concept, form, practice, and system was formed, effectively solving the problems of one-sided education and one-sided talent cultivation in China's education field. In the later development process, some key universities have begun to actively explore the field of basic English courses, seeking suitable content for the current economic development model and social system needs of China, and actively promoting the standardized development of basic English course education. The 13th Five Year Plan officially proposed a new educational and training mechanism that combines English education with traditional educational concepts. Since then, the wave of reform in basic English education in China has officially started, and various modes of English education dissemination have emerged and achieved good development. The importance of basic English curriculum education is evident in the process of cultivating versatile talents. Looking back at the practical process of English education in universities, college students have gradually developed from the initial "What is English education, what is the significance of conducting basic English courses, and how to effectively carry out English education" to a systematic development

TABLE 6. The NXR = $\begin{bmatrix} NXR_{ij} \end{bmatrix}_{5 \times 4}$.

Alternatives	XG_2	XG_2
XA ₁	([0.1524,0.4531], [0.2678,0.5961])	([0.1647,0.3251], [0.2780,0.3806])
XA_2	([01732,0.2105], [0.2208,0.3085])	([0.1316,0.3036], [0.1721,0.5098])
XA ₃	([0.1109,0.4514], [0.3580,0.5031])	([0.2345,0.5016], [0.1867,0.3084])
XA ₄	([0.3105,0.5026], [0.2157,0.6123])	([0.3031,0.6953], [0.1428,0.3125])
XA ₅	([0.2976,03595], [0.1456,0.2297])	([0.2789,0.6547], [0.1106,0.3572])
Alternatives	XG ₃	XG ₄
XA ₁	([0.1367,0.3478], [0.3652,0.6194])	([0.2845,0.5956], [0.1767,0.3778])
XA_2	([0.2846,0.5953], [0.1785,0.5909])	([0.3456,0.7519], [0.1484,0.4136])
XA_3	([0.3115,0.6947], [0.1215,0.2835])	([0.1439,0.4601], [0.1812,0.3338])
XA_4	([0.1165,0.7145], [0.1875,0.2654])	([0.1216,0.3014], [0.1886,0.3872])

TABLE 7. The attribute weight information.

	XG_1	XG_2	XG ₃	XG_4
w	0.2760	0.2959	0.1616	0.2665
lativa wai	aht information			
relative wei	ght information.			
relative wei	-	YC.	YC.	XG4
relative wei	ght information. XG ₁	XG ₂	XG ₃	XG ₄
e relative wei	-	XG ₂	XG ₃	

model that combines ideas with practice, teachers and students work together, and students consciously learn. At present, the reform of basic English curriculum education has made preliminary achievements, but it is still far from enough in terms of the national talent demand and the current situation of talent cultivation. In the future, under the promotion of smart classroom theory, there is still a long way to go for the reform of basic English curriculum education. The smart classroom teaching evaluation of basic English is MAGDM issue. Therefore, the smart classroom teaching evaluation of basic English is presented to demonstrate the approach developed in this paper. There is a panel with five English colleges XA_i (i = 1, 2, 3, 4, 5) to choose. The experts select four attributes to assess the five English colleges:

XG₁ is teachers' teaching ability; $@XG_2$ is teaching cost; $@XG_3$ is teaching content; $@XG_4$ is subject knowledge. The XG₂ is cost attribute and others attributes are beneficial. The five possible English colleges XA_i (i = 1, 2, 3, 4, 5) are to be evaluated with IVPFNs with the four criteria by three experts XE_i (t = 1, 2, 3) (Suppose expert's weight is (0.2595, 0.3897, 0.3508).

The IVPF-ExpTODIM-MACONT technique is used to solve the smart classroom teaching evaluation of basic English.

Step 1. Construct the $XR = \left[XR_{ij}^t\right]_{5\times 4}$ through linguistic scale (See Table 2-4).

Then according to IVPFWA technique, the $XR = [XR_{ij}]_{5\times4}$ is obtained (See Table 5).

TABLE 8.

TABLE 9. The *DDRV* = $\left(DDRV_{ij}\right)_{5\times 4}$.

	XG_1	XG_2	XG ₃	XG ₄
XA ₁	3.4572	-4.3251	1.7211	-3.7494
XA_2	-2.0610	3.4310	-5.9212	2.8724
XA ₃	-2.4566	-0.2111	-0.6050	3.9416
XA ₄	1.4786	0.5687	-1.5765	2.0549
XA ₅	-1.9779	-2.6519	-3.0593	-5.1291

TABLE 10. The *NDDRV*⁽¹⁾_{*ij*}.

	XG ₁	XG_2	XG ₃	XG ₄
XA ₁	1.0000	0.0000	1.0000	0.0000
XA_2	0.0669	1.0000	0.0000	0.9075
XA ₃	0.0000	0.5304	0.7905	1.0000
XA4	0.6859	0.6000	0.6979	0.6766
XA ₅	0.0535	0.3601	0.4229	0.1077

TABLE 11. The NDDRV $_{ij}^{(2)}$.

	XG_1	XG_2	XG ₃	XG_4
XA_1	0.7389	0.0003	0.6827	0.0087
XA_2	0.0181	0.9349	0.0015	0.3435
XA ₃	0.0139	0.0216	0.1890	0.4992
XA_4	0.2120	0.0377	0.1070	0.1350
XA ₅	0.0172	0.0055	0.0198	0.0135

IEEE Access

TABLE	12.	The $NDDRV_{ij}^{(3)}$.
-------	-----	--------------------------

	XG ₁	XG_2	XG ₃	XG_4
XA_1	1.0000	0.0003	1.0000	0.0175
XA_2	0.0245	1.0000	0.0022	0.6881
XA ₃	0.0188	0.0231	0.2768	1.0000
XA_4	0.2868	0.0404	0.1568	0.2704
XA_5	0.0232	0.0059	0.0290	0.0271

TABLE 13. The NDDRV_{ii}.

	XG_1	XG ₂	XG ₃	XG ₄
XA_1	0.9130	0.0002	0.8942	0.0088
XA_2	0.0365	0.9783	0.0012	0.6464
XA ₃	0.0109	0.1917	0.4188	0.8331
XA4	0.3949	0.2260	0.3206	0.3607
XA_5	0.0313	0.1238	0.1572	0.0494

TABLE 15. The FCV.

_

TABLE 14. The $SCV_i^{(1)}$ and $SCV_i^{(2)}$.

	$XQ_i^{(1)}$	$XQ_i^{(2)}$	$SCV_i^{(1)}$	$SCV_i^{(2)}$
XA_1	-0.2364	0.8457	-0.0687	-0.0112
XA_2	-0.0089	0.8516	0.2115	0.0109
XA ₃	0.0800	0.1452	0.1290	0.0185
XA_4	0.3177	1.1815	0.7012	0.0825
XA ₅	-0.0683	0.8203	0.1303	0.0048

Step 2. Normalize the $XR = [XR_{ij}]_{5\times 4}$ into $NXR = [NXR_{ij}]_{5\times 4}$ (See Table 6).

Step 3. Implement the weight values (See Table 7):

Step 4. Produce the relative weight (See Table 8):

Step 5. Implement the $DDRV = (DDRV_{ij})_{5\times 4}$ (See table 9):

Alternative	FCV	Order
XA_1	-0.0994	5
XA_2	0.1690	3
XA ₃	0.1719	2
XA_4	0.8296	1
XA_5	0.0930	4

Step 6. Standardize DDRV matrix through using three normalization techniques(See table 10-12):

Step 7. Integrating three normalization techniques (See table 13):

TABLE 16. Order of these different techniques.

Methods	Order
IVPFWA technique[48]	$XA_4 \succ XA_3 \succ XA_2 \succ XA_5 \succ XA_1$
IVPFWG technique[48]	$XA_4 \succ XA_3 \succ XA_5 \succ XA_2 \succ XA_1$
WIVPFPA technique [56]	$XA_4 \succ XA_3 \succ XA_2 \succ XA_5 \succ XA_1$
WIVPFPG technique [56]	$XA_4 \succ XA_3 \succ XA_5 \succ XA_2 \succ XA_1$
IVPFAAWA technique [55]	$XA_4 \succ XA_3 \succ XA_2 \succ XA_5 \succ XA_1$
IVPF-WASPAS technique [57]	$XA_4 \succ XA_3 \succ XA_2 \succ XA_5 \succ XA_1$
IVPF-QUALIFLEX technique [58]	$XA_4 \succ XA_3 \succ XA_2 \succ XA_5 \succ XA_1$
IVPF-TOPSIS technique [59]	$XA_4 \succ XA_3 \succ XA_2 \succ XA_5 \succ XA_1$
IVPF-TODIM technique [60]	$XA_4 \succ XA_3 \succ XA_2 \succ XA_5 \succ XA_1$
IVPF-ExpTODIM-MACONT technique	$XA_4 \succ XA_3 \succ XA_2 \succ XA_5 \succ XA_1$

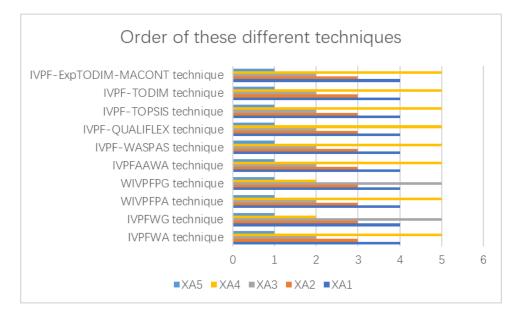


FIGURE 2. Order of these different techniques.

Step 8. Manage the SCV for each alternative (See table 14):

Step 9. Manage the FCV for each alternative (See table 15):

Step 10. Finally, the order is obtained: $XA_4 > XA_3 > XA_2 > XA_5 > XA_1$, and thus the optimal English college is XA_4 .

B. COMPARATIVE ANALYSIS

Then, the IVPF-ExpTODIM-MACONT technique is compared with IVPFWA technique [48], IVPFWG technique [48], interval-valued Pythagorean fuzzy Aczel-Alsina weighted average (IVPFAAWA) technique [55], weighted interval-valued Pythagorean fuzzy power average (WIVPFPA) technique [56], weighted interval-valued Pythagorean fuzzy power geometric (WIVPFPG) technique [56], intervalvalued Pythagorean fuzzy WASPAS (IVPF-WASPAS) technique [57] and interval-valued Pythagorean fuzzy QUALIFLEX (IVPF-QUALIFLEX) technique [58], intervalvalued Pythagorean fuzzy TOPSIS (IVPF-TOPSIS) technique [59] and IVPF-TODIM technique [60]. The comparative decision results are shown in Table 16 and Figure 2.

In accordance with WS coefficients [61], [62], the WS coefficient calculation between IVPFWA technique [48], IVPFWG technique [48], IVPFAAWA technique [55], WIVPFPA technique [56], WIVPFPG technique [56], IVPF-WASPAS technique [57], IVPF-QUALIFLEX technique [58], IVPF-TOPSIS technique [59], IVPF-TODIM technique [60] and the proposed IVPF-ExpTODIM-MACONT technique is 1.0000, 0.7266, 1.0000, 0.7266, 1.0000, 1.0000, 1.0000, 1.0000, 1.0000, respectively. The WS coefficient calculation shows that the order result of IVPFWA technique [48], IVPFAAWA technique [55], WIVPFPA technique [56], IVPF-WASPAS technique [57], IVPF-QUALIFLEX technique [58], IVPF-TOPSIS technique [59] and IVPF-TODIM technique [60] is same with the order result of the proposed IVPF-ExpTODIM-MACONT technique. The WS coefficient calculation shows that the order result of IVPFWG technique [48] and WIVPFPG technique [56] is slight different from the order result of the proposed IVPF-ExpTODIM-MACONT technique. However, these techniques have the same optimal English college and worst English college. This shows the IVPF-ExpTODIM-MACONT technique is reasonable and effective.

V. CONCLUSION

The smart classroom teaching evaluation of basic English (IP) perspective is the MAGDM. Recently, the Exponential TODIM (ExpTODIM) technique and MACONT technique has been used to cope with MAGDM issues. The IVPFSs are used as a technique for characterizing uncertain information during the smart classroom teaching evaluation of basic English. In this manuscript, the IVPF-ExpTODIM-MACONT technique is built to solve the MAGDM under IVPFSs. In the end, a numerical case study for smart classroom teaching evaluation of basic English is implemented to validate the proposed technique. The main contributions of this paper are outlined: (1) the ExpTODIM and MACONT technique has been extended to IVPFSs; (2) Information Entropy technique is employed

to manage the weight values under IVPFSs. (3) the IVPF-ExpTODIM-MACONT technique is founded to implement the MAGDM under IVPFSs; (4) a numerical case study for smart classroom teaching evaluation of basic English and some comparative analysis is supplied to verify the IVPF-ExpTODIM-MACONT technique.

There may be some possible limitations of this work, which could be further managed the smart classroom teaching evaluation of basic English: (1) It is a worthwhile research work to manage consensus [63], [64], [65] to smart classroom teaching evaluation of basic English under IVPFSs; (2) It is also worthwhile research work to manage regret theory to the smart classroom teaching evaluation of basic English under IVPFSs [66], [67], [68]; (3) In subsequent research works, the integration of TODIM techniques with other decision-making techniques employing different fuzzy extensions could be managed for smart classroom teaching evaluation of basic English [69], [70].

REFERENCES

- J. Aguilar, J. Altamiranda, and F. Díaz, "Specification of a managing agent of emergent serious games for a smart classroom," *IEEE Latin Amer. Trans.*, vol. 18, no. 1, pp. 51–58, Jan. 2020.
- [2] J. Aguilar, J. Cordero, and O. Buendía, "Specification of the autonomic cycles of learning analytic tasks for a smart classroom," J. Educ. Comput. Res., vol. 56, no. 6, pp. 866–891, Oct. 2018.
- [3] A. M. Alfoudari, C. M. Durugbo, and F. M. Aldhmour, "Understanding socio-technological challenges of smart classrooms using a systematic review," *Comput. Educ.*, vol. 173, Nov. 2021, Art. no. 104282.
- [4] M. Bhatia and A. Kaur, "Quantum computing inspired framework of student performance assessment in smart classroom," *Trans. Emerg. Telecommun. Technol.*, vol. 32, no. 9, p. 22, Sep. 2021.
- [5] J. P. Bishop, "'She's always been the smart one. I've always been the dumb one': Identities in the mathematics classroom," *J. Res. Math. Educ.*, vol. 43, no. 1, pp. 34–74, Jan. 2012.
- [6] M. Burunkaya and K. Duraklar, "Design and implementation of an IoTbased smart classroom incubator," *Appl. Sci.*, vol. 12, no. 4, p. 2233, Feb. 2022.
- [7] G. Cebrián, R. Palau, and J. Mogas, "The smart classroom as a means to the development of ESD methodologies," *Sustainability*, vol. 12, no. 7, p. 3010, Apr. 2020.
- [8] L. Chamba and J. Aguilar, "Design of an augmented reality component from the theory of agents for smart classrooms," *IEEE Latin Amer. Trans.*, vol. 14, no. 8, pp. 3826–3837, Aug. 2016.
- [9] M. Corella, "Talking 'smart': Academic language and indexical competence in peer interactions in an elementary classroom," *Linguistics Educ.*, vol. 55, Feb. 2020, Art. no. 100755.
- [10] Y. Dai, Z. Liu, and D. Luo, "Research on the interactive psychological adaptability of teachers and students in smart classroom," *Basic Clin. Pharmacol. Toxicol.*, vol. 127, pp. 62–63, Nov. 2020.
- [11] S. Fakhar, J. Baber, S. U. Bazai, S. Marjan, M. Jasinski, E. Jasinska, M. U. Chaudhry, Z. Leonowicz, and S. Hussain, "Smart classroom monitoring using novel real-time facial expression recognition system," *Appl. Sci.*, vol. 12, no. 23, p. 12134, Nov. 2022.
- [12] N. Gligoric, A. Uzelac, S. Krco, I. Kovacevic, and A. Nikodijevic, "Smart classroom system for detecting level of interest a lecture creates in a classroom," *J. Ambient Intell. Smart Environments*, vol. 7, no. 2, pp. 271–284, 2015.
- [13] A. Gómez, L. A. Chamba Eras, and J. Aguilar, "Multi-agent systems for the management of resources and activities in a smart classroom," *IEEE Latin Amer. Trans.*, vol. 19, no. 9, pp. 1511–1519, Sep. 2021.
- [14] S. K. Gupta, T. S. Ashwin, and R. M. R. Guddeti, "Students' affective content analysis in smart classroom environment using deep learning techniques," *Multimedia Tools Appl.*, vol. 78, no. 18, pp. 25321–25348, Sep. 2019.

- [15] M. S. Haghighi, A. Sheikhjafari, A. Jolfaei, F. Farivar, and S. Ahmadzadeh, "Automation of recording in smart classrooms via deep learning and Bayesian maximum a posteriori estimation of instructor's pose," *IEEE Trans. Ind. Informat.*, vol. 17, no. 4, pp. 2813–2820, Apr. 2021.
- [16] L.-S. Huang, J.-Y. Su, and T.-L. Pao, "A context aware smart classroom architecture for smart campuses," *Appl. Sci.*, vol. 9, no. 9, p. 1837, May 2019.
- [17] P. W. Kim, "Ambient intelligence in a smart classroom for assessing students' engagement levels," J. Ambient Intell. Humanized Comput., vol. 10, no. 10, pp. 3847–3852, Oct. 2019.
- [18] Z. Zhan, Q. Wu, Z. Lin, and J. Cai, "Smart classroom environments affect teacher-student interaction: Evidence from a behavioural sequence analysis," *Australas. J. Educ. Technol.*, vol. 37, no. 2, pp. 96–109, May 2021.
- [19] X. Zhang and L. Chen, "College English smart classroom teaching model based on artificial intelligence technology in mobile information systems," *Mobile Inf. Syst.*, vol. 2021, pp. 1–12, Oct. 2021.
- [20] M. Kwet and P. Prinsloo, "The 'smart'classroom: A new frontier in the age of the smart university," *Teaching Higher Educ.*, vol. 25, no. 4, pp. 510–526, May 2020.
- [21] Y. Li, H. H. Yang, and J. MacLeod, "Preferences toward the constructivist smart classroom learning environment: Examining pre-service teachers' connectedness," *Interact. Learn. Environ.*, vol. 27, no. 3, pp. 349–362, Apr. 2019.
- [22] H. Liao, R. Tan, and M. Tang, "An overlap graph model for largescale group decision making with social trust information considering the multiple roles of experts," *Exp. Syst.*, vol. 38, no. 3, p. 20, May 2021.
- [23] M. Tang, H. Liao, and G. Kou, "Type a and type γ consensus for multi-stage emergency group decision making based on mining consensus sequences," J. Oper. Res. Soc., vol. 73, no. 2, pp. 365–381, Feb. 2022.
- [24] M. Tang, H. Liao, J. Xu, D. Streimikiene, and X. Zheng, "Adaptive consensus reaching process with hybrid strategies for large-scale group decision making," *Eur. J. Oper. Res.*, vol. 298, pp. 957–971, May 2022.
- [25] M. Akram, C. Kahraman, and K. Zahid, "Extension of TOPSIS model to the decision-making under complex spherical fuzzy information," *Soft Comput.*, vol. 25, no. 16, pp. 10771–10795, Aug. 2021.
- [26] B. Oztaysi, S. C. Onar, and C. Kahraman, "Waste disposal location selection by using Pythagorean fuzzy REGIME method," J. Intell. Fuzzy Syst., vol. 42, no. 1, pp. 401–410, Dec. 2021.
- [27] L. A. Zadeh, "Fuzzy sets," Inf. Control, vol. 8, no. 3, pp. 338–353, 1965.
- [28] K. T. Atanassov, "More on intuitionistic fuzzy sets," Fuzzy Sets Syst., vol. 33, no. 1, pp. 37–45, Oct. 1989.
- [29] R. R. Yager, "Pythagorean membership grades in multicriteria decision making," *IEEE Trans. Fuzzy Syst.*, vol. 22, no. 4, pp. 958–965, Aug. 2014.
- [30] P. Rani, A. R. Mishra, R. Krishankumar, K. S. Ravichandran, and A. H. Gandomi, "A new Pythagorean fuzzy based decision framework for assessing healthcare waste treatment," *IEEE Trans. Eng. Manag.*, vol. 69, no. 6, pp. 2915–2929, Dec. 2022.
- [31] R. Muhammad Zulqarnain, I. Siddique, R. Ali, J. Awrejcewicz, H. Karamti, D. Grzelczyk, A. Iampan, and M. Asif, "Einstein ordered weighted aggregation operators for Pythagorean fuzzy hypersoft set with its application to solve MCDM problem," *IEEE Access*, vol. 10, pp. 95294–95320, 2022.
- [32] P. A. Ejegwa, S. Wen, Y. Feng, W. Zhang, and J. Liu, "A three-way Pythagorean fuzzy correlation coefficient approach and its applications in deciding some real-life problems," *Int. J. Speech Technol.*, vol. 53, no. 1, pp. 226–237, Jan. 2023.
- [33] A. Kehagias and M. Konstantinidou, "L-fuzzy valued inclusion measure, L-fuzzy similarity and L-fuzzy distance," *Fuzzy Sets Syst.*, vol. 136, no. 3, pp. 313–332, Jun. 2003.
- [34] Q. Zhang, J. Ma, Z.-P. Fan, and W.-C. Chiang, "A statistical approach to multiple-attribute decision-making with interval numbers," *Int. J. Syst. Sci.*, vol. 34, nos. 12–13, pp. 683–692, Oct. 2003.
- [35] W. Liang, X. Zhang, and M. Liu, "The maximizing deviation method based on interval-valued Pythagorean fuzzy weighted aggregating operator for multiple criteria group decision analysis," *Discrete Dyn. Nature Soc.*, vol. 2015, pp. 1–15, Aug. 2015.
- [36] E. Haktanir and C. Kahraman, "A novel interval-valued Pythagorean fuzzy QFD method and its application to solar photovoltaic technology development," *Comput. Ind. Eng.*, vol. 132, pp. 361–372, Jun. 2019.

- [37] Y. Tang and Y. Yang, "Sustainable e-bike sharing recycling supplier selection: An interval-valued Pythagorean fuzzy MAGDM method based on preference information technology," *J. Cleaner Prod.*, vol. 287, Mar. 2021, Art. no. 125530.
- [38] X. Fu, T. Ouyang, Z. Yang, and S. Liu, "A product ranking method combining the features-opinion pairs mining and interval-valued Pythagorean fuzzy sets," *Appl. Soft Comput.*, vol. 97, Dec. 2020, Art. no. 106803.
- [39] M. Alrasheedi, A. Mardani, A. R. Mishra, P. Rani, and N. Loganathan, "An extended framework to evaluate sustainable suppliers in manufacturing companies using a new Pythagorean fuzzy entropy-SWARA-WASPAS decision-making approach," *J. Enterprise Inf. Manag.*, vol. 35, no. 2, pp. 333–357, Mar. 2022.
- [40] V. Mohagheghi and S. M. Mousavi, "A new framework for hightechnology project evaluation and project portfolio selection based on Pythagorean fuzzy WASPAS, MOORA and mathematical modeling," *Iran. J. Fuzzy Syst.*, vol. 16, pp. 89–106, Dec. 2019.
- [41] G. Bakioglu and A. O. Atahan, "AHP integrated TOPSIS and VIKOR methods with Pythagorean fuzzy sets to prioritize risks in self-driving vehicles," *Appl. Soft Comput.*, vol. 99, Feb. 2021, Art. no. 106948.
- [42] A. B. Leoneti and L. F. A. M. Gomes, "A novel version of the TODIM method based on the exponential model of prospect theory: The ExpTODIM method," *Eur. J. Oper. Res.*, vol. 295, no. 3, pp. 1042–1055, Dec. 2021.
- [43] H. Sun, Z. Yang, Q. Cai, G. Wei, and Z. Mo, "An extended exp-TODIM method for multiple attribute decision making based on the Z-Wasserstein distance," *Exp. Syst. Appl.*, vol. 214, Mar. 2023, Art. no. 119114.
- [44] Z. Wen, H. Liao, and E. K. Zavadskas, "MACONT: Mixed aggregation by comprehensive normalization technique for multi-criteria analysis," *Informatica*, vol. 31, pp. 1–24, Jan. 2020.
- [45] Z. Wen and H. Liao, "Pension service institution selection by a personalized quantifier-based macont method," *Int. J. Strategic Property Manag.*, vol. 25, no. 6, pp. 446–458, Sep. 2021.
- [46] Z. Wen and H. Liao, "PL-MACONT-I: A probabilistic linguistic MACONT-I method for multi-criterion sorting," Int. J. Inf. Technol. Decis. Making, vol. 20, pp. 1–20, Feb. 2023.
- [47] R. R. Yager and A. M. Abbasov, "Pythagorean membership grades, complex numbers, and decision making," *Int. J. Intell. Syst.*, vol. 28, no. 5, pp. 436–452, May 2013.
- [48] H. Garg, "A novel accuracy function under interval-valued Pythagorean fuzzy environment for solving multicriteria decision making problem," *J. Intell. Fuzzy Syst.*, vol. 31, no. 1, pp. 529–540, Jun. 2016.
- [49] R. Tan, W. Zhang, and S. Chen, "Decision-making method based on grey relation analysis and trapezoidal fuzzy neutrosophic numbers under double incomplete information and its application in typhoon disaster assessment," *IEEE Access*, vol. 8, pp. 3606–3628, 2020.
- [50] J. H. Kim and B. S. Ahn, "The hierarchical VIKOR method with incomplete information: Supplier selection problem," *Sustainability*, vol. 12, no. 22, p. 9602, Nov. 2020.
- [51] P. Liu and W. Liu, "Multiple-attribute group decision-making method of linguisticq-rung orthopair fuzzy power Muirhead mean operators based on entropy weight," *Int. J. Intell. Syst.*, vol. 34, no. 8, pp. 1755–1794, Aug. 2019.
- [52] C. E. Shannon, "A mathematical theory of communication," *Bell Syst. Tech. J.*, vol. 27, no. 3, pp. 379–423, Jul. 1948.
- [53] D. Kahneman and A. Tversky, "Prospect theory: An analysis of decision under risk," *Econometrica*, vol. 47, no. 2, pp. 263–291, Mar. 1979.
- [54] J. Wang, "Extended TODIM method based on VIKOR for quality evaluation of higher education scientific research management under interval-valued Pythagorean fuzzy sets," J. Intell. Fuzzy Syst., vol. 45, no. 4, pp. 5277–5289, Oct. 2023.
- [55] T. Senapati, A. R. Mishra, A. Saha, V. Simic, P. Rani, and R. Ali, "Construction of interval-valued Pythagorean fuzzy Aczel–Alsina aggregation operators for decision making: A case study in emerging IT software company selection," *Sadhana*, vol. 47, no. 4, p. 18, Nov. 2022.
- [56] D. Liang, A. P. Darko, and J. Zeng, "Interval-valued Pythagorean fuzzy power average-based MULTIMOORA method for multi-criteria decisionmaking," *J. Experim. Theor. Artif. Intell.*, vol. 32, no. 5, pp. 845–874, Sep. 2020.
- [57] A. Al-Barakati, A. R. Mishra, A. Mardani, and P. Rani, "An extended interval-valued Pythagorean fuzzy WASPAS method based on new similarity measures to evaluate the renewable energy sources," *Appl. Soft Comput.*, vol. 120, May 2022, Art. no. 108689.

- [58] X. Zhang, "Multicriteria Pythagorean fuzzy decision analysis: A hierarchical QUALIFLEX approach with the closeness index-based ranking methods," *Inf. Sci.*, vol. 330, pp. 104–124, Feb. 2016.
- [59] H. Garg, "A new improved score function of an interval-valued Pythagorean fuzzy set based topsis methoD," *Int. J. Uncertainty Quantification*, vol. 7, no. 5, pp. 463–474, 2017.
- [60] M. Zhao, G. Wei, C. Wei, and J. Wu, "TODIM method for interval-valued Pythagorean fuzzy MAGDM based on cumulative prospect theory and its application to green supplier selection," *Arabian J. Sci. Eng.*, vol. 46, no. 2, pp. 1899–1910, Feb. 2021.
- [61] W. Salabun and K. Urbaniak, "A new coefficient of rankings similarity in decision-making problems," in *Proc. 20th Annu. Int. Conf. Comput. Sci.* (*ICCS*). Cham, Switzerland: Springer, 2020, pp. 632–645.
- [62] W. Salabun, J. Watróbski, and A. Shekhovtsov, "Are MCDA methods benchmarkable? A comparative study of TOPSIS, VIKOR, COPRAS, and PROMETHEE II methods," *Symmetry*, vol. 12, no. 9, p. 1549, Sep. 2020.
- [63] P. Wu, F. Li, J. Zhao, L. Zhou, and L. Martínez, "Consensus reaching process with multiobjective optimization for large-scale group decision making with cooperative game," *IEEE Trans. Fuzzy Syst.*, vol. 31, no. 1, pp. 293–306, Jan. 2023.
- [64] X. Xu, Z. Gong, E. Herrera-Viedma, G. Kou, and F. J. Cabrerizo, "Consensus reaching in group decision making with linear uncertain preferences and asymmetric costs," *IEEE Trans. Syst., Man, Cybern., Syst.*, vol. 53, no. 5, pp. 2887–2899, May 2023.
- [65] H. Zhang and Y. Dai, "Consensus improvement model in group decision making with hesitant fuzzy linguistic term sets or hesitant fuzzy linguistic preference relations," *Comput. Ind. Eng.*, vol. 178, Apr. 2023, Art. no. 109015.
- [66] Y. Lin, Y.-M. Wang, and S.-Q. Chen, "Hesitant fuzzy multiattribute matching decision making based on regret theory with uncertain weights," *Int. J. Fuzzy Syst.*, vol. 19, no. 4, pp. 955–966, Aug. 2017.
- [67] X. Jia, X. Wang, Y. Zhu, L. Zhou, and H. Zhou, "A two-sided matching decision-making approach based on regret theory under intuitionistic fuzzy environment," *J. Intell. Fuzzy Syst.*, vol. 40, no. 6, pp. 11491–11508, Jun. 2021.

- [68] X. Tian, Z. Xu, J. Gu, and F. Herrera, "A consensus process based on regret theory with probabilistic linguistic term sets and its application in venture capital," *Inf. Sci.*, vol. 562, pp. 347–369, Jul. 2021.
- [69] X. Peng, J. Dai, and F. Smarandache, "Research on the assessment of project-driven immersion teaching in extreme programming with neutrosophic linguistic information," *Int. J. Mach. Learn. Cybern.*, vol. 14, no. 3, pp. 873–888, Mar. 2023.
- [70] X. Peng and J. Dai, "Research on the assessment of classroom teaching quality with q-rung orthopair fuzzy information based on multiparametric similarity measure and combinative distance-based assessment," *Int. J. Intell. Syst.*, vol. 34, no. 7, pp. 1588–1630, Jul. 2019.



YAMIN LIANG was born in Xianyang, Shaanxi, China, in 1978. She received the master's degree from the Shandong University of Science and Technology, China. She is currently a Teacher with the School of Foreign Languages, Xianyang Normal University. Her research interests include English teaching and translation.

...