



## Group decision-making based on distance measures settings for single-valued neutrosophic fuzzy soft expert environment

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### Abstract

A soft expert set is a concept that combines elements of soft sets and expert systems. It aims to incorporate expert knowledge and uncertainty-handling capabilities into the analysis and decision-making processes. On the other hand, the idea of single neutrosophic sets (SVNSs) and fuzzy sets (FSs) are imported models for handling the uncertainty data. In this work, the authors combine the critical features of FSs and SVNSs under expert systems in one model. Accordingly, this model worked to provide decision-makers with more flexibility in the process of interpreting uncertain information. From a scientific point of view, the process of evaluating this high-performance SVNFSSES disappears. Therefore, in this paper, we initiated a new approach known as single-valued neutrosophic fuzzy soft expert sets (SVNFSESS) as a new development in a fuzzy soft computing environment. We investigate some fundamental operations on SVNFSSESS along with their basic properties. Also, we investigate AND and OR operations between two SVNFSSESS as well as several numerical examples to clarify the above fundamental operations. Finally, we have given distance measures (DM) between two SVNFSSESS to construct a new algorithm that is used to demonstrate the effectiveness of the method in handling some real-life applications.

**Keywords:** Neutrosophic sets; Neutrosophic soft sets; Single-valued neutrosophic soft sets; Expert soft set; Optimization; Decision Making

### 1 Introduction

Neutrosophic set theory<sup>1</sup> is an extension of classical set theory that aims to handle uncertainty, vagueness, and indeterminacy in a more comprehensive way. It was introduced by Florentin Smarandache in 1995 and has since found applications in various fields, including decision-making, expert systems, image processing, and artificial intelligence. A neutrosophic set is characterized by three components: the membership function, the

non-membership function, and the indeterminacy function. These functions assign degrees of truth, falsity, and indeterminacy, respectively, to each element in the set. Unlike classical sets, which assign elements either a membership degree of 1 or 0, neutrosophic sets allow for partial membership, non-membership, or indeterminacy. The membership function represents the degree to which an element belongs to the set. It ranges between 0 and 1, where 0 indicates no membership, 1 indicates full membership, and values in between represent degrees of partial membership. The non-membership function represents the degree to which an element does not belong to the set. It also ranges between 0 and 1, where 0 indicates no non-membership, 1 indicates full non-membership, and values in between represent degrees of partial non-membership. The indeterminacy function represents the degree of indeterminacy or uncertainty associated with an element's membership or non-membership. It ranges between 0 and 1, where 0 represents complete determinacy, 1 represents complete indeterminacy, and values in between represent degrees of partial indeterminacy. Neutrosophic sets provide a more nuanced and flexible approach to handling uncertainty and imprecision in various domains. They can capture situations where the degree of membership or non-membership is uncertain or ambiguous, allowing for more realistic modeling of complex and uncertain systems. Neutrosophic set theory has been applied in decision-making processes to handle conflicting and uncertain information. Neutrosophic logic and inference have been developed to reason with neutrosophic sets and make decisions based on the degrees of truth, falsity, and indeterminacy. Moreover, neutrosophic fuzzy sets combine neutrosophic sets and fuzzy sets,<sup>2</sup> incorporating degrees of truth, falsity, and indeterminacy into fuzzy logic systems. This hybrid approach enhances the modeling of uncertainty, vagueness, and ambiguity in decision-making processes. Overall, neutrosophic set theory offers a valuable framework for dealing with uncertainty and indeterminacy, providing a more comprehensive representation and analysis of complex systems and decision problems. Its applications continue to evolve, and researchers are exploring new extensions and variations to address specific challenges in different domains.

Soft set<sup>3</sup> theory is a mathematical framework that provides a flexible and intuitive way to handle uncertainty, vagueness, and imprecision in decision-making and data analysis. It was introduced by Molodtsov in 1999 as a generalization of classical set theory. In soft set theory, a soft set is defined as a collection of objects with a characteristic function that assigns degrees of membership to each object in a set. Unlike classical sets, where an object is either a member or non-member, soft sets allow for partial membership or degrees of uncertainty associated with membership. SSS have received wide attention from researchers around the world and they have introduced a lot of works for example, neutrosophic soft set(NSS),<sup>4,5</sup> interval-NSS,<sup>6,7</sup> complex interval-NSS,<sup>8</sup> weighted Similarity Measure on neutrosophic soft set<sup>9,10</sup> and a lot of models employed in solving real-life applications see.<sup>11-16</sup> In other hande, Soft Expert Set<sup>17-19</sup> is a concept that combines elements of soft sets and expert systems. It aims to incorporate expert knowledge and uncertainty handling capabilities into the analysis and decision-making processes.

In traditional expert systems, human expertise is captured and represented using rules or knowledge bases. These systems rely on the knowledge and experience of domain experts to make informed decisions. However, they may not adequately handle uncertainties or imprecise information. With this design, many research works emerged that combine NS and SES like neutrosophic soft expert sets (NSESs),<sup>20</sup> interval-NSESs,<sup>21</sup> Generalized neutrosophic soft expert set,<sup>22</sup> neutrosophic soft expert graphs,<sup>23</sup> possibility neutrosophic soft expert sets<sup>24,25</sup> and a lot of models employed in solving real-life applications see.<sup>26-31</sup> In this work, we will collect a number of properties present in each of fuzzy set, neutrosophic set, and soft expert set under a single value in one model called Single value neutrosophic fuzzy soft expert set (SVNFSES). Based on this model, we present several distance measures and explain the mechanism for their use in solving a decision-making problem. This article is organized as follows: We review some basic definitions of the associated studies in Section 2. Section 3 presents the formulation of the SVNFSE-set and its operations. While in section 4, we demonstrate the set-theoretic operations of SVNFSE-sets together with some propositions and examples. Section 5 discusses the applications of the SVNFSE-set in decision-making problems based on Distance Measure on SVNFSE-sets. Finally, conclusions and suggestions for further studies are pointed out in section 6.

## 2 Preliminaries

In this section, we recapitulate some of the ideas like FS, NS, SVNS, SS, and SVNSS that are considered beneficial in developing our new concept.

**Definition 2.1.** <sup>2</sup> Assume that  $\hat{U} = \{u_1, u_2, u_3, \dots, u_n\}$  be a reference set. Then the FS formed as following structure:

$$Q = \left\{ \left( u, \left\langle \check{\partial}_Q^t(u_i) \right\rangle \right) \mid u \in \hat{U} \right\}$$

where  $\check{\partial}_Q^t(u_i)$  refer to true membership of object  $u_i$  in  $\hat{U}$  and persistent as a mapping:  $\check{\partial}_Q^t : \hat{U} \rightarrow [0, 1]$ .

**Definition 2.2.** <sup>2</sup> Let  $Q_1 = \left\{ \left( u, \left\langle \check{\partial}_{Q_1}^t(u_i) \right\rangle \right) \mid u \in \hat{U} \right\}$  and  $Q_2 = \left\{ \left( \tau, \left\langle \check{\partial}_{Q_2}^t(u_i) \right\rangle \right) \mid u \in \hat{U} \right\}$  be two FS on reference set  $\hat{U}$ . Then the fundamental operation on FSs defined as following:

**1.Union**  $Q_3 = \left\{ \left( u, \left\langle \max \left( \check{\partial}_{Q_1}^t(u_i), \check{\partial}_{Q_2}^t(u_i) \right) \right\rangle \right) \mid u \in \hat{T} \right\}$ .

**2.Intersection**  $Q_3 = \left\{ \left( \tau, \left\langle \min \left( \check{\partial}_{Q_1}^t(u_i), \check{\partial}_{Q_2}^t(u_i) \right) \right\rangle \right) \mid \partial \in \hat{U} \right\}$ .

**3.Complement**  $Q_1^c = \left\{ \left( u, \left\langle \left( 1 - \check{\partial}_{Q_1}^t(u_i) \right) \right\rangle \right) \mid u \in \hat{U} \right\}$ .

**4.Subset**  $Q_1 \subseteq Q_2$  if  $\check{\partial}_{Q_1}^t(u_i) \leq \check{\partial}_{Q_2}^t(u_i)$ .

**Definition 2.3.** <sup>1</sup> Assume that  $\hat{U} = \{u_1, u_2, \tau_3, \dots, u_n\}$  be a reference set  $\hat{U}$ . Then the NS formed as following structure:

$$\hat{A}_{NS} = \left\{ \left( u, \left\langle \check{\partial}_{\hat{A}}^t(u_i), \check{\partial}_{\hat{A}}^i(u_i), \check{\partial}_{\hat{A}}^f(u_i) \right\rangle \right) \mid u \in \hat{U} \right\}$$

where  $\check{\partial}_{\hat{A}}^t(u_i), \check{\partial}_{\hat{A}}^i(u_i), \check{\partial}_{\hat{A}}^f(u_i)$  refer to true membership, indeterminacy membership and falsehood membership of object  $u_i$  in  $\hat{U}$  and persistent as a mapping:  $\check{\partial}_{\hat{A}}^t(u_i), \check{\partial}_{\hat{A}}^i(u_i), \check{\partial}_{\hat{A}}^f(u_i) : \hat{U} \rightarrow [0, 1]$ .

**Definition 2.4.** <sup>24</sup> Let  $\hat{A}_{SVNS} = \left\{ \left( u, \left\langle \check{\partial}_{\hat{A}}^t(u_i), \check{\partial}_{\hat{A}}^i(u_i), \check{\partial}_{\hat{A}}^f(u_i) \right\rangle \right) \mid u \in \hat{U} \right\}$  and  $\hat{B}_{SVNS} = \left\{ \left( u, \left\langle \check{\partial}_{\hat{B}}^t(u_i), \check{\partial}_{\hat{B}}^i(u_i), \check{\partial}_{\hat{B}}^f(u_i) \right\rangle \right) \mid u \in \hat{U} \right\}$  be two SVNS on reference set  $\hat{U}$  Then the fundamental operation on SVNS defined as following:

**1.Union**  
 $\hat{C}_{SVNS} = \left\{ \left( u, \left\langle \max \left[ \check{\partial}_{\hat{A}}^t(u_i), \check{\partial}_{\hat{B}}^t(u_i) \right], \min \left[ \check{\partial}_{\hat{A}}^i(u_i), \check{\partial}_{\hat{B}}^i(u_i) \right], \min \left[ \check{\partial}_{\hat{A}}^f(u_i), \check{\partial}_{\hat{B}}^f(u_i) \right] \right\rangle \right) \mid u \in U \right\}$ .

**2.Intersection**  
 $\hat{C}_{SVNS} = \left\{ \left( u, \left\langle \min \left[ \check{\partial}_{\hat{A}}^t(u_i), \check{\partial}_{\hat{B}}^t(u_i) \right], \max \left[ \check{\partial}_{\hat{A}}^i(u_i), \check{\partial}_{\hat{B}}^i(u_i) \right], \max \left[ \check{\partial}_{\hat{A}}^f(u_i), \check{\partial}_{\hat{B}}^f(u_i) \right] \right\rangle \right) \mid u \in U \right\}$ .

**3.Complement**  
 $\hat{A}_{SVNS}^c = \left\{ \left( u, \left\langle \left[ \check{\partial}_{\hat{A}}^f(u_i) \right], \left[ 1 - \check{\partial}_{\hat{A}}^i(u_i) \right], \left[ \check{\partial}_{\hat{A}}^t(u_i) \right] \right\rangle \right) \mid u \in U \right\}$ .

**4.Subset**  
 $\hat{A}_{SVNS} \subseteq \hat{B}_{SVNS}$  if  $\check{\partial}_{\hat{A}}^t(u_i) \leq \check{\partial}_{\hat{B}}^t(u_i), \check{\partial}_{\hat{A}}^i(u_i) \geq \check{\partial}_{\hat{B}}^i(u_i), \check{\partial}_{\hat{A}}^f(u_i) \geq \check{\partial}_{\hat{B}}^f(u_i)$ .

**Definition 2.5.** <sup>24</sup> Let  $\hat{U} = \{u_1, u_2, u_3, \dots, u_n\}$  and  $\hat{E} = \{z_1, z_2, z_3, \dots, z_m\}$  be a reference set and attribute set, respectively. Then a SS over  $\hat{U}$  given as structures as follows:

$$\vec{S} = \left\{ \left( z, \left\langle \vec{S}(z_i) \right\rangle \right) \mid z \in \hat{E} \right\}$$

where the function  $\overset{\leftrightarrow}{S}$  given by following mapping:

$$\overset{\leftrightarrow}{S} = E \rightarrow P(U)$$

Here  $P(U)$  refer to collection of subsets of reference set  $\hat{U}$ .

**Definition 2.6.** <sup>24</sup> A term  $\mathcal{P}^{svnss}$  is said to be SVNSS on soft reference set  $(\hat{U}, \hat{E})$ , where  $\mathcal{P}^{svnss} : \hat{E} \rightarrow SVN(P)^{(U)}$ , such that  $SVN(P)^{(U)}$  is a collection of all SVNS-subsets over  $\hat{U}$ .

### 3 Structuring the concept of single value neutrosophic fuzzy soft expert set (SVNFSESS)

In this section, we introduce the concept of a SVNFES and define some properties of this model, namely, the null of the SVNFES, the absolute of the SVNFES, a subset of the SVNFES, and the equality of the SVNFES. Illustrated examples are also given.

**Definition 3.1.** The ordered pair  $(H, O)$  pointing to single-neutrosophic soft expert set (SNFSESS) on  $U$ . If

1. The mapping  $H : O \rightarrow SVNFN^U$  where  $O \subseteq Z = M \times N \times Y$ , such that for all  $z \in Z$  then  $z = (m \times n \times y = 0 \text{ or } 1)$

2. Here  $U = \{u_1, u_2, u_3, \dots, u_s\}$ ,  $M = \{m_1, m_2, m_3, \dots, m_s\}$ ,  $N = \{n_1, n_2, n_3, \dots, n_s\}$  represent to reference set, attribute set, set of experts respectively and  $Y = \{0, 1\}$ .

3. A single neutrosophic soft set (PIVNSS)  $\mathcal{H}$  on  $\hat{U}$  has the structures as follows:

$$\mathcal{H}^{svnfses} = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z \in \hat{Z} \right) \right\}$$

Where  $\ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j)$  represent of the three SVNFSESS memberships as a single real number and  $\mathcal{P}^{svnss}$  refer to SVNFSESS degree of element  $u_i \in \hat{U}$  to  $\mathcal{H}^{svnss}$  who also can denoted by  $\mathcal{H}_O = (H, O \subseteq Z)$

**Example 3.2.** Consider that a tourism company would like to evaluate a group of hotels it owns to see who is suitable. This evaluation was based on two experts working for the company. Now we assume that  $\hat{U}$  includes three hotels  $\{u_1, u_2, u_3\}$ , and the object is evaluated by two experts  $N = \{n_1, n_2\}$ , and for the criteria that were adopted in this evaluation process, it can be represented by  $M = \{m_1, m_2, m_3\}$  such that  $m_1 = \text{Food services}, m_2 = \text{Staff}, m_3 = \text{Number of rooms}$ . Now for  $O \subseteq Z = M \times N \times Y$ , Now our concept presents the opinions of the two experts as follows:

$$\begin{aligned} \mathcal{H}(z_1 = (m_1, n_1, 1)) &= \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.3)}, \left( \frac{u_2}{(0.3, 0.6, 0.7)}, \left( \frac{u_3}{(0.8, 0.1, 0.6)} \right) \right) \right\} \cdot \\ \mathcal{H}(z_2 = (m_1, n_2, 1)) &= \left\{ \left( \frac{u_1}{(0.5, 0.4, 0.2)}, \left( \frac{u_2}{(0.1, 0.5, 0.7)}, \left( \frac{u_3}{(0.2, 0, 0.8)} \right) \right) \right\} \cdot \\ \mathcal{H}(z_3 = (m_2, n_1, 1)) &= \left\{ \left( \frac{u_1}{(0.7, 0.6, 0.2)}, \left( \frac{u_2}{(0.6, 0.3, 0.1)}, \left( \frac{u_3}{(0.2, 0.3, 0.5)} \right) \right) \right\} \cdot \\ \mathcal{H}(z_4 = (m_2, n_2, 1)) &= \left\{ \left( \frac{u_1}{(0.5, 0.3, 0.2)}, \left( \frac{u_2}{(0.6, 0.4, 0.7)}, \left( \frac{u_3}{(0.5, 0.4, 0.3)} \right) \right) \right\} \cdot \\ \mathcal{H}(z_5 = (m_3, n_1, 1)) &= \left\{ \left( \frac{u_1}{(0.7, 0.5, 0.9)}, \left( \frac{u_2}{(0.2, 0.6, 0.7)}, \left( \frac{u_3}{(0.3, 0.4, 0.6)} \right) \right) \right\} \cdot \\ \mathcal{H}(z_6 = (m_3, n_2, 1)) &= \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.3)}, \left( \frac{u_2}{(0.3, 0.6, 0.7)}, \left( \frac{u_3}{(0.8, 0.1, 0.6)} \right) \right) \right\} \cdot \\ \mathcal{H}(z_7 = (m_1, n_1, 0)) &= \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.3)}, \left( \frac{u_2}{(0.3, 0.6, 0.7)}, \left( \frac{u_3}{(0.8, 0.1, 0.6)} \right) \right) \right\} \cdot \\ \mathcal{H}(z_8 = (m_1, n_2, 0)) &= \left\{ \left( \frac{u_1}{(0.4, 0.6, 0.2)}, \left( \frac{u_2}{(0.7, 0.2, 0.5)}, \left( \frac{u_3}{(0.1, 0.1, 0.3)} \right) \right) \right\} \cdot \\ \mathcal{H}(z_9 = (m_2, n_1, 0)) &= \left\{ \left( \frac{u_1}{(0.4, 0.5, 0.8)}, \left( \frac{u_2}{(0.9, 0.8, 0.7)}, \left( \frac{u_3}{(0.1, 0.6, 0.9)} \right) \right) \right\} \cdot \end{aligned}$$

$$\begin{aligned} \mathcal{H}(z_{10} = (m_2, n_2, 0)) &= \left\{ \left( \frac{u_1}{(0.3, 0.4, 0.7)} \right), \left( \frac{u_2}{(0.3, 0.4, 0.2)} \right), \left( \frac{u_3}{(0.6, 0.6, 0.2)} \right) \right\}. \\ \mathcal{H}(z_{11} = (m_3, n_1, 0)) &= \left\{ \left( \frac{u_1}{(0.4, 0.5, 0.6)} \right), \left( \frac{u_2}{(0.1, 0.2, 0.8)} \right), \left( \frac{u_3}{(0.2, 0.4, 0.7)} \right) \right\}. \\ \mathcal{H}(z_{12} = (m_3, n_2, 0)) &= \left\{ \left( \frac{u_1}{(0.6, 0.8, 0.6)} \right), \left( \frac{u_2}{(0.2, 0.5, 0.6)} \right), \left( \frac{u_3}{(0.4, 0.9, 0.9)} \right) \right\}. \end{aligned}$$

Also  $\mathcal{H}(z_i)$  can represent as a matrix as a following form:

$$\mathcal{H}(z_i) = \begin{pmatrix} ((0.2, 0.5, 0.3)) & ((0.3, 0.6, 0.7)) & ((0.8, 0.1, 0.6)) \\ ((0.5, 0.4, 0.2)) & ((0.1, 0.5, 0.7)) & ((0.2, 0.0, 0.8)) \\ ((0.7, 0.6, 0.2)) & ((0.6, 0.3, 0.1)) & ((0.2, 0.3, 0.5)) \\ ((0.5, 0.3, 0.2)) & ((0.6, 0.4, 0.7)) & ((0.5, 0.4, 0.3)) \\ ((0.7, 0.5, 0.9)) & ((0.2, 0.6, 0.7)) & ((0.3, 0.4, 0.6)) \\ ((0.2, 0.5, 0.3)) & ((0.3, 0.6, 0.7)) & ((0.8, 0.1, 0.6)) \\ ((0.4, 0.6, 0.2)) & ((0.7, 0.2, 0.5)) & ((0.1, 0.1, 0.3)) \\ ((0.4, 0.5, 0.8)) & ((0.9, 0.8, 0.7)) & ((0.1, 0.6, 0.9)) \\ ((0.3, 0.4, 0.7)) & ((0.3, 0.4, 0.2)) & ((0.6, 0.6, 0.2)) \\ ((0.4, 0.5, 0.6)) & ((0.1, 0.2, 0.8)) & ((0.2, 0.4, 0.7)) \\ ((0.6, 0.8, 0.6)) & ((0.2, 0.5, 0.6)) & ((0.4, 0.9, 0.9)) \end{pmatrix}$$

**Definition 3.3. (Agree SVNFSSES):** Agree SVNFSSES  $\mathcal{H}_1$  represents agreement of all expert’s opinions and is defined as follows:

$$\mathcal{H}_O = \{H_O(z_i) : z_i \in M \times N \times \{1\}\}$$

**Example 3.4.** Take the terms  $\mathcal{H}_O(z_1)$  in Example 3.2. above

$$\mathcal{K}(z_1 = (m_1, n_1, 1)) = \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.3)} \right), \left( \frac{u_2}{(0.3, 0.6, 0.7)} \right), \left( \frac{u_3}{(0.8, 0.1, 0.6)} \right) \right\}.$$

**Definition 3.5. (Disagree SVNFSSES):** disagree SVNFSSES  $\mathcal{H}_0$  represents disagreement of all expert’s opinions and is defined as follows:

$$\mathcal{H}_O = \{H_O(z_i) : z_i \in M \times N \times \{0\}\}$$

**Example 3.6.** Take the terms  $\mathcal{H}_O(z_9)$  in Example 3.2. above

$$\mathcal{K}(z_9 = (m_2, n_1, 0)) = \left\{ \left( \frac{u_1}{(0.4, 0.5, 0.8)} \right), \left( \frac{u_2}{(0.9, 0.8, 0.7)} \right), \left( \frac{u_3}{(0.1, 0.6, 0.9)} \right) \right\}.$$

**Definition 3.7. (SVNFSE-subset):** Let  $\mathcal{H}_O = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  and  $\mathcal{K}_P = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  be two SVNFSSE-sets on reference set  $\hat{U}$ . Then  $\mathcal{H}_O$  is said SVNFSSE-subset of  $\mathcal{K}_P$  and denoted by  $\mathcal{H}_O \subseteq \mathcal{K}_P$  if:

1.  $\mathcal{H}_O(u_i)$  is SVNFSSE-subset of  $\mathcal{K}_P(u_i), \forall u_i \in \hat{U}$ .

2.  $\mathcal{O} \subseteq P$ .

**Example 3.8.** Take the terms  $\mathcal{H}_O(z_i)$  in Example 3.2. above and take

$$\begin{aligned} \mathcal{K}(z_1 = (m_1, n_1, 1)) &= \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.3)} \right), \left( \frac{u_2}{(0.3, 0.6, 0.7)} \right), \left( \frac{u_3}{(0.8, 0.1, 0.6)} \right) \right\}. \\ \mathcal{K}(z_9 = (m_2, n_1, 0)) &= \left\{ \left( \frac{u_1}{(0.4, 0.5, 0.8)} \right), \left( \frac{u_2}{(0.9, 0.8, 0.7)} \right), \left( \frac{u_3}{(0.1, 0.6, 0.9)} \right) \right\}. \end{aligned}$$

Now, its clear the two terms are  $\subseteq \mathcal{H}_O$ .

**Definition 3.9. (Equality of SVNFSSE-sets):** Let  $\mathcal{H}_O = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  and  $\mathcal{K}_P = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  be two SVNFSSE-sets on reference set  $\hat{U}$ . Then  $\mathcal{H}_O = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  is called equal of  $\mathcal{K}_P = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  and denoted by  $\mathcal{H}_O = \mathcal{K}_P$  if:

1.  $\mathcal{H}(u_i)$  is SVNFSSE-subset of  $\mathcal{K}(u_i)$  and  $\mathcal{K}(u_i)$  is SVNFSSE-subset of  $\mathcal{H}(u_i), \forall u_i \in \hat{U}$ .

2.  $\mathcal{O}$  is subset of  $\mathcal{P}$  and  $\mathcal{P}$  is subset of  $\mathcal{O}$ ,  $\forall u_i \in \hat{\mathcal{U}}$ .

**Example 3.10.** Consider  $\mathcal{H}_{\mathcal{O}}(z_i)$  in Example 3.2. above and take  $\mathcal{H}_{\mathcal{O}} =$

$$\begin{pmatrix} (0.2, 0.50.3) & (0.3, 0.6, 0.8) & (0.8, 0.1, 0.6) \\ (0.5, 0.80.2) & (0.2, 0.4, 0.8) & (0.8, 0, 0.2) \\ (0.4, 0.50.3) & (0.9, 0.6, 0.8) & (0.8, 0.7, 0.7) \end{pmatrix} \text{ and}$$

$\mathcal{G}_{\mathcal{C}} =$

$$\begin{pmatrix} (0.1, 0.30.3) & (0.6, 0.8, 0.8) & (0.8, 0.1, 0.1) \\ (0.5, 0.80.2) & (0.2, 0.4, 0.8) & (0.8, 0, 0.2) \\ (0.4, 0.7, 0.8) & (0.4, 0.6, 0.8) & (0.8, 0.5, 0.4) \end{pmatrix} \text{ and}$$

$\mathcal{K}_{\mathcal{P}} =$

$$\begin{pmatrix} (0.2, 0.50.3) & (0.3, 0.6, 0.8) & (0.8, 0.1, 0.6) \\ (0.5, 0.80.2) & (0.2, 0.4, 0.8) & (0.8, 0, 0.2) \\ (0.4, 0.50.3) & (0.9, 0.6, 0.8) & (0.8, 0.7, 0.7) \end{pmatrix}$$

Then, its clear  $\mathcal{H}_{\mathcal{O}} = \mathcal{K}_{\mathcal{P}}$  and  $\mathcal{H}_{\mathcal{O}} \neq \mathcal{G}_{\mathcal{C}}$ .

**Definition 3.11. ( null SVNFSSE-set):** Let  $\mathcal{H}_{\mathcal{O}} = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  be SVNFSSE-set on reference set  $\hat{\mathcal{U}}$ . Then we say that  $\mathcal{H}_{\mathcal{O}}$  is " null SVNFSSE-set" and denoted by  $\hat{\Phi}_{(0)}$  if  $\mathcal{H}(u_i) = (0, 1, 1)$  and  $\Theta(z_i) = 0, \forall u_i \in \hat{\mathcal{U}}$ .

**Example 3.12.** Taking into account the matrix notation of  $\mathcal{H}_{\mathcal{O}}$  as an Example 3.2, it can be observed that we possess.

$$\hat{\Phi}_{(0)} = \begin{pmatrix} ((0, 1, 1)) & ((0, 1, 1)) & ((0, 1, 1)) \\ ((0, 1, 1)) & ((0, 1, 1)) & ((0, 1, 1)) \\ ((0, 1, 1)) & ((0, 1, 1)) & ((0, 1, 1)) \end{pmatrix}$$

**Definition 3.13. ( absolute SVNFSSE-set):** Let  $\mathcal{H}_{\mathcal{O}} = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  be SVNFSSE-set on reference set  $\hat{\mathcal{U}}$ . Then we say that  $\mathcal{H}_{\mathcal{O}}$  is " absolute SVNFSSE-set" and denoted by  $\hat{\Omega}_{(1)}$  if  $\mathcal{H}(u_i) = (1, 0, 0)$  and  $\Theta(z_i) = 1, \forall u_i \in \hat{\mathcal{U}}$ .

**Example 3.14.** Taking into account the matrix notation of  $\mathcal{H}_{\mathcal{O}}$  as an Example 3.2, it can be observed that we possess.

$$\hat{\Phi}_{(0)} = \begin{pmatrix} ((1, 0, 0)) & ((1, 0, 0)) & ((1, 0, 0)) \\ ((1, 0, 0)) & ((1, 0, 0)) & ((1, 0, 0)) \\ ((1, 0, 0)) & ((1, 0, 0)) & ((1, 0, 0)) \end{pmatrix}$$

**Definition 3.15. (Complement operation of SVNFSSE-set):** Let  $\mathcal{H}_{\mathcal{O}} = (\mathcal{H}, \mathcal{O} \subseteq \mathcal{Z})$  be SVNFSSE-set on reference set  $\hat{\mathcal{U}}$  and defined as follows

$$\mathcal{H}_{\mathcal{O}} = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{\mathcal{U}}, z \in \hat{\mathcal{Z}} \right) \right\}$$

Then, complement operation of PIVNS-set defined as follows

$$\mathcal{H}_{\mathcal{O}}^c = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j), 1 - \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j) \right\rangle \mid u \in \hat{\mathcal{U}}, z \in \hat{\mathcal{Z}} \right) \right\}$$

**Here:** we follow the complement role of SVNS-complement.

**Example 3.16.** Take the terms  $\mathcal{H}_{\mathcal{O}}(z_{i=1,9})$  in Example 3.2. above and take

$$\mathcal{H}(z_1 = (m_1, n_1, 1)) = \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.3)} \right), \left( \frac{u_2}{(0.3, 0.6, 0.7)} \right), \left( \frac{u_3}{(0.8, 0.1, 0.6)} \right) \right\} \cdot$$

$$\mathcal{H}(z_9 = (m_2, n_1, 0)) = \left\{ \left( \frac{u_1}{(0.4, 0.5, 0.8)} \right), \left( \frac{u_2}{(0.9, 0.8, 0.7)} \right), \left( \frac{u_3}{(0.1, 0.6, 0.9)} \right) \right\} \cdot$$

Then the complement of them given as following:

$$\mathcal{H}^c(z_1 = (m_1, n_1, 1)) = \left\{ \left( \frac{u_1}{(0.3, 0.5, 0.2)} \right), \left( \frac{u_2}{(0.7, 0.4, 0.3)} \right), \left( \frac{u_3}{(0.6, 0.9, 0.8)} \right) \right\} \cdot$$

$$\mathcal{H}^c(z_9 = (m_2, n_1, 0)) = \left\{ \left( \frac{u_1}{(0.8, 0.5, 0.4)} \right), \left( \frac{u_2}{(0.7, 0.2, 0.9)} \right), \left( \frac{u_3}{(0.9, 0.4, 0.1)} \right) \right\} \cdot$$

**4 The set-theoretic operations pertaining to SVNSE-sets**

Now, in this section, we introduce the set-theoretic operations pertaining to SVNSE-sets as well as some properties and numerical examples that illustrate how these tools work in algebraic environments.

**Definition 4.1. (Union of SVNFSE-sets)** Let

$$\mathcal{H}_O = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\}$$

and

$$\mathcal{K}_P = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{K}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{K}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{K}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\}$$

be two SVNFSE-sets on reference set  $\hat{U}$ . Then, the union of SVNFSE-sets denoted by  $\mathcal{H}_O \hat{\cup} \mathcal{K}_P$  and defined as following:  $\mathcal{D}_S = \mathcal{H}_O \hat{\cup} \mathcal{K}_P$ , where  $\hat{\cup}$  denotes SVNSE-sets-union.

**Example 4.2.** Taking into account one part of SVNFSES  $\mathcal{H}_O$  as an Example 3.2, and  $\mathcal{K}_P$  given in an Example 3.4,

$$\mathcal{H}(z_2 = (m_1, n_2, 1)) = \left\{ \left( \frac{u_1}{(0.5, 0.4, 0.2)}, \left( \frac{u_2}{(0.1, 0.5, 0.7)}, \left( \frac{u_3}{(0.2, 0, 0.8)} \right) \right) \right\}.$$

and

$$\mathcal{K}(z_2 = (m_1, n_2, 1)) = \left\{ \left( \frac{u_1}{(0.4, 0.5, 0.8)}, \left( \frac{u_2}{(0.9, 0.8, 0.7)}, \left( \frac{u_3}{(0.1, 0.6, 0.9)} \right) \right) \right\}.$$

then, the union of SVNFSE-sets can be possess as following :

$$\mathcal{D}(z_2 = (m_1, n_2, 1)) = \left\{ \left( \frac{u_1}{(0.5, 0.4, 0.2)}, \left( \frac{u_2}{(0.9, 0.5, 0.7)}, \left( \frac{u_3}{(0.2, 0, 0.9)} \right) \right) \right\}.$$

Also  $\mathcal{D}(u_i)$  can represent as a matrix as a following form:

$$\mathcal{D}_{\Psi} =$$

$$\left( ((0.5, 0.4, 0.2)) \quad ((0.9, 0.5, 0.7)) \quad ((0.2, 0, 0.9)) \right)_{1 \times 3}$$

**Definition 4.3. (Intersection of SVNFSE-sets)** Let

$$\mathcal{H}_O = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\}$$

and

$$\mathcal{K}_P = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{K}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{K}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{K}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\}$$

be two SVNFSE-sets on reference set  $\hat{U}$ . Then, the intersection of SVNFSE-sets denoted by  $\mathcal{H}_O \hat{\cap} \mathcal{K}_P$  and defined as following:  $\mathcal{C}_L = \mathcal{H}_O \hat{\cap} \mathcal{K}_P$ , where  $\hat{\cap}$  denotes SVNSE-sets-intersection.

**Example 4.4.** Taking into account one part of SVNFSES  $\mathcal{H}_O$  as an Example 3.2, and  $\mathcal{K}_P$  given in an Example 3.4,

$$\mathcal{H}(z_2 = (m_1, n_2, 1)) = \left\{ \left( \frac{u_1}{(0.5, 0.4, 0.2)}, \left( \frac{u_2}{(0.1, 0.5, 0.7)}, \left( \frac{u_3}{(0.2, 0, 0.8)} \right) \right) \right\}.$$

and

$$\mathcal{K}(z_2 = (m_1, n_2, 1)) = \left\{ \left( \frac{u_1}{(0.4, 0.5, 0.8)}, \left( \frac{u_2}{(0.9, 0.8, 0.7)}, \left( \frac{u_3}{(0.1, 0.6, 0.9)} \right) \right) \right\}.$$

then, the intersection of SVNFSE-sets can be possess as following :

$$\mathcal{C}(z_2 = (m_1, n_2, 1)) = \left\{ \left( \frac{u_1}{(0.4, 0.5, 0.8)}, \left( \frac{u_2}{(0.1, 0.8, 0.7)}, \left( \frac{u_3}{(0.1, 0, 0.8)} \right) \right) \right\}.$$

Also  $\mathcal{C}(u_i)$  can represent as a matrix as a following form:

$$\mathcal{C}_{\mathcal{L}} =$$

$$\left( ((0.4, 0.5, 0.8)) \quad ((0.1, 0.8, 0.7)) \quad ((0.1, 0, 0.8)) \right)_{1 \times 3}$$

**Proposition 4.5.** *Let*

$$\mathcal{H}_O = \left\{ \left( u, \left\langle \ddot{\partial}_H^t(z_i)(u_j), \ddot{\partial}_H^i(z_i)(u_j), \ddot{\partial}_H^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\}$$

be a SVNFSSE-set on reference set  $\hat{U}$ . Then, the following statements hold:

1.  $\mathcal{H}_O \hat{\cup} \mathcal{H}_O = \mathcal{H}_O$ .
2.  $\mathcal{H}_O \hat{\cap} \mathcal{H}_O = \mathcal{H}_O$ .
3.  $\mathcal{H}_O \hat{\cup} \widehat{\Phi}_{(0)} = \mathcal{H}_O$ .
4.  $\mathcal{H}_O \hat{\cap} \widehat{\Phi}_{(0)} = \widehat{\Phi}_{(0)}$ .
5.  $\mathcal{H}_O \hat{\cup} \widehat{\Omega}_{(1)} = \widehat{\Omega}_{(1)}$ .
6.  $\mathcal{H}_O \hat{\cap} \widehat{\Omega}_{(1)} = \mathcal{H}_O$ .

**Proposition 4.6.** *Let*

$$\mathcal{H}_O = \left\{ \left( u, \left\langle \ddot{\partial}_H^t(z_i)(u_j), \ddot{\partial}_H^i(z_i)(u_j), \ddot{\partial}_H^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\},$$

$$\mathcal{K}_P = \left\{ \left( u, \left\langle \ddot{\partial}_K^t(z_i)(u_j), \ddot{\partial}_K^i(z_i)(u_j), \ddot{\partial}_K^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\} \text{ and}$$

$\mathcal{G}_C = \left\{ \left( u, \left\langle \ddot{\partial}_G^t(z_i)(u_j), \ddot{\partial}_G^i(z_i)(u_j), \ddot{\partial}_G^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\}$  be three SVNFSSE-sets on reference set  $\hat{U}$ . Then, the following statements hold:

1.  $\mathcal{H}_O \hat{\cup} \mathcal{G}_C = \mathcal{G}_C \hat{\cup} \mathcal{H}_O$ .
2.  $\mathcal{H}_O \hat{\cap} \mathcal{G}_C = \mathcal{G}_C \hat{\cap} \mathcal{H}_O$ .
3.  $\mathcal{H}_O \hat{\cup} (\mathcal{G}_C \hat{\cup} \mathcal{K}_P) = (\mathcal{H}_O \hat{\cup} \mathcal{G}_C) \hat{\cup} \mathcal{K}_P$ .
4.  $\mathcal{H}_O \hat{\cap} (\mathcal{G}_C \hat{\cap} \mathcal{K}_P) = (\mathcal{H}_O \hat{\cap} \mathcal{G}_C) \hat{\cap} \mathcal{K}_P$ .
5.  $\mathcal{H}_O \hat{\cup} (\mathcal{G}_C \hat{\cap} \mathcal{K}_P) = (\mathcal{H}_O \hat{\cup} \mathcal{G}_C) \hat{\cap} (\mathcal{H}_O \hat{\cup} \mathcal{K}_P)$ .
6.  $\mathcal{H}_O \hat{\cap} (\mathcal{G}_C \hat{\cup} \mathcal{K}_P) = (\mathcal{H}_O \hat{\cap} \mathcal{G}_C) \hat{\cup} (\mathcal{H}_O \hat{\cap} \mathcal{K}_P)$ .

*Proof. 1.* We take the left side  $\mathcal{H}_O \hat{\cup} \mathcal{G}_C =$

$$H_O = \left\{ \left( u, \left\langle \ddot{\partial}_H^t(z_i)(u_j), \ddot{\partial}_H^i(z_i)(u_j), \ddot{\partial}_H^f(z_i)(u_j) \right\rangle \mid u \in U, z \in Z \right) \right\} \cup$$

$$G_C = \left\{ \left( u, \left\langle \ddot{\partial}_G^t(z_i)(u_j), \ddot{\partial}_G^i(z_i)(u_j), \ddot{\partial}_G^f(z_i)(u_j) \right\rangle \mid u \in U, z \in Z \right) \right\}$$

$$= \left\{ \left( u, \left\langle \max \left( \ddot{\partial}_H^t(z_i)(u_j), \ddot{\partial}_G^t(z_i)(u_j) \right), \min \left( \ddot{\partial}_H^i(z_i)(u_j), \ddot{\partial}_G^i(z_i)(u_j) \right), \min \left( \ddot{\partial}_H^f(z_i)(u_j), \ddot{\partial}_G^f(z_i)(u_j) \right) \right\rangle \right) \right\}$$

$$= \left\{ \left( u, \left\langle \max \left( \ddot{\partial}_G^t(z_i)(u_j), \ddot{\partial}_H^t(z_i)(u_j) \right), \min \left( \ddot{\partial}_G^i(z_i)(u_j), \ddot{\partial}_H^i(z_i)(u_j) \right), \min \left( \ddot{\partial}_G^f(z_i)(u_j), \ddot{\partial}_H^f(z_i)(u_j) \right) \right\rangle \right) \right\}$$

$$= G_C = \left\{ \left( u, \left\langle \ddot{\partial}_G^t(z_i)(u_j), \ddot{\partial}_G^i(z_i)(u_j), \ddot{\partial}_G^f(z_i)(u_j) \right\rangle \mid u \in U, z \in Z \right) \right\} \cup$$

$$H_O = \left\{ \left( u, \left\langle \ddot{\partial}_H^t(z_i)(u_j), \ddot{\partial}_H^i(z_i)(u_j), \ddot{\partial}_H^f(z_i)(u_j) \right\rangle \mid u \in U, z \in Z \right) \right\}.$$

$$= \mathcal{G}_C \hat{\cup} \mathcal{H}_O.$$

□

*Note:* The rest of the proof is similar to proof method 1

**Proposition 4.7.** *Let*

$$\mathcal{H}_O = \left\{ \left( u, \left\langle \ddot{\partial}_H^t(z_i)(u_j), \ddot{\partial}_H^i(z_i)(u_j), \ddot{\partial}_H^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\},$$

$\mathcal{K}_P = \left\{ \left( u, \left\langle \ddot{\partial}_K^t(z_i)(u_j), \ddot{\partial}_K^i(z_i)(u_j), \ddot{\partial}_K^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\}$  be two SVNFSSE-sets on reference set  $\hat{U}$ . Then, the following statements hold:

1.  $(\mathcal{H}_O^c)^c = \mathcal{H}_O$ .
2.  $(\mathcal{H}_O \hat{\cup} \mathcal{K}_P)^c = \mathcal{H}_O^c \hat{\cap} \mathcal{K}_P^c$ .
3.  $(\mathcal{H}_O \hat{\cap} \mathcal{K}_P)^c = \mathcal{H}_O^c \hat{\cup} \mathcal{K}_P^c$ .

Here, paragraphs 2 and 3 refer to De Morgan’s law.

*Proof.* **1.**  $\mathcal{H}_O = \left\{ u_j \left( \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j) \right) \mid z_i \in Z, u_j \in U \right\}$   
 $\mathcal{H}_O^c = \left\{ u_j \left( \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j), 1 - \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j) \right) \mid z_i \in Z, u_j \in U \right\}$   
 $(\mathcal{H}_O^c)^c = \left\{ u_j \left( \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), 1 - \left( 1 - \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j) \right), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j) \right) \mid z_i \in Z, u_j \in U \right\}$   
 $= \left\{ u_j \left( \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j) \right) \mid z_i \in Z, u_j \in U \right\}$   
 $= \mathcal{H}_O$

**Note:** The rest of proof is similar to proof method 1 □

**Definition 4.8. (AND of SVNFSSE-sets):** Let  $\mathcal{H}_O$  and  $\mathcal{G}_C$  be two SVNFSSE-sets on reference set  $\hat{U}$ . Then the "AND" operation of both  $\mathcal{H}_O$  and  $\mathcal{G}_C$  defined as  $\mathcal{H}_O \wedge \mathcal{G}_C = \mathcal{R}_L$  such that  $\mathcal{R}_L(u_i, u_j)(z_i) = \mathcal{H}_O(u_j)(z_i) \cap \mathcal{G}_C(u_j)(z_i)$  Here  $\cap$  refer to the intersection of SVNFS.

**Example 4.9.** Taking into account the SVNFSSE  $\mathcal{H}_O$  as an Example 3.2, and  $\mathcal{G}_C$  given in an Example 3.4, then, **(AND of SVNFSSE-sets):** can be possess as following :

$$\mathcal{R}_L(z_1 \times z_1) = \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right), \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right), \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right) \right\} \right\}.$$

$$\mathcal{R}_L(z_1 \times z_2) = \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right), \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right), \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right) \right\} \right\}.$$

**Definition 4.10. (OR of SVNFSSE-sets):** Let  $\mathcal{H}_O$  and  $\mathcal{G}_C$  be two SVNFSSE-sets on reference set  $\hat{U}$ . Then the "OR" operation of both  $\mathcal{H}_O$  and  $\mathcal{G}_C$  defined as  $\mathcal{H}_O \vee \mathcal{G}_C = \mathcal{F}_V$  such that  $\mathcal{F}_V(u_i, u_j)(z_i) = \mathcal{H}_O(u_j)(z_i) \cup \mathcal{G}_C(u_j)(z_i)$  Here  $\cup$  refer to the union of SVNFS.

**Example 4.11.** Taking into account the SVNFSSE  $\mathcal{H}_O$  as an Example 3.2, and  $\mathcal{G}_C$  given in an Example 3.4, then, **(OR of SVNFSSE-sets):** can be possess as following :

$$\mathcal{R}_L(z_1 \times z_1) = \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right), \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right), \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right) \right\} \right\}.$$

$$\mathcal{R}_L(z_1 \times z_2) = \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right), \left\{ \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right), \left( \frac{u_1}{(0.2, 0.5, 0.1)} \right) \right\} \right\}.$$

### 5 Distance Measure on SVNFSSE-sets

In this part, we introduce and study the distance measure (DM) of SVNFSSES in order to calculate the ratio of DM between two SVNFSSES. After that, we will employ these DMs in one application in DM problem by proposing an algorithm shown in Figure1.

**Definition 5.1.** Let

$$\mathcal{H}_O = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{H}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{H}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\},$$

$$\mathcal{K}_P = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{K}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{K}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{K}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\} \text{ and}$$

$\mathcal{G}_C = \left\{ \left( u, \left\langle \ddot{\partial}_{\mathcal{G}}^t(z_i)(u_j), \ddot{\partial}_{\mathcal{G}}^i(z_i)(u_j), \ddot{\partial}_{\mathcal{G}}^f(z_i)(u_j) \right\rangle \mid u \in \hat{U}, z_i \in \hat{Z} \right) \right\}$  be three SVNFSSES on reference set  $\hat{U}$ . Then, a function  $\mathbb{D}:\text{SVNFSSES} \times \text{SVNFSSES} \rightarrow [0, 1]$  is called distance measure (SVNFSSES( $\hat{U}$ )) if  $\mathbb{D}$  fulfilled the following notes:

i.  $\mathbb{D}(\mathcal{H}_O, \mathcal{G}_C) \geq 0$ , and  $\mathbb{D}(\mathcal{H}_O, \mathcal{G}_C) = 0$ , iff both  $\mathcal{H}_O = \mathcal{G}_C$ .

ii.  $\mathbb{D}(\mathcal{H}_O, \mathcal{G}_C) = \mathbb{D}(\mathcal{G}_C, \mathcal{H}_O)$

iii.  $\mathbb{D}(\mathcal{H}_O, \mathcal{K}_P) \leq \mathbb{D}(\mathcal{H}_O, \mathcal{K}_P) + \mathbb{D}(\mathcal{G}_C, \mathcal{K}_P)$  (triangle inequality).

Now based on definition 5.1, we will define the following distance measures:

**1.Hamming distance**

$$\mathbb{D}_{SVNFSSES}^H(\mathcal{H}_O, \mathcal{G}_C) = \frac{1}{3} \sum_i^n \sum_j^m \left[ \frac{|\Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^t| + |\Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^i| + |\Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^f|}{3} \right] \text{ where}$$

$$\Delta_{ij} \ddot{\partial}_{\mathcal{H}, \mathcal{G}}^t = \left( \ddot{\partial}_{\mathcal{H}_O}^t - \ddot{\partial}_{\mathcal{G}_C}^t \right),$$

$$\Delta_{ij} \ddot{\partial}_{\mathcal{H}, \mathcal{G}}^i = \left( \ddot{\partial}_{\mathcal{H}_O}^i - \ddot{\partial}_{\mathcal{G}_C}^i \right),$$

$$\Delta_{ij} \ddot{\partial}_{\mathcal{H}, \mathcal{G}}^f = \left( \ddot{\partial}_{\mathcal{H}_O}^f - \ddot{\partial}_{\mathcal{G}_C}^f \right),$$

**2.Normalized Hamming distance**

$$\mathbb{D}_{SVNFSSES}^{NH}(\mathcal{H}_O, \mathcal{G}_C) = \frac{D_{SVNFSSES}^H(\mathcal{P}_O, \mathcal{G}_C)}{mn}.$$

**3. Euclidean distance**

$$D_{SVNFSSES}^E(\mathcal{H}_O, \mathcal{G}_C) = \sqrt{\frac{1}{3} \sum_i^n \sum_j^m \left[ \frac{|\Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^t| + |\Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^i| + |\Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^f|}{6} \right]}$$

where

$$\Delta_{ij} \ddot{\partial}_{\mathcal{H}, \mathcal{G}}^t = \left( \ddot{\partial}_{\mathcal{H}_O}^t - \ddot{\partial}_{\mathcal{G}_C}^t \right),$$

$$\Delta_{ij} \ddot{\partial}_{\mathcal{H}, \mathcal{G}}^i = \left( \ddot{\partial}_{\mathcal{H}_O}^i - \ddot{\partial}_{\mathcal{G}_C}^i \right),$$

$$\Delta_{ij} \ddot{\partial}_{\mathcal{H}, \mathcal{G}}^f = \left( \ddot{\partial}_{\mathcal{H}_O}^f - \ddot{\partial}_{\mathcal{G}_C}^f \right),$$

**4. Normalized Euclidean distance**

$$\mathbb{D}_{SVNFSSES}^{NE}(\mathcal{H}_O, \mathcal{G}_O) = \frac{D_{SVNFSSES}^E(\mathcal{P}_O, \mathcal{G}_O)}{\sqrt{mn}}.$$

**Theorem 5.2.** The function  $\mathbb{D}_{SVNFSSES}^H(\mathcal{H}_O, \mathcal{G}_C)$ ,  $\mathbb{D}_{SVNFSSES}^{NH}(\mathcal{H}_O, \mathcal{G}_C)$ ,  $D_{SVNFSSES}^E(\mathcal{H}_O, \mathcal{G}_C)$  and  $\mathbb{D}_{SVNFSSES}^{NE}(\mathcal{H}_O, \mathcal{G}_C) : \text{SVNFSSES}(\hat{U}) \rightarrow R^+$  that given by Definition 5.1 respectively are metrics, where  $R^+$  is the set all non-negative real-numbers.

*Proof.* The proof is straightforward. □

Sometimes users resort to adjusting the resulting values when using the tools given above, using weight values that are specified by the user. Now we will redefine the above tools, but combined with weight values.

Now based on definition 5.1, we will define the following distance measures:

**1. Weighted Hamming distance:**

$$D_{SVNFSES}^{wH}(\mathcal{H}_O, \mathcal{G}_C) = \left\{ \frac{1}{3} \sum_i^n \sum_j^m w_i \left[ \left| \Delta_{ij} \ddot{\partial}_{\mathcal{P}_O, \mathcal{G}_C}^t \right| + \left| \Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_O}^i \right| + \left| \Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^f \right| \right]^\gamma \right\}^{\frac{1}{\gamma}}$$

where

$$\begin{aligned} \Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^t &= \left| \ddot{\partial}_{\mathcal{P}_O}^t - \ddot{\partial}_{\mathcal{G}_C}^t \right|^\gamma \\ \Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^i &= \left| \ddot{\partial}_{\mathcal{P}_O}^i - \ddot{\partial}_{\mathcal{G}_C}^i \right|^\gamma \\ \Delta_{ij} \ddot{\partial}_{\mathcal{H}_O, \mathcal{G}_C}^f &= \left| \ddot{\partial}_{\mathcal{P}_O}^f - \ddot{\partial}_{\mathcal{G}_C}^f \right|^\gamma \end{aligned}$$

**2. Weighted Normalized Hamming distance:**

$$\mathbb{D}_{SVNFSES}^{wNH}(\mathcal{H}_O, \mathcal{G}_C) = \frac{D_{SVNFSES}^{wH}(\mathcal{H}_O, \mathcal{G}_C)}{mn}$$

**6 Application of SVNFSE-sets in real-life situations based on DM**

In this section of the current research, we will create a new algorithm based on the tools presented in this work to solve one of the decision-making problems (to help a couple choose a new home in one of the residential complexes). This algorithm will present its steps in Figure 1 as following:

**6.1 Statement of the problem**

Assume that the married couple, Mr. Xu and Mrs. Xu wants to purchase a house in one of the low-cost residential complexes. In the low-cost residential complexes, there are two houses that represent by reference set  $\hat{U} = \{u_1, u_2\}$ . The two couples in their selection focus on observing the attributes that can be represented by the following attribute set  $\mathbb{M} = \{m_1, m_2, m_3\}$  such that  $m_1 = \text{House area}$ ,  $m_2 = \text{House price}$ , and  $m_3 = \text{Materials used in building the house}$ . In this scenario, the couple resorts to two experts  $\{n_1, n_2\}$  in real estate issues for the purpose of consultation. Here we will analyze the expert's opinions by building two models (SVNFSES – memberships) from our proposed evaluation of this evaluation as a follows:

**Step 1.** Build three SVNFSES models  $\mathcal{K}_P$  represents optimal evaluation and  $\mathcal{H}_O$  and  $\mathcal{G}_C$  represents the opinion of the first expert ( $n_1$ ) and the second expert ( $n_2$ ) for both first house ( $u_1$ ) and the second house ( $u_2$ ):

$$\begin{aligned} \mathcal{K}_C &= \\ \left\{ \mathcal{K}(z_1 = (m_1, n_1, 1)) &= \left\{ \left( \frac{u_1}{(1,0,0)} \right), \left( \frac{u_2}{(1,0,0)} \right) \right\} \cdot \right. \\ \mathcal{K}(z_2 = (m_1, n_2, 1)) &= \left\{ \left( \frac{u_1}{(1,0,0)} \right), \left( \frac{u_3}{(1,0,0)} \right) \right\} \cdot \end{aligned}$$

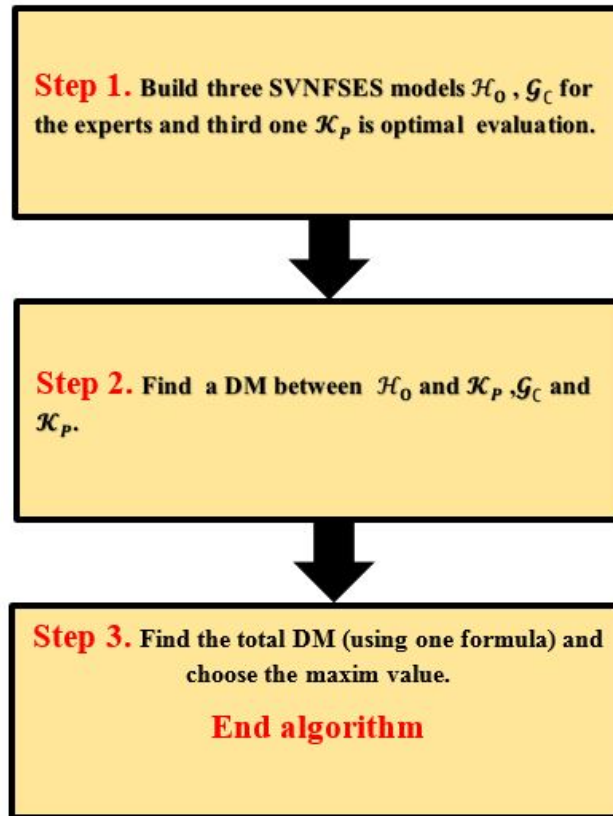


Figure 1: The propose algorithm based on DM technical

$$\begin{aligned}
 \mathcal{K}(z_3 = (m_2, n_1, 1)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_4 = (m_2, n_2, 1)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_5 = (m_3, n_1, 1)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_6 = (m_3, n_2, 1)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_7 = (m_1, n_1, 0)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_8 = (m_1, n_2, 0)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_9 = (m_2, n_1, 0)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_{10} = (m_2, n_2, 0)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_{11} = (m_3, n_1, 0)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot \\
 \mathcal{K}(z_{12} = (m_3, n_2, 0)) &= \left\{ \left( \frac{u_1}{\langle 1, 0, 0 \rangle} \right), \left( \frac{u_2}{\langle 1, 0, 0 \rangle} \right) \right\} \cdot
 \end{aligned}$$

$$\begin{aligned}
 \mathcal{H}_O = \\
 \mathcal{H}(z_1 = (m_1, n_1, 1)) &= \left\{ \left( \frac{u_1}{\langle 0.2, 0.5, 0.3 \rangle} \right), \left( \frac{u_2}{\langle 0.8, 0.1, 0.6 \rangle} \right) \right\} \cdot \\
 \mathcal{H}(z_2 = (m_1, n_2, 1)) &= \left\{ \left( \frac{u_1}{\langle 0.5, 0.4, 0.2 \rangle} \right), \left( \frac{u_3}{\langle 0.2, 0, 0.8 \rangle} \right) \right\} \cdot \\
 \mathcal{H}(z_3 = (m_2, n_1, 1)) &= \left\{ \left( \frac{u_1}{\langle 0.7, 0.6, 0.2 \rangle} \right), \left( \frac{u_2}{\langle 0.2, 0.3, 0.5 \rangle} \right) \right\} \cdot \\
 \mathcal{H}(z_4 = (m_2, n_2, 1)) &= \left\{ \left( \frac{u_1}{\langle 0.5, 0.3, 0.2 \rangle} \right), \left( \frac{u_2}{\langle 0.5, 0.4, 0.3 \rangle} \right) \right\} \cdot \\
 \mathcal{H}(z_5 = (m_3, n_1, 1)) &= \left\{ \left( \frac{u_1}{\langle 0.7, 0.5, 0.9 \rangle} \right), \left( \frac{u_2}{\langle 0.3, 0.4, 0.6 \rangle} \right) \right\} \cdot \\
 \mathcal{H}(z_6 = (m_3, n_2, 1)) &= \left\{ \left( \frac{u_1}{\langle 0.2, 0.5, 0.3 \rangle} \right), \left( \frac{u_2}{\langle 0.8, 0.1, 0.6 \rangle} \right) \right\} \cdot
 \end{aligned}$$

$$\begin{aligned} \mathcal{H}(z_7 = (m_1, n_1, 0)) &= \left\{ \left( \frac{u_1}{\langle 0.2, 0.5, 0.3 \rangle} \right), \left( \frac{u_2}{\langle 0.8, 0.1, 0.6 \rangle} \right) \right\}. \\ \mathcal{H}(z_8 = (m_1, n_2, 0)) &= \left\{ \left( \frac{u_1}{\langle 0.4, 0.6, 0.2 \rangle} \right), \left( \frac{u_2}{\langle 0.1, 0.1, 0.3 \rangle} \right) \right\}. \\ \mathcal{H}(z_9 = (m_2, n_1, 0)) &= \left\{ \left( \frac{u_1}{\langle 0.4, 0.5, 0.8 \rangle} \right), \left( \frac{u_2}{\langle 0.1, 0.6, 0.9 \rangle} \right) \right\}. \\ \mathcal{H}(z_{10} = (m_2, n_2, 0)) &= \left\{ \left( \frac{u_1}{\langle 0.3, 0.4, 0.7 \rangle} \right), \left( \frac{u_2}{\langle 0.6, 0.6, 0.2 \rangle} \right) \right\}. \\ \mathcal{H}(z_{11} = (m_3, n_1, 0)) &= \left\{ \left( \frac{u_1}{\langle 0.4, 0.5, 0.6 \rangle} \right), \left( \frac{u_2}{\langle 0.2, 0.4, 0.7 \rangle} \right) \right\}. \\ \mathcal{H}(z_{12} = (m_3, n_2, 0)) &= \left\{ \left( \frac{u_1}{\langle 0.6, 0.8, 0.6 \rangle} \right), \left( \frac{u_2}{\langle 0.4, 0.9, 0.9 \rangle} \right) \right\} \end{aligned}$$

$$\begin{aligned} \mathcal{G}_C = \\ \left\{ \mathcal{G}(z_1 = (m_1, n_1, 1)) = \left\{ \left( \frac{u_1}{\langle 0.1, 0.4, 0.2 \rangle} \right), \left( \frac{u_2}{\langle 0.3, 0.2, 0.7 \rangle} \right) \right\} \right. \\ \mathcal{G}(z_2 = (m_1, n_2, 1)) = \left\{ \left( \frac{u_1}{\langle 0.8, 0.7, 0.1 \rangle} \right), \left( \frac{u_3}{\langle 0.3, 0.4, 0.6 \rangle} \right) \right\}. \\ \mathcal{G}(z_3 = (m_2, n_1, 1)) = \left\{ \left( \frac{u_1}{\langle 0.9, 0.2, 0.1 \rangle} \right), \left( \frac{u_2}{\langle 0.2, 0.5, 0.9 \rangle} \right) \right\}. \\ \mathcal{G}(z_4 = (m_2, n_2, 1)) = \left\{ \left( \frac{u_1}{\langle 0.6, 0.3, 0.4 \rangle} \right), \left( \frac{u_2}{\langle 0.4, 0.4, 0.6 \rangle} \right) \right\}. \\ \mathcal{G}(z_5 = (m_3, n_1, 1)) = \left\{ \left( \frac{u_1}{\langle 0.4, 0.5, 0.9 \rangle} \right), \left( \frac{u_2}{\langle 0.8, 0.7, 0. \rangle} \right) \right\}. \\ \mathcal{G}(z_6 = (m_3, n_2, 1)) = \left\{ \left( \frac{u_1}{\langle 0.3, 0.5, 0.6 \rangle} \right), \left( \frac{u_2}{\langle 0.9, 0.2, 0.5 \rangle} \right) \right\}. \\ \mathcal{G}(z_7 = (m_1, n_1, 0)) = \left\{ \left( \frac{u_1}{\langle 0.2, 0.5, 0.3 \rangle} \right), \left( \frac{u_2}{\langle 0.2, 0.4, 0.9 \rangle} \right) \right\}. \\ \mathcal{G}(z_8 = (m_1, n_2, 0)) = \left\{ \left( \frac{u_1}{\langle 0.9, 0.1, 0.2 \rangle} \right), \left( \frac{u_2}{\langle 0.5, 0.3, 0.3 \rangle} \right) \right\}. \\ \mathcal{G}(z_9 = (m_2, n_1, 0)) = \left\{ \left( \frac{u_1}{\langle 0.6, 0.4, 0.8 \rangle} \right), \left( \frac{u_2}{\langle 0.2, 0.3, 0.9 \rangle} \right) \right\}. \\ \mathcal{G}(z_{10} = (m_2, n_2, 0)) = \left\{ \left( \frac{u_1}{\langle 0.6, 0.4, 0.7 \rangle} \right), \left( \frac{u_2}{\langle 0.8, 0.9, 0.1 \rangle} \right) \right\}. \\ \mathcal{G}(z_{11} = (m_3, n_1, 0)) = \left\{ \left( \frac{u_1}{\langle 0.5, 0.6, 0.2 \rangle} \right), \left( \frac{u_2}{\langle 0.7, 0.2, 0.9 \rangle} \right) \right\}. \\ \left. \mathcal{G}(z_{12} = (m_3, n_2, 0)) = \left\{ \left( \frac{u_1}{\langle 0.7, 0.4, 0.2 \rangle} \right), \left( \frac{u_2}{\langle 0.5, 0.6, 0.8 \rangle} \right) \right\} \right\} \end{aligned}$$

**Step 2.** Find the DM between  $\mathcal{K}_P$  and  $\mathcal{H}_O$  according definition 5.1 in part 5, (the DM value presented in Table 1.) . After that, find the DM between  $\mathcal{K}_P$  and  $\mathcal{G}_C$  according definition 5.1 in part 5, (the DM value presented in Table 2.)

Table 1: Valudes of DM Between  $\mathcal{K}(u_i), \mathcal{H}(u_i)$

$\bar{D}_k(\mathcal{K}(u_i), \mathcal{H}(u_i))$	Degree of DM
$\bar{D}_1^H(\mathcal{K}(z_1), \mathcal{H}(z_1))$	0.354
$\bar{D}_1^H(\mathcal{K}(z_2), \mathcal{H}(z_2))$	0.514
$\bar{D}_1^H(\mathcal{K}(z_3), \mathcal{H}(z_3))$	0.589
$\bar{D}_1^H(\mathcal{K}(z_4), \mathcal{H}(z_4))$	0.724
$\bar{D}_1^H(\mathcal{K}(z_5), \mathcal{H}(z_5))$	0.474
$\bar{D}_1^H(\mathcal{K}(z_6), \mathcal{H}(z_6))$	0.414
$\bar{D}_1^H(\mathcal{K}(z_7), \mathcal{H}(z_7))$	0.571
$\bar{D}_1^H(\mathcal{K}(z_8), \mathcal{H}(z_8))$	0.571
$\bar{D}_1^H(\mathcal{K}(z_9), \mathcal{H}(z_9))$	0.637
$\bar{D}_1^H(\mathcal{K}(z_{10}), \mathcal{H}(z_{10}))$	0.592
$\bar{D}_1^H(\mathcal{K}(z_{11}), \mathcal{H}(z_{11}))$	0.469
$\bar{D}_1^H(\mathcal{K}(z_{12}), \mathcal{H}(z_{12}))$	0.469
<b>Total of <math>\mathbb{D}^H(\mathcal{K}_P, \mathcal{H}_O) =</math></b>	<b>X=6.408</b>

**Step 3.** The value of  $X \geq Y$ , therefor the two couple will choose house  $u_1$ .

Table 2: Valudes of DM Between  $\mathcal{K}(u_i), \mathcal{G}(u_i)$ 

$\check{D}_k(\mathcal{K}(u_i), \mathcal{G}(u_i))$	Degree of DM
$\check{D}_1^H(\mathcal{K}(z_1), \mathcal{G}(z_1))$	0.437
$\check{D}_1^H(\mathcal{K}(z_2), \mathcal{G}(z_2))$	0.543
$\check{D}_1^H(\mathcal{K}(z_3), \mathcal{G}(z_3))$	0.363
$\check{D}_1^H(\mathcal{K}(z_4), \mathcal{G}(z_4))$	0.368
$\check{D}_1^H(\mathcal{K}(z_5), \mathcal{G}(z_5))$	0.325
$\check{D}_1^H(\mathcal{K}(z_6), \mathcal{G}(z_6))$	0.532
$\check{D}_1^H(\mathcal{K}(z_7), \mathcal{G}(z_7))$	0.453
$\check{D}_1^H(\mathcal{K}(z_8), \mathcal{G}(z_8))$	0.764
$\check{D}_1^H(\mathcal{K}(z_9), \mathcal{G}(z_9))$	0.538
$\check{D}_1^H(\mathcal{K}(z_{10}), \mathcal{G}(z_{10}))$	0.428
$\check{D}_1^H(\mathcal{K}(z_{11}), \mathcal{G}(z_{11}))$	0.476
$\check{D}_1^H(\mathcal{K}(z_{12}), \mathcal{G}(z_{12}))$	0.547
<b>Total of <math>\mathbb{D}^H(\mathcal{K}_P, \mathcal{G}_C) =</math></b>	<b>Y=5.784</b>

## 7 Conclusion

In this work, the notion of a SVNFSSES as combining the critical features of FSs and SVNFSs under expert systems in one model or as an extension to SES is introduced. The basic properties of this model namely null, absolute, subset, equality, and complement are presented. Also, the basic set theory like union, intersection, OR, and AND operations as well as some properties on SVNFSSESs are discussed. Finally, we presented a decision-making method based on SVNFSSES and gave an application of this method to solve a decision-making problem. For future research work, users can combine these tools with other fuzzy algebraic tools.

**Funding:** We thank the Deanship of Scientific Research, Vice President for Graduate Studies and Scientific Research, King Faisal University, Saudi Arabia for their support under [Grant No. GRANT5916].

**Acknowledgments:** This work was supported by the Deanship of Scientific Research, Vice Presidency for Graduate Studies and Scientific Research, King Faisal University, Saudi Arabia [Grant No. GRANT5916].

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