Research Article

Similarity Measures between Temporal Complex Intuitionistic Fuzzy Sets and Application in Pattern Recognition and Medical Diagnosis

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This work addresses the issue of similarity measures between two temporal complex Atanassov’s intuitionistic fuzzy sets, many measures of similarity between complex Atanassov’s intuitionistic fuzzy sets. What was proposed before did not consider the abstention group influence, which may lead to counterintuitive results in some cases. A new structure of temporal complex Atanassov’s intuitionistic fuzzy sets is obtained. This set is formally generalized from a conventional Atanassov’s intuitionistic complex fuzzy sets. Here we analyze the limitations of the existing similarity measures. Then, a new similarity measure of temporal complex Atanassov’s intuitionistic fuzzy sets is proposed and several numeric examples are given to demonstrate the validity of the proposed measure. Finally, the proposed similarity measure is applied to pattern recognition and medical diagnosis.

1. Introduction

Fuzzy set theory was conferred by Zadeh [1] to solve difficulties in dealing with uncertainties. Since then, the theories of fuzzy sets and fuzzy logic have been examined by many researchers to solve many real life problems involving ambiguous and uncertain environment. By adding a new component the idea of the concept of Atanassov’s intuitionistic fuzzy set (AIFS) was introduced [2]. Applications of these sets have been broadly studied in other aspects such as image processing [3], multicriteria decision making [4], pattern recognition [5], etc. Buckley [6] and Nguyen et al. [7] combined complex numbers with fuzzy sets. On the other hand, the innovative complex fuzzy set is introduced. The complex fuzzy set is characterized by a membership function, \( \mu_A(x) \), whose range is not limited to \([0,1]\) but extended to the unit circle in the complex plane. Hence, \( \mu_A(x) \) is a complex-Valued function that assigns a grade of membership of the form \( r_A(x)e^{i\omega_A(x)} \), \( i = \sqrt{-1} \) to any element in the universe of discourse. The value of \( \mu_A(x) \) is defined by the two variables, \( r_A(x) \) and \( \omega_A(x) \), both real-valued, with \( \mu_A(x) \in [0,1] \). Complex fuzzy set theory modifies the original concept of fuzzy membership by asserting that, at least in some instances, it is necessary to add a second dimension to the expression of membership. However, this added dimension does not alter the basic concept of fuzziness. Membership in a complex fuzzy set remains “as fuzzy” as membership in a traditional fuzzy set. The fuzziness of membership, i.e., the representation of membership as a value in the range \([0,1]\), is retained in complex fuzzy sets through the amplitude of the grade of membership, \( r_A(x) \). The novelty of complex fuzzy sets is manifested in the additional dimension of membership: the phase of the grade of membership, \( \omega_A(x) \). The properties of membership phase are discussed at length in this section. Ramot et al. [8, 9] extended the range of membership to “unit circle in the complex plane”, unlike others who limited the range to \([0,1]\). Omar [10] studied similarity measures between temporal intuitionistic fuzzy sets. As the complex fuzzy membership grade is two-dimensional (amplitude and phase), a complex fuzzy set can
be visually represented by a three-dimensional graph where the universe of discourse is the third axis. Figure 1 shows the complex fuzzy set.

We divide the paper into four main sections. In the first section preliminaries and basic definitions, we provide some details about the complex fuzzy sets. In the second section, detail is given about the complex version of temporal complex intuitionistic fuzzy set, which is an extension of complex intuitionistic fuzzy set by adding the times and studied the correlation coefficient between two temporal complex intuitionistic fuzzy set. In the third section, details is given about similarity measures between other extensions of temporal complex intuitionistic fuzzy set and extend the method proposed by Chaira [12] for intuitionistic fuzzy set based on the Sugeno [13] and Omar [10] intuitionistic fuzzy generator. In the fourth section, we give application in pattern recognition, medical diagnosis, and topology.

2. Preliminaries and Basic Definitions

Definition 1 (see [8]). A complex fuzzy set (CFS) \( A \) defined on a universe \( X \) is an object of the form \( A \) defined on a universe of discourse \( X \) which is an object of the form

\[
A = \{ x, \mu_A(x) e^{i \alpha_A(x)} : x \in X \},
\]

where \( i = \sqrt{-1}, \mu_A(x) \in [0, 1], \) and \( 0 \leq \omega_A(x) \leq 2\pi. \)

Definition 2 (see [2]). A complex intuitionistic fuzzy set (CIFS) \( A \) defined on a universe of discourse \( X \) is an object of the form

\[
A = \{ x, \mu_A(x) e^{i \alpha_A(x)}, \nu_A(x) e^{i \beta_A(x)} : x \in X \},
\]

where \( i = \sqrt{-1}, \mu_A(x), \nu_A(x) \in [0, 1], \alpha_A(x), \beta_A(x) \in [0, 2\pi], \) and \( 0 \leq \mu_A(x) + \nu_A(x) \leq 1. \)

Definition 3 (see [14]). Let \( A \) and \( B \) be two CIFSs in \( X, \)

\[
A = \{ x, \mu_A(x) e^{i \alpha_A(x)}, \nu_A(x) e^{i \beta_A(x)} : x \in X \}
\]

\[
B = \{ x, \mu_B(x) e^{i \alpha_B(x)}, \nu_B(x) e^{i \beta_B(x)} : x \in X \}.
\]

Then, \( A \cup B \) is given as

\[
A \cup B = \{ x, \mu_{A\cup B}(x) e^{i \alpha_{A\cup B}(x)}, \nu_{A\cup B}(x) e^{i \beta_{A\cup B}(x)} : x \in X \}
\]

where

\[
\mu_{A\cup B}(x) = \max\{ \mu_A(x), \mu_B(x) \},
\]

\[
\nu_{A\cup B}(x) = \max\{ \nu_A(x), \nu_B(x) \}.
\]

Then, for all \( x \in X, \)

(1) \( A \subset B \) if and only if \( \mu_A(x) < \mu_B(x); \nu_A(x) > \nu_B(x). \)

For amplitude terms and \( \alpha_A(x) < \alpha_B(x), \beta_A(x) > \beta_B(x) \) for phase terms.

(2) \( A = B \) if and only if \( \mu_A(x) = \mu_B(x); \nu_A(x) = \nu_B(x). \)

For amplitude terms and \( \alpha_A(x) = \alpha_B(x), \beta_A(x) = \beta_B(x) \) for phase terms.

Definition 4 (see [15]). Let \( A \) and \( B \) be two CIF-sets in \( X, \)

\[
A = \{ x, \mu_A(x) e^{i \alpha_A(x)}, \nu_A(x) e^{i \beta_A(x)} : x \in X \}
\]

\[
B = \{ x, \mu_B(x) e^{i \alpha_B(x)}, \nu_B(x) e^{i \beta_B(x)} : x \in X \}.
\]

Then, for all \( x \in X, \)

(1) \( A \cap B = \{ x, \mu_{A\cap B}(x) e^{i \alpha_{A\cap B}(x)}, \nu_{A\cap B}(x) e^{i \beta_{A\cap B}(x)} : x \in X \}
\]

(2) \( A \cap B = \{ x, \mu_{A\cap B}(x) e^{i \alpha_{A\cap B}(x)}, \nu_{A\cap B}(x) e^{i \beta_{A\cap B}(x)} : x \in X \}. \)
where
\[\mu_{\mathcal{A}\cap\mathcal{B}}(x) e^{i\alpha_{\mathcal{A}\cap\mathcal{B}}(x)} = \left\{ x, \left[ \mu_{\mathcal{A}}(x) \wedge \nu_{\mathcal{B}}(x) \right] e^{i\alpha_{\mathcal{A}\cap\mathcal{B}}(x)} \right\},\]

\[\nu_{\mathcal{A}\cup\mathcal{B}}(x) e^{i\beta_{\mathcal{A}\cup\mathcal{B}}(x)} = \left\{ x, \left[ \nu_{\mathcal{A}}(x) \vee \nu_{\mathcal{B}}(x) \right] e^{i\beta_{\mathcal{A}\cup\mathcal{B}}(x)} \right\},\]

(8)

(2) The complex fuzzy complement of \( \mathcal{A} \), denoted by \( \overline{\mathcal{A}} \), is specified by a function
\[\overline{\mathcal{A}} = \left\{ x, (1 - \mu_{\mathcal{A}}(x)) e^{i(2\pi - \alpha_{\mathcal{A}}(x))}, (1 - \nu_{\mathcal{B}}(x)) e^{i(2\pi - \beta_{\mathcal{B}}(x))} : x \in X \right\},\]

(9)

(3) \( \overline{\mathcal{I}} = \{ x, (e^{i2\pi}, 0) : x \in X \} \) and \( \overline{\partial} = \{ x, (0, e^{i2\pi}) : x \in X \} \).

**Example 6.** Consider \( X = \{ a, b, c, d \} \). Let \( \mathcal{A}, \mathcal{B} \) be a CIFS-subset of \( X \), as given by
\[
\mathcal{A} = \left( \frac{0.2 e^{i0.3\pi}}{a}, \frac{0.4 e^{i0.5\pi}}{c}, \frac{1.0 e^{i1.5\pi}}{d}, \frac{0.0 e^{i0.5\pi}}{b} \right),
\]
\[
\mathcal{B} = \left( \frac{0.7 e^{i0.3\pi}}{a}, \frac{0.2 e^{i1.5\pi}}{c}, \frac{0.8 e^{i1.1\pi}}{d}, \frac{0.1 e^{i0.7\pi}}{b} \right),
\]

(10)

Then
\[\mathcal{A} \cap \mathcal{B} = \left( \frac{0.1 e^{i0.1\pi}}{a}, \frac{0.1 e^{i1.5\pi}}{c}, \frac{0.3 e^{i0.9\pi}}{d}, \frac{0.3 e^{i0.9\pi}}{b} \right),\]

\[\overline{\mathcal{A}} = \left( \frac{0.8 e^{i0.7\pi}}{a}, \frac{0.6 e^{i0.5\pi}}{c}, \frac{0.1 e^{i1.5\pi}}{d}, \frac{0.0 e^{i0.7\pi}}{b} \right),\]

\[\mathcal{B} \cup \mathcal{B} = \left( \frac{0.2 e^{i1.3\pi}}{a}, \frac{0.3 e^{i0.4\pi}}{c}, \frac{1.0 e^{i1.5\pi}}{d}, \frac{0.0 e^{i0.5\pi}}{b} \right),\]

\[\overline{\mathcal{I}} = \left( \frac{0.7 e^{i0.3\pi}}{a}, \frac{0.2 e^{i0.5\pi}}{c}, \frac{0.8 e^{i1.1\pi}}{d}, \frac{0.1 e^{i0.5\pi}}{b} \right),\]

And \( \mathcal{A} \notin \mathcal{B} \) and \( \mathcal{A} \neq \mathcal{B} \).

**3. Temporal Complex Intuitionistic Fuzzy Set**

**Definition 7.** Let \( X \) be a universe, \( T \) be a nonempty set of time moments, and \( \mathcal{A} \subseteq X \). A temporal complex intuitionistic fuzzy set (TCIFS) \( \mathcal{A} \) defined on a universe of discourse \( X \) is an object of the form
\[\mathcal{A}(T) = \left\{ \left( (x, t), \mu_{\mathcal{A}}(x, t) e^{i\alpha_{\mathcal{A}}(x, t)}, \nu_{\mathcal{A}}(x, t) e^{i\beta_{\mathcal{A}}(x, t)} \right) \right\},\]

(12)

where \( \mu_{\mathcal{A}} : \mathcal{A} \times T \rightarrow [0,1] \) and \( \nu_{\mathcal{A}} : \mathcal{A} \times T \rightarrow [0,1] \) such that \( 0 \leq \mu_{\mathcal{A}}(x, t) e^{i\alpha_{\mathcal{A}}(x, t)} + \nu_{\mathcal{A}}(x, t) e^{i\beta_{\mathcal{A}}(x, t)} \leq 1 \), \( \mu_{\mathcal{A}}(x, t) \), and \( \nu_{\mathcal{A}}(x, t) \) being the degrees of membership and nonmembership, respectively, of the element \( x \in X \) at the moment \( t \in T \). And \( \alpha_{\mathcal{A}}(x, t), \beta_{\mathcal{A}}(x, t) \in [0, 2\pi] \) at the moment \( t \in T \), where \( i = \sqrt{-1} \).

The hesitation degree of a TCIFS \( \mathcal{A} \) is defined by
\[\pi_{\mathcal{A}} : \mathcal{A} \times T \rightarrow [0,1] \text{ such that } 0 \leq \pi_{\mathcal{A}}(x, t) \leq 1 \text{ for each } (x, t) \in \mathcal{A} \times T.\]

For brevity we will write \( \mathcal{A} \) instead of \( \mathcal{A}(T) \) when this does not cause confusion.

**Example 8.** Suppose that \( X \) is a universal, with respect to the time set \( T_1 = \{ t \mid t = 6k, k = 0, 1, \ldots, 100 \} \), \( T_2 = \{ t \mid t = 2k, k = 0, 1, \ldots, 100 \} \), and \( \alpha_{\mathcal{A}}(x, t) = \beta_{\mathcal{A}}(x, t) = e^{i2\pi} \). Then, TCIFSs \( \mathcal{A}, \mathcal{B} \) are defined by
\[
\mathcal{A}(T_1)(x) = \begin{cases} \frac{1}{2}, & t = 4k, k = 1, \ldots, 100 \\ \frac{1}{3}, & t = 4k + 2, k = 1, \ldots, 100 \\ \frac{3}{5}, & t = 3k + 1, k = 1, \ldots, 100 \\ \frac{7}{11}, & t = 3k + 2, k = 1, \ldots, 100 \end{cases}
\]

\[
\mathcal{B}(T_2)(x) = \begin{cases} \frac{2}{3}, & t = 3k, k = 1, \ldots, 100 \\ \frac{3}{5}, & t = 3k + 1, k = 1, \ldots, 100 \\ \frac{7}{11}, & t = 3k + 2, k = 1, \ldots, 100 \end{cases}
\]

**Example 9.** Suppose \( X = \{ x_1, x_2, x_3 \} \) with respect to the time set \( T = \{ t_1, t_2, t_3 \} \). Then, the details of a TCIFS \( \mathcal{A} \) are explained in Tables 1, 2, and 3.

**Definition 10.** Let \( \mathcal{A}(T_1) \) and \( \mathcal{B}(T_2) \) be two TCIFSs. Then
\[
\mathcal{A}(T_1) \cap \mathcal{B}(T_2) = \left\{ (x, \min \left( \mu_{\mathcal{A}}(x, t) e^{i\alpha_{\mathcal{A}}(x, t)}, \mu_{\mathcal{B}}(x, t) e^{i\alpha_{\mathcal{B}}(x, t)} \right), \right\},
\]

\[
\mathcal{A}(T_1) \cup \mathcal{B}(T_2) = \left\{ (x, \max \left( \nu_{\mathcal{A}}(x, t) e^{i\beta_{\mathcal{A}}(x, t)}, \nu_{\mathcal{B}}(x, t) e^{i\beta_{\mathcal{B}}(x, t)} \right), \right\}.
\]

\[
\mathcal{A}(T_1) \cup \mathcal{B}(T_2) = \left\{ (x, \min \left( \nu_{\mathcal{A}}(x, t) e^{i\beta_{\mathcal{A}}(x, t)}, \nu_{\mathcal{B}}(x, t) e^{i\beta_{\mathcal{B}}(x, t)} \right), \right\}.
\]
We define the following two operators $f$ and $g$ over a TIFS $\mathcal{A}$:

$$f(\mathcal{A}(T)) = \left\{ (x, \left\{ x \left( \max_{t \in T} \mu_{\mathcal{A}}(x, t) e^{i\alpha_{\mathcal{A}}(x,t)} \right), \min_{t \in T} \nu_{\mathcal{A}}(x, t) e^{i\beta_{\mathcal{A}}(x,t)} \right\} : (x, t) \in X \times T \right\}$$

$$g(\mathcal{A}(T)) = \left\{ (x, \left\{ x \left( \min_{t \in T} \mu_{\mathcal{A}}(x, t) e^{i\alpha_{\mathcal{A}}(x,t)} \right), \max_{t \in T} \nu_{\mathcal{A}}(x, t) e^{i\beta_{\mathcal{A}}(x,t)} \right\} : (x, t) \in X \times T \right\}$$

$$h(\mathcal{A}(T)) = \left\{ (x, \left\{ x \left( \max_{t \in X} \mu_{\mathcal{A}}(x, t) e^{i\alpha_{\mathcal{A}}(x,t)} \right), \min_{t \in X} \nu_{\mathcal{A}}(x, t) e^{i\beta_{\mathcal{A}}(x,t)} \right\} : (x, t) \in X \times T \right\}$$

Table 1: TIFS $\mathcal{A}$.

<table>
<thead>
<tr>
<th>$x_1$</th>
<th>$x_2$</th>
<th>$x_3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$(0.2, 0.1)$</td>
<td>$(0.1, 0.6)$</td>
<td>$(0.3, 0.5)$</td>
</tr>
<tr>
<td>$(0.6, 0.1)$</td>
<td>$(0.1, 0.9)$</td>
<td>$(0.6, 0.4)$</td>
</tr>
<tr>
<td>$(0.7, 0.1)$</td>
<td>$(0.1, 0.7)$</td>
<td>$(0.8, 0.5)$</td>
</tr>
</tbody>
</table>

Table 2: $\alpha_{\mathcal{A}}(x, t) = \beta_{\mathcal{A}}(x, t)$.

<table>
<thead>
<tr>
<th>$x_1$</th>
<th>$x_2$</th>
<th>$x_3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$e^{i(3\pi/2)}$</td>
<td>$e^{i(\pi/2)}$</td>
<td>$e^{i(2\pi/3)}$</td>
</tr>
<tr>
<td>$e^{i(2\pi)}$</td>
<td>$e^{i(3\pi/2)}$</td>
<td>$e^{i(\pi/2)}$</td>
</tr>
<tr>
<td>$e^{i(\pi/2)}$</td>
<td>$e^{i(3\pi/2)}$</td>
<td>$e^{i(\pi/2)}$</td>
</tr>
</tbody>
</table>

Table 3: TCIFS $\mathcal{A}$.

<table>
<thead>
<tr>
<th>$x_1$</th>
<th>$x_2$</th>
<th>$x_3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$(0.2, 0.1i)$</td>
<td>$(0.1i, 0.9i)$</td>
<td>$(0.6i, 0.4i)$</td>
</tr>
<tr>
<td>$(0.6, 0.1)$</td>
<td>$(0.1i, 0.9i)$</td>
<td>$(0.6i, 0.4i)$</td>
</tr>
<tr>
<td>$(-0.7i, -0.1i)$</td>
<td>$(-0.1i, -0.7i)$</td>
<td>$(0.8i, 0.5i)$</td>
</tr>
</tbody>
</table>

**Theorem 12.** $f(\mathcal{A}(T))$ and $g(\mathcal{A}(T))$ are TCIFSs.

**Proof.** Suppose that

$$\max_{t \in T} \mu_{\mathcal{A}}(x, t) e^{i\alpha_{\mathcal{A}}(x,t)} = \mu_{\mathcal{A}}(x, t_1) e^{i\alpha_{\mathcal{A}}(x,t_1)},$$

for some $t_1 \in T$, and

$$\min_{t \in T} \nu_{\mathcal{A}}(x, t) e^{i\beta_{\mathcal{A}}(x,t)} = \nu_{\mathcal{A}}(x, t_2) e^{i\beta_{\mathcal{A}}(x,t_2)},$$

for some $t_2 \in T$.

Then $f(\mathcal{A}(T))$ is TCIFSs. Also, by the same fashion $g(\mathcal{A}(T))$ are TCIFSs.

**Definition II.** We define the following two operators $f$ and $g$ over a TIFS $\mathcal{A}$:

$\overline{\mathcal{A}}(T_1) = \left\{ (x, \left( \overline{\mu_{\mathcal{A}}}(x, t) e^{i(2\pi-\beta_{\mathcal{A}}(x,t))}, \overline{\nu_{\mathcal{A}}}(x, t) e^{i(\alpha_{\mathcal{A}}(x,t))} \right) : (x, t) \in X \times T_1 \right\},$

(14)

where

$$\overline{\mu_{\mathcal{A}}}(x, t) = \begin{cases} \mu_{\mathcal{A}}(x, t) & t \in T_1 \\ 0 & t \in T_2 - T_1 \end{cases}$$

$$\overline{\nu_{\mathcal{A}}}(x, t) = \begin{cases} \nu_{\mathcal{A}}(x, t) & t \in T_1 \\ 1 & t \in T_2 - T_1 \end{cases}$$

$$\overline{\mu_{\mathcal{A}}}(x, t) = \begin{cases} \mu_{\mathcal{A}}(x, t) & t \in T_1 \\ 0 & t \in T_2 - T_1 \end{cases}$$

$$\overline{\nu_{\mathcal{A}}}(x, t) = \begin{cases} \nu_{\mathcal{A}}(x, t) & t \in T_1 \\ 1 & t \in T_2 - T_1 \end{cases}$$

**Theorem 13.** For every TCIFS $\mathcal{A}(T)$,

(1) $f(f(\mathcal{A}(T))) = f(\mathcal{A}(T))$;

(2) $g(g(\mathcal{A}(T))) = g(\mathcal{A}(T))$;

(3) $f(g(\mathcal{A}(T))) = g(\mathcal{A}(T))$;

(4) $g(f(\mathcal{A}(T))) = f(\mathcal{A}(T))$.

**Proof.** The proof is obvious.
Proof. (1)

\[
\begin{align*}
 h(f(\mathcal{A}(T))) &= \left\{ (x, t) \left\{ x, \max_{t \in X} \max_{t \in T} \mu_{\mathcal{A}}(x, t) \right\} \right\} : (x, t) \\
 &\in X \times T \left\{ (x, t) \left\{ x, \min_{t \in T} \min_{t \in T} \mu_{\mathcal{A}}(x, t) e^{\beta t(x,t)} \right\} \right\} : (x, t) \in X \times T \\
 = f(h(\mathcal{A}(T)))
\end{align*}
\]

(2) By the same fashion, one has the following. \(\square\)

Definition 15. Let \(\mathcal{A}\) and \(\mathcal{B}\) be two TCIFSs defined on the universe of discourse \(X = \{x_1, x_2, x_3, \ldots, x_n\}\) and the time moments \(T = \{t_1, t_2, t_3, \ldots, t_m\}\). The correlation coefficient of \(\mathcal{A}\) and \(\mathcal{B}\) is given by

\[
k(\mathcal{A}, \mathcal{B}) = \frac{C(\mathcal{A}, \mathcal{B})}{\sqrt{T(\mathcal{A})T(\mathcal{B})}}
\]

where

\[
C(\mathcal{A}, \mathcal{B}) = \sum_{i=1}^{n} \sum_{j=1}^{m} (\mu_{\mathcal{A}}(x_i, t_j) e^{\alpha_{\mathcal{A}}(x_i,t_j)} \mu_{\mathcal{B}}(x_i, t_j) e^{\beta_{\mathcal{B}}(x_i,t_j)}) + \nu_{\mathcal{A}}(x_i, t_j) \nu_{\mathcal{B}}(x_i, t_j) e^{\gamma_{\mathcal{B}}(x_i,t_j)}
\]

is the correlation of two TCIFSs \(\mathcal{A}\) and \(\mathcal{B}\), and

\[
T(\mathcal{A}) = \sum_{i=1}^{n} \sum_{j=1}^{m} (\mu_{\mathcal{A}}^2(x_i, t_j) e^{2\alpha_{\mathcal{A}}(x_i,t_j)} + \nu_{\mathcal{A}}^2(x_i, t_j) e^{2\gamma_{\mathcal{A}}(x_i,t_j)})
\]

\[
T(\mathcal{B}) = \sum_{i=1}^{n} \sum_{j=1}^{m} (\mu_{\mathcal{B}}^2(x_i, t_j) e^{2\beta_{\mathcal{B}}(x_i,t_j)} + \nu_{\mathcal{B}}^2(x_i, t_j) e^{2\gamma_{\mathcal{B}}(x_i,t_j)})
\]

are the information temporal complex intuitionistic energies of \(\mathcal{A}\) and \(\mathcal{B}\), respectively.

Example 16. Suppose that \(X = \{x_1, x_2, x_3\}\) with respect to the time set \(T = \{t_1, t_2, t_3\}\). The details of a TCIFS \(\mathcal{A}(T)\) are explained in Table 4, Table 5 explained TCIFS \(\mathcal{B}(T)\), and Table 6 explained the correlation coefficient \(k(\mathcal{A}, \mathcal{B})\) between TCIFS \(\mathcal{A}(T)\) and TCIFS \(\mathcal{B}(T)\).

<table>
<thead>
<tr>
<th>(x)</th>
<th>(t_1)</th>
<th>(t_2)</th>
<th>(t_3)</th>
</tr>
</thead>
<tbody>
<tr>
<td>(x_1)</td>
<td>(0.21, 0.31)</td>
<td>(-0.11, -0.31)</td>
<td>(0.31, 0.51)</td>
</tr>
<tr>
<td>(x_2)</td>
<td>(0.6, 0.11)</td>
<td>(0.11, 0.91)</td>
<td>(0.61, 0.41)</td>
</tr>
<tr>
<td>(x_3)</td>
<td>(-0.71, -0.11)</td>
<td>(-0.81, -0.11)</td>
<td>(0.81, 0.51)</td>
</tr>
<tr>
<td>(x_4)</td>
<td>(0.6, 0.11)</td>
<td>(0.11, 0.91)</td>
<td>(0.61, 0.41)</td>
</tr>
<tr>
<td>(x_5)</td>
<td>(0.21, 0.11)</td>
<td>(-0.1, -0.6)</td>
<td>(0.31, 0.51)</td>
</tr>
<tr>
<td>(x_6)</td>
<td>(0.61, 0.11)</td>
<td>(0.11, 0.91)</td>
<td>(0.61, 0.46)</td>
</tr>
</tbody>
</table>

Proof. Let \(t \in T_1\). From Definition 10, \(\overline{\mu}_\mathcal{A}(x, t) = \mu_{\mathcal{A}}(x, t)\), \(\overline{\nu}_\mathcal{A}(x, t) = \nu_{\mathcal{A}}(x, t)\), \(\overline{\mu}_\mathcal{B}(x, t) = \mu_{\mathcal{B}}(x, t)\), \(\overline{\nu}_\mathcal{B}(x, t) = \nu_{\mathcal{B}}(x, t)\), and then

(1)

\[
T(\mathcal{A}) = \sum_{i=1}^{m} \sum_{j=1}^{m} (\mu_{\mathcal{A}}^2(x_i, t_j) e^{2\alpha_{\mathcal{A}}(x_i,t_j)} + \nu_{\mathcal{A}}^2(x_i, t_j) e^{2\gamma_{\mathcal{A}}(x_i,t_j)})
\]

(2) If \(\alpha_{\mathcal{A}}(x, t) = 2\pi\) and \(\beta_{\mathcal{A}}(x, t) = 2\pi\), then \(C(\mathcal{A}, \mathcal{A}) = T(\mathcal{A})\); and

(3) \(C(\mathcal{A}, \mathcal{B}) = C(\mathcal{B}, \mathcal{A})\).
\[ k(\mathcal{A}, \mathcal{B}) = \frac{C(\mathcal{A}, \mathcal{B})}{\sqrt{T(\mathcal{A}) T(\mathcal{B})}} \] (27)

If \( \mathcal{A} = \mathcal{B} \), then \( \sqrt{T(\mathcal{A}) T(\mathcal{B})} = \sqrt{T(\mathcal{A})^2} = T(\mathcal{A}) \). Then from Proposition 17 (2), one has the following.

If \( \alpha_{\mathcal{A}}(x, t) = 2\pi \) and \( \beta_{\mathcal{A}}(x, t) = 2\pi \), then \( C(\mathcal{A}, \mathcal{A}) = T(\mathcal{A}) \); then

\[ k(\mathcal{A}, \mathcal{A}) = \frac{T(\mathcal{A})}{T(\mathcal{A})} = 1 \] (28)

(2) From Proposition 17 (3), \( C(\mathcal{A}, \mathcal{B}) = C(\mathcal{B}, \mathcal{A}) \). Then

\[ k(\mathcal{A}, \mathcal{B}) = \frac{C(\mathcal{B}, \mathcal{A})}{\sqrt{T(\mathcal{B}) T(\mathcal{A})}} = k(\mathcal{B}, \mathcal{A}) \] (29)

(3) We will prove that \( k(\mathcal{A}, \mathcal{B}) < 1 \) such that it is evident \( 0 < k(\mathcal{A}, \mathcal{B}) < 1 \), so suppose that

\[ \sum_{i=1}^{n} \sum_{j=1}^{m} \mu_{\mathcal{A}}^2(x_i, t_j) e^{i(2\pi - \beta_{\mathcal{A}}(x_i, t_j))} = \alpha_1, \]

\[ \sum_{i=1}^{n} \sum_{j=1}^{m} \mu_{\mathcal{B}}^2(x_i, t_j) e^{i(2\pi - \beta_{\mathcal{B}}(x_i, t_j))} = \alpha_2, \]

\[ \sum_{i=1}^{n} \sum_{j=1}^{m} \nu_{\mathcal{A}}^2(x_i, t_j) e^{i(2\pi - \beta_{\mathcal{A}}(x_i, t_j))} = \alpha_3, \]

\[ \sum_{i=1}^{n} \sum_{j=1}^{m} \nu_{\mathcal{B}}^2(x_i, t_j) e^{i(2\pi - \beta_{\mathcal{B}}(x_i, t_j))} = \alpha_4. \] (30)

Then

\[ T(\mathcal{A}) T(\mathcal{B}) = \left( \sum_{i=1}^{n} \sum_{j=1}^{m} \mu_{\mathcal{A}}^2(x_i, t_j) \right) \cdot e^{i(2\pi - \beta_{\mathcal{A}}(x_i, t_j))} + \nu_{\mathcal{A}}^2(x_i, t_j) e^{i(2\pi - \beta_{\mathcal{A}}(x_i, t_j))} = \left( \sum_{i=1}^{n} \sum_{j=1}^{m} \mu_{\mathcal{B}}^2(x_i, t_j) \right) \cdot e^{i(2\pi - \beta_{\mathcal{B}}(x_i, t_j))} + \nu_{\mathcal{B}}^2(x_i, t_j) e^{i(2\pi - \beta_{\mathcal{B}}(x_i, t_j))} \] (31)

Then

\[ C(\mathcal{A}, \mathcal{B}) = (\alpha_1 + \alpha_3)^{1/2} \times (\alpha_2 + \alpha_4)^{1/2} \] (32)

\[ C(\mathcal{A}, \mathcal{B}) = (\alpha_1 + \alpha_3)^{1/2} + (\alpha_2 + \alpha_4)^{1/2} \] (33)

Then

\[ C(\mathcal{A}, \mathcal{B}) = (\alpha_1 \times \alpha_2)^{1/2} + (\alpha_3 \times \alpha_4)^{1/2} \] (34)
\[ k^2 (\mathcal{A}, \mathcal{B}) \leq \frac{\alpha_1 \times \alpha_2 + 2 (\alpha_1 \times \alpha_3 \times \alpha_4)^{1/2} + \alpha_3 \times \alpha_4}{(\alpha_1 + \alpha_3) \times (\alpha_2 + \alpha_4)} \quad (34) \]

But

\[ k^2 (\mathcal{A}, \mathcal{B}) - 1 \leq \frac{\alpha_1 \times \alpha_2 + 2 (\alpha_1 \times \alpha_3 \times \alpha_4)^{1/2} + \alpha_3 \times \alpha_4 - (\alpha_1 + \alpha_3) \times (\alpha_2 + \alpha_4)}{(\alpha_1 + \alpha_3) \times (\alpha_2 + \alpha_4)} - 1 = \frac{[\alpha_1 \times \alpha_4]^{1/2} - (\alpha_2 \times \alpha_3)^{1/2} - 1}{(\alpha_1 + \alpha_3) \times (\alpha_2 + \alpha_4)} \leq 0 \]

Hence \( k^2 (\mathcal{A}, \mathcal{B}) - 1 \leq 0 \), and then \( k (\mathcal{A}, \mathcal{B}) \leq 1 \). \( \square \)

**Definition 19.** Let \( S : \text{TCIFSs} (X, T) \times \text{TCIFSs} (X, T) \rightarrow [0, 1] \) be a function, and let \( \mathcal{A}(T), \mathcal{B}(T) \) and \( C(T) \) be TCIFSs in the universal \( X = \{x_1, x_2, x_3, \ldots, x_n\} \) with respect to the time set \( T = \{t_1, t_2, t_3, \ldots, t_m\} \). Then \( S(\mathcal{A}, \mathcal{B}) \) is said to be the similarity degree between \( \text{TCIFSs} \mathcal{A} \) and \( \mathcal{B} \), satisfying the following statements:

1. \( 0 \leq S(\mathcal{A}, \mathcal{B}) \leq 1 \).
2. \( S(\mathcal{A}, \mathcal{B}) = 1 \) if \( \mathcal{A} = \mathcal{B} \).
3. \( S(\mathcal{A}, \mathcal{B}) = S(\mathcal{B}, \mathcal{A}) \).
4. If \( \mathcal{A} \subseteq \mathcal{B} \subseteq C \). Then \( S(\mathcal{A}, C) \leq S(\mathcal{A}, \mathcal{B}), S(\mathcal{A}, C) \leq S(\mathcal{B}, C) \).

Let

\[ S_1 (\mathcal{A}, \mathcal{B}) = 1 - \frac{1}{2mn} \sum_{i=1}^{n} \sum_{j=1}^{m} |S_{d\mathcal{A}}(i, j) - S_{d\mathcal{B}}(i, j)|, \]

\[ S_2 (\mathcal{A}, \mathcal{B}) = 1 - \frac{1}{2mn} \sum_{i=1}^{n} \sum_{j=1}^{m} \left| \mu_{d\mathcal{A}}(x_i, t_j) e^{\alpha_{d\mathcal{A}}(x_i,t_j)} - \mu_{d\mathcal{B}}(x_i, t_j) e^{\alpha_{d\mathcal{A}}(x_i,t_j)} \right| + \left| \nu_{d\mathcal{A}}(x_i, t_j) e^{\beta_{d\mathcal{A}}(x_i,t_j)} - \nu_{d\mathcal{B}}(x_i, t_j) e^{\beta_{d\mathcal{A}}(x_i,t_j)} \right|, \]

\[ S_0 (\mathcal{A}, \mathcal{B}) = 1 - \frac{1}{2mn} \sum_{i=1}^{n} \sum_{j=1}^{m} \left( (\mu_{d\mathcal{A}}(x_i, t_j) e^{\alpha_{d\mathcal{A}}(x_i,t_j)} - \mu_{d\mathcal{B}}(x_i, t_j) e^{\alpha_{d\mathcal{A}}(x_i,t_j)})^2 + (\nu_{d\mathcal{A}}(x_i, t_j) e^{\beta_{d\mathcal{A}}(x_i,t_j)} - \nu_{d\mathcal{B}}(x_i, t_j) e^{\beta_{d\mathcal{A}}(x_i,t_j)})^2 \right) \]

\[ S_3 (\mathcal{A}, \mathcal{B}) = 1 - \frac{1}{\sqrt{2}mn} \left( \sum_{i=1}^{m} \sum_{j=1}^{n} \left| \psi_{d\mathcal{A}}(i, j) - \psi_{d\mathcal{B}}(i, j) \right| \right)^y, \quad 1 \leq y < +\infty, \]

\[ S_4 (\mathcal{A}, \mathcal{B}) = \frac{\sum_{i=1}^{n} \sum_{j=1}^{m} \max \left( \left| \mu_{d\mathcal{A}}(x_i, t_j) e^{\alpha_{d\mathcal{A}}(x_i,t_j)} \right|, \left| \mu_{d\mathcal{B}}(x_i, t_j) e^{\alpha_{d\mathcal{A}}(x_i,t_j)} \right| \right) + \min \left( \left| 1 - \psi_{d\mathcal{A}}(i, j) \right|, \left| 1 - \psi_{d\mathcal{B}}(i, j) \right| \right)}{\sum_{i=1}^{n} \sum_{j=1}^{m} \max \left( \left| \mu_{d\mathcal{A}}(x_i, t_j) e^{\alpha_{d\mathcal{A}}(x_i,t_j)} \right|, \left| \mu_{d\mathcal{B}}(x_i, t_j) e^{\alpha_{d\mathcal{A}}(x_i,t_j)} \right| \right) + \min \left( \left| 1 - \psi_{d\mathcal{A}}(i, j) \right|, \left| 1 - \psi_{d\mathcal{B}}(i, j) \right| \right)} \]

From a comparison between similarity measures \( S_0 (\mathcal{A}, \mathcal{B}), S_1 (\mathcal{A}, \mathcal{B}), S_2 (\mathcal{A}, \mathcal{B}), S_3 (\mathcal{A}, \mathcal{B}), S_4 (\mathcal{A}, \mathcal{B}) \), we give the following example.

Example 20. Suppose that \( \mathcal{A}(T) \) and \( \mathcal{B}(T) \) is TCIFSs defined on \( X = \{x_1, x_2, x_3, x_4, x_5, x_6\} \) with respect to the time set \( T = \{t_1, t_2, t_3, t_4, t_5, t_6\} \). The details of a TCIFS \( \mathcal{A}(T) \)
are explained in Table 7, Table 8 explained TCIFS $\mathcal{B}(T)$, and Table 9 explained TCIFS $\mathcal{A}(T)$.

### 4. Similarity Measures between Other Extensions of Temporal Complex Intuitionistic Fuzzy Set


**Definition 21.** If $\mu_{\mathcal{A}}(x, t)e^{i\alpha_{\mathcal{A}}(x,t)}$ is the degrees of membership function of the element $x \in X$ at the moment $t \in T$, then nonmembership function $\nu_{\mathcal{A}}(x)e^{i\beta_{\mathcal{A}}(x)} = G(\mu_{\mathcal{A}}(x, t)e^{i\alpha_{\mathcal{A}}(x,t)})$, where

$$G(\mu_{\mathcal{A}}(x, t)e^{i\alpha_{\mathcal{A}}(x,t)}) = \frac{1 - \mu_{\mathcal{A}}(x, t)e^{i\alpha_{\mathcal{A}}(x,t)}}{1 + \alpha \mu_{\mathcal{A}}(x, t)e^{i\alpha_{\mathcal{A}}(x,t)}},$$  

(38)

$$\alpha > 0$$

And $G(1) = 0, G(0) = 1$, and by help of the Sugeno [6] intuitionistic fuzzy generator, TCIFS $\mathcal{A}$ is given by

$$\mathcal{A}(T) = \left\{ (x, t) \cdot \mu_{\mathcal{A}}(x, t), e^{i\alpha_{\mathcal{A}}(x,t)}, \nu_{\mathcal{A}}(x) \right\} | (x, t) \in X \times T \right\}$$

(39)

The hesitation degree of a TCIFS $\mathcal{A}$ is

$$\pi_{\mathcal{A}}(x, t) = 1 - \frac{\mu_{\mathcal{A}}(x, t) e^{i\alpha_{\mathcal{A}}(x,t)}}{1 + \alpha \mu_{\mathcal{A}}(x, t)e^{i\alpha_{\mathcal{A}}(x,t)}}$$

(40)

**Example 22.** Suppose that $\mathcal{A}(T)$ is TCIFS defined on $X = \{x_1, x_2, x_3\}$ with respect to the time set $T = \{t_1, t_2, t_3\}$. The details of a TCIFS $\mathcal{A}(T)$ are explained in Tables 10, 11, and 12. Table 13 explained TCIFS $\mathcal{B}(T)$ when $\alpha = 1$, and Table 14 explained the hesitation degree of a TCIFS $\mathcal{A}$. If $\alpha = 1$, then one has the following (see Tables 13 and 14).

**Definition 23.** Suppose that $\mathcal{A}(T)$ and $\mathcal{B}(T)$ are TCIFSs in the universal $X = \{x_1, x_2, x_3, \ldots, x_n\}$ with respect to the time set $T = \{t_1, t_2, t_3, \ldots, t_m\}$. Then a cosine similarity measure between $\mathcal{A}(T)$ and $\mathcal{B}(T)$ is proposed as follows:

$$C_T(\mathcal{A}\left(x_i, t_j\right), \mathcal{B}\left(x_i, t_j\right))$$

$$= \frac{1}{mn} \sum_{i=1}^{n} \sum_{j=1}^{m} \mu_{\mathcal{A}}(x_i, t_j)e^{i\alpha_{\mathcal{A}}(x_i,t_j)} \mu_{\mathcal{B}}(x_i, t_j)e^{i\alpha_{\mathcal{B}}(x_i,t_j)} + \nu_{\mathcal{A}}(x_i, t_j)e^{i\beta_{\mathcal{A}}(x_i)} + \nu_{\mathcal{B}}(x_i, t_j)e^{i\beta_{\mathcal{B}}(x_i)} \sqrt{\mu_{\mathcal{A}}^2(x_i,t_j)e^{i(2\alpha-\alpha_{\mathcal{A}}(x_i))} + \nu_{\mathcal{A}}^2(x_i,t_j)e^{i(2\beta-\beta_{\mathcal{A}}(x_i))}} \sqrt{\mu_{\mathcal{B}}^2(x_i,t_j)e^{i(2\alpha-\alpha_{\mathcal{B}}(x_i))} + \nu_{\mathcal{B}}^2(x_i,t_j)e^{i(2\beta-\beta_{\mathcal{B}}(x_i))}}$$

(41)

where $i = 1, 2, 3, \ldots, n; j = 1, 2, 3, \ldots, m$.
Table 9: A comparison between similarity measures $S_0(\mathbf{A}, \mathbf{B}), S_1(\mathbf{A}, \mathbf{B}), S_2(\mathbf{A}, \mathbf{B}), S_3(\mathbf{A}, \mathbf{B}), S_4(\mathbf{A}, \mathbf{B})$.

<table>
<thead>
<tr>
<th>$t_1$</th>
<th>$t_2$</th>
<th>$t_3$</th>
<th>$t_4$</th>
<th>$t_5$</th>
<th>$t_6$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$S_0(\mathbf{A}, \mathbf{B})$</td>
<td>$S_1(\mathbf{A}, \mathbf{B})$</td>
<td>$S_2(\mathbf{A}, \mathbf{B})$</td>
<td>$S_3(\mathbf{A}, \mathbf{B})$</td>
<td>$S_4(\mathbf{A}, \mathbf{B})$</td>
<td></td>
</tr>
<tr>
<td>0.9795</td>
<td>0.9646</td>
<td>1 - 0.235i</td>
<td>0.9795</td>
<td>1.0</td>
<td></td>
</tr>
<tr>
<td>0.9986</td>
<td>0.9938</td>
<td>0.9972</td>
<td>0.9980</td>
<td>0.9980</td>
<td>1.</td>
</tr>
<tr>
<td>1.00</td>
<td>1.0</td>
<td>0.9972</td>
<td>0.9980</td>
<td>0.9980</td>
<td>1.</td>
</tr>
<tr>
<td>0.9941</td>
<td>0.9823</td>
<td>0.9882</td>
<td>0.9916</td>
<td>0.9916</td>
<td>1.</td>
</tr>
<tr>
<td>0.8882</td>
<td>0.9035</td>
<td>0.8589</td>
<td>0.9128</td>
<td>0.9128</td>
<td>1.</td>
</tr>
<tr>
<td>0.9</td>
<td>0.9575</td>
<td>1 - 0.0263i</td>
<td>0.9302</td>
<td>1 - 0.0263i</td>
<td></td>
</tr>
<tr>
<td>0.9944</td>
<td>0.9819</td>
<td>0.9930</td>
<td>0.9980</td>
<td>0.9980</td>
<td>1.</td>
</tr>
<tr>
<td>0.9764</td>
<td>0.9233</td>
<td>0.9705</td>
<td>0.9916</td>
<td>0.9916</td>
<td>1.</td>
</tr>
<tr>
<td>0.7333</td>
<td>0.5474</td>
<td>0.8193</td>
<td>0.7448</td>
<td>0.7448</td>
<td>1.</td>
</tr>
<tr>
<td>1.0</td>
<td>1.0</td>
<td>0.9764</td>
<td>0.9973</td>
<td>0.9973</td>
<td>1.</td>
</tr>
<tr>
<td>0.9302 - 0.0816i</td>
<td>0.8129</td>
<td>1 - 0.0241i</td>
<td>0.9302</td>
<td>1 - 0.0241i</td>
<td></td>
</tr>
<tr>
<td>0.9891</td>
<td>0.9805</td>
<td>0.9896</td>
<td>0.9896</td>
<td>0.9896</td>
<td>1.</td>
</tr>
<tr>
<td>0.9981</td>
<td>0.9981</td>
<td>0.9981</td>
<td>0.9981</td>
<td>0.9981</td>
<td>1.</td>
</tr>
<tr>
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<td>0.9175</td>
<td>0.9941</td>
<td>0.9941</td>
<td>0.9941</td>
<td>1.</td>
</tr>
<tr>
<td>0.87977</td>
<td>0.9137</td>
<td>0.8743</td>
<td>0.8797</td>
<td>0.9137</td>
<td>1.</td>
</tr>
<tr>
<td>0.9288 - 0.0171i</td>
<td>0.9845 + 0.0697i</td>
<td>1 - 0.0331i</td>
<td>0.9288 - 0.0171i</td>
<td>1 - 0.01171</td>
<td></td>
</tr>
<tr>
<td>0.9929</td>
<td>0.9868</td>
<td>0.9944</td>
<td>0.9972</td>
<td>0.9972</td>
<td>1.</td>
</tr>
<tr>
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<td>0.9931</td>
<td>0.9931</td>
<td>0.9931</td>
<td>0.9931</td>
<td>1.</td>
</tr>
<tr>
<td>0.9899</td>
<td>0.9444</td>
<td>0.9764</td>
<td>0.9882</td>
<td>0.9882</td>
<td>1.</td>
</tr>
<tr>
<td>0.6853</td>
<td>0.8502</td>
<td>0.8015</td>
<td>0.8576</td>
<td>0.8576</td>
<td>1.</td>
</tr>
<tr>
<td>0.9288 - 0.0471i</td>
<td>0.8945 + 0.0697i</td>
<td>1 - 0.0331i</td>
<td>0.9288 - 0.0241i</td>
<td>0.9288 - 0.01171</td>
<td></td>
</tr>
<tr>
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<td>0.9868</td>
<td>0.9944</td>
<td>0.9972</td>
<td>0.9972</td>
<td>1.</td>
</tr>
<tr>
<td>0.9944</td>
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<td>0.9944</td>
<td>0.9944</td>
<td>0.9944</td>
<td>1.</td>
</tr>
<tr>
<td>0.9764</td>
<td>0.9444</td>
<td>0.9764</td>
<td>0.9823</td>
<td>0.9823</td>
<td>1.</td>
</tr>
<tr>
<td>0.7507</td>
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<td>0.8015</td>
<td>0.8372</td>
<td>0.8372</td>
<td>1.</td>
</tr>
</tbody>
</table>
Theorem 24. Suppose that \( \mathcal{A}(T) \) and \( \mathcal{B}(T) \) are TCIFSs in the universal \( X \) with respect to the time set \( T \). Then

1. \( C_T(\mathcal{A}, \mathcal{B}) = 1 \) if \( \mathcal{A} = \mathcal{B} \) and \( \alpha_{\mathcal{A}}(x, t) = \beta_{\mathcal{B}}(x, t) = \frac{1}{2\pi} \).

2. \( C_T(\mathcal{A}, \mathcal{B}) = C_T(\mathcal{B}, \mathcal{A}) \).

3. \(-1 \leq C_T(\mathcal{A}, \mathcal{B}) \leq 1\).

4. If \( n = m = 1 \), then \( C_T(\mathcal{A}, \mathcal{B}) = k(\mathcal{A}, \mathcal{B}) \).

Proof. (1) Let \( \mathcal{A} \) and \( \mathcal{B} \) be two TCIFSs defined on the universe of discourse \( X = \{x_1, x_2, x_3, \ldots, x_n\} \) and the time moments \( T = \{t_1, t_2, t_3, \ldots, t_m\} \). The cosine similarity measure between \( \mathcal{A}(T) \) and \( \mathcal{B}(T) \) is given by

\[
C_T(\mathcal{A}, \mathcal{B}) = \frac{1}{mn} \left( \sum_{i=1}^{n} \sum_{j=1}^{m} \mu_{\mathcal{A}}(x_i, t_j) e^{i\alpha_{\mathcal{A}}(x_i, t_j)} \mu_{\mathcal{B}}(x_i, t_j) e^{i\beta_{\mathcal{B}}(x_i, t_j)} + \nu_{\mathcal{A}}(x_i, t_j) e^{i\alpha_{\mathcal{A}}(x_i, t_j)} \nu_{\mathcal{B}}(x_i, t_j) e^{i\beta_{\mathcal{B}}(x_i, t_j)} \right)
\]

If \( \mathcal{A} = \mathcal{B} \) and \( \alpha_{\mathcal{A}}(x, t) = \beta_{\mathcal{B}}(x, t) = \frac{1}{2\pi} \), then

\[
C_T(\mathcal{A}, \mathcal{B}) = 1
\]

(2) By the same way in (3) in Theorem 2.1, one has the following.

(4) If \( n = m = 1 \), then

\[
k(\mathcal{A}, \mathcal{B})
\]
Definition 25. Suppose that \( \mathcal{A}(T) \) and \( \mathcal{B}(T) \) is TCIFSs in the universal \( X = \{x_1, x_2, x_3, \ldots, x_n\} \) with respect to the time set \( T = \{t_1, t_2, t_3, \ldots, t_m\} \). Then, the distance measure of the angle is proposed as follows:

\[
d(\mathcal{A}, \mathcal{B}) = \cos^{-1}(C_T(\mathcal{A}, \mathcal{B}))
\]  

(46)

Theorem 26. Suppose that \( \mathcal{A}(T) \) and \( \mathcal{B}(T) \) are TCIFSs in the universal \( X \) with respect to the time set \( T \).

1. \( C_T(\mathcal{A}, \mathcal{B}) = 1 \) and \( \alpha_{\mathcal{A}}(x, t) = \beta_{\mathcal{A}}(x, t) = 2\pi \); then \( d(\mathcal{A}, \mathcal{B}) = 0 \).

2. \( C_T(\mathcal{A}, \mathcal{B}) = C_T(\mathcal{B}, \mathcal{A}) \), then \( d(\mathcal{A}, \mathcal{B}) = d(\mathcal{B}, \mathcal{A}) \).

3. if \( -1 \leq C_T(\mathcal{A}, \mathcal{B}) \leq 1 \), then \( d(\mathcal{A}, \mathcal{B}) \geq 0 \).

4. if \( \mathcal{A} \subseteq \mathcal{B} \subseteq C \), then \( d(\mathcal{A}, C) \leq d(\mathcal{A}, \mathcal{B}) + d(\mathcal{B}, C) \).

\[
C_T(\mathcal{A}, \mathcal{B}) = \frac{1}{mn} \left[ \sum_{i=1}^{n} \sum_{j=1}^{m} \left( \mu_{\mathcal{A}}(x_i, t_j) e^{i\alpha_{\mathcal{A}}(x_i,t_j)} \mu_{\mathcal{B}}(x_i, t_j) e^{i\beta_{\mathcal{B}}(x_i,t_j)} + \nu_{\mathcal{A}}(x_i, t_j) e^{i\beta_{\mathcal{B}}(x_i,t_j)} \nu_{\mathcal{B}}(x_i, t_j) e^{i\beta_{\mathcal{B}}(x_i,t_j)} \right) \right] \tag{47}
\]

Proof. (1), (2), and (3) are simple proof.

(4) Let \( \mathcal{A}(T), \mathcal{B}(T) \) and \( C(T) \) be TCIFSs in the universal \( X = \{x_1, x_2, x_3, \ldots, x_n\} \) with respect to the time set \( T = \{t_1, t_2, t_3, \ldots, t_m\} \). Then, the distance measure of the angle is proposed as follows:

\[
d_{(i,j)}(\mathcal{A}(x_i, t_j), \mathcal{B}(x_j, t_j)) = \cos^{-1}(C_T(\mathcal{A}(x_i, t_j), \mathcal{B}(x_j, t_j))) \tag{48}
\]

where \( i, j = 1, 2, 3, \ldots, n, j = 1, 2, 3, \ldots, m \), and

If \( \mathcal{A}(x_i, t_j) \subseteq \mathcal{B}(x_j, t_j) \subseteq C(x_i, t_j) \), for each \( i = 1, 2, 3, \ldots, n \), \( j = 1, 2, 3, \ldots, m \), then

\[
d_{(i,j)}(\mathcal{A}(x_i, t_j), \mathcal{B}(x_j, t_j)) + d_{(i,j)}(\mathcal{B}(x_j, t_j), C(x_i, t_j)) \geq d_{(i,j)}(\mathcal{A}(x_i, t_j), C(x_i, t_j)).
\]  

(49)

Definition 27. Let \( \mathcal{A} \) and \( \mathcal{B} \) be two TCIFSs defined on the universe of discourse \( X = \{x_1, x_2, x_3, \ldots, x_n\} \) and the time moments \( T = \{t_1, t_2, t_3, \ldots, t_m\} \). Suppose that \( k(\mathcal{A}, \mathcal{B}) \) is correlation coefficient of \( \mathcal{A} \) and \( \mathcal{B} \). Then a weight similarity measure between TCIFSs \( \mathcal{A}(T) \) and \( \mathcal{B}(T) \) is proposed as follows:

\[
\rho_T(\mathcal{A}(x_i, t_j), \mathcal{B}(x_j, t_j)) = \left( \sum_{i=1}^{n} \sum_{j=1}^{m} k(\mathcal{A}(x_i, t_j), \mathcal{B}(x_j, t_j)) \right)^{-1} \mu_{\mathcal{A}}(x_i, t_j) e^{i\alpha_{\mathcal{A}}(x_i,t_j)} \mu_{\mathcal{B}}(x_j, t_j) e^{i\beta_{\mathcal{B}}(x_j,t_j)} + \nu_{\mathcal{A}}(x_i, t_j) e^{i\beta_{\mathcal{B}}(x_i,t_j)} \nu_{\mathcal{B}}(x_j, t_j) e^{i\beta_{\mathcal{B}}(x_j,t_j)} \right) \sqrt{\mu_{\mathcal{A}}^2(\mathcal{A}(x_i, t_j)) + \mu_{\mathcal{B}}^2(\mathcal{B}(x_j, t_j)) + \nu_{\mathcal{A}}^2(\mathcal{A}(x_i, t_j)) + \nu_{\mathcal{B}}^2(\mathcal{B}(x_j, t_j))} \tag{50}
\]
Remark 28. \(\rho(T) = \{a, b, c, d\}\) is a TIFS defined on \(X = \{x_1, x_2, x_3, x_4, x_5, x_6\}\) with respect to the time set \(T = \{t_1, t_2, t_3, t_4, t_5, t_6\}\). The details of a TIFS \(\mathcal{A}(T)\) are explained in Table 15, Table 16 explained TCIFS \(\mathcal{B}(T)\), and Table 17 explained a comparison between similarity measures between \(C_T(\mathcal{A}, \mathcal{B}), \rho_T(\mathcal{A}, \mathcal{B})\).

4.1. Application in Pattern Recognition and Medical Diagnosis. Let \(L = \{x_1, x_2, x_3, x_4, x_5, x_6\}\) be the set of symptoms of the diseases with respect to the time set \(T = \{t_1, t_2, t_3, t_4, t_5, t_6\}\) and \(L_1\) be the set of diagnoses. By using the similarity measures \(C_T(L, L_1)\) we try to discover that the patient may suffer from one from diseases \(L\) which have symptoms \(x_1\) at the time \(t_1\), and we let \(L\) be standard case symptoms of one of diseases (Table 18 and L1 be any case (Table 19); Table 20 explained the similarity measures \(C_T(L, L_1)\) between a standard case \(L\) and any case \(L_1\).

And we define the symptoms of case by Table 19.

Then Table 20 explained the similarity measures \(C_T(L, L_1)\) between a standard case \(L\) and any case \(L_1\).

When the similarity measures \(-1 \leq C_T(L, L_1) \leq 1\) are small, then probability that the patient is suffering from the disease \(x\) at the time \(t\) is big and the conversely is true.

4.2. Complex Intuitionistic Fuzzy Topology

Definition 30. An intuitionistic complex fuzzy topology on \(X\) is a family \(T\) of CIF-sets in \(X\) which satisfies the following properties:

1. \(\emptyset, X \in T\),
2. if \(\mathcal{A}, \mathcal{B} \in T\), then \(\mathcal{A} \cap \mathcal{B} \in T\),
3. if \(\mathcal{A}_i \in T\) for each \(i \in \Gamma\), then \(\bigcup_{i \in \Gamma} \mathcal{A}_i \in T\).

Then \((X, T)\) is called complex intuitionistic fuzzy topological space. The elements of \(T\) are called CIFO-sets and the complement of the CIFO-set is called CIFC-sets.

Example 31. Consider \(X = \{a, b, c, d\}\). Let \(\mathcal{A}, \mathcal{B}\) be a CIFO-subset of \(X\), as given by

\[
\mathcal{A} = \left(\frac{0.2e^{0.3\pi}, 0.4e^{0.5\pi}}{a}, \frac{1.0e^{0.5\pi}, 0.0e^{0.5\pi}}{b}\right),
\mathcal{B} = \left(\frac{0.7e^{0.3\pi}, 0.2e^{0.5\pi}}{c}, \frac{0.8e^{0.1\pi}, 0.1e^{0.2\pi}}{d}\right)
\]

Then \(\mathcal{A} \cap \mathcal{B}, \mathcal{A} \cup \mathcal{B}\) is an complex intuitionistic fuzzy topology on \(X\).
Definition 32. If $(X, \tau)$ is called complex intuitionistic fuzzy topological space, $A \subseteq X$, then the interior of $A$ is defined as the union of all $CIF$-subsets of $A$ and it is denoted by $A'$. That is, $A'$ is the largest $CIF$-subset of $A$. The closure of $A$ is defined as the intersection of all $CIFC$ sets containing $A$ and it is denoted by $A^-$. That is, $A^-$ is the smallest $CIPC$ set containing $A$.

Example 33. Consider $X = \{a, b\}$. Let $\mathcal{A}, \mathcal{B}$ be a $CIF$-subset of $X$, as given by

$$\mathcal{A} = \left( \frac{0.2e^{10.3\pi}, 0.4e^{10.5\pi}}{a}, \frac{0.2e^{10.1\pi}, 0.1e^{10.7\pi}}{b} \right),$$

$$\mathcal{B} = \left( \frac{0.2e^{10.9\pi}, 0.3e^{10.4\pi}}{a}, \frac{0.5e^{10.2\pi}, 0.1e^{10.5\pi}}{b} \right).$$

Then $\tau = \{\emptyset, \bar{T}, \mathcal{A}, \mathcal{B}\}$ is an intuitionistic complex fuzzy topology on $X$, $C^0 = \emptyset = \{x, (0, e^{2\pi i}) : x \in X\}$ and $C^- = \bar{T} = \{x, (e^{2\pi i}, 0) : x \in X\}$.

Definition 34. An intuitionistic complex fuzzy topological space $(X, \tau)$ is said to be extremely disconnected, if the closure of each $CISO$-set is $CISO$-set.

Definition 35. Let $(X, \tau)$ be an intuitionistic complex fuzzy topological space. A subset $\mathcal{A}$ of $X$ is said to be $CIF$ semiopen set (by short $CIFSO$-set) (resp., $CIF$ preopen (by short $CIFPO$-set), $CIF$ $\alpha$-open (by short $CIFA\alpha$O-set), $CIF$ $\beta$-open (by short $CIF\beta$O-set), and $CIFb$-open (by short $CIFBo$-set). If $\mathcal{A} \subseteq \mathcal{A}'$ (resp., $\mathcal{A} \subseteq \mathcal{A}^-$, $\mathcal{A} \subseteq \mathcal{A}^{--}$, $\mathcal{A} \subseteq \mathcal{A}^{---}$, and $\subseteq \mathcal{A}^{--} \cup \mathcal{A}^{---}$), the family of all $CISO$-set (resp., $CISO$-set, $CIFA\alpha$O-set, and $CIF\beta$O-set) in $X$ is denoted by $CISO(X)$ (resp., $CIFO(X)$, $CIFA\alpha O(X)$, and $CIF\beta O(X)$).

The implications between these concepts in the following diagram and the converse are not true in general,

\[
\begin{array}{ccc}
\mathcal{CISO} & \supseteq & \mathcal{CIFPO} \\
\uparrow & & \uparrow \\
\mathcal{CIFA\alpha O} & \supseteq & \mathcal{CIFBo} \\
\uparrow & & \uparrow \\
CIF\beta O \\
\end{array}
\]

(54)

5. Conclusion

In this paper we introduced and studied a temporal complex intuitionistic fuzzy sets as generalization of complex Atanassov’s intuitionistic fuzzy sets by taking the time in the moving of the point; a correlation between two temporal complex intuitionistic fuzzy sets is discussed. A similarity between temporal complex intuitionistic fuzzy sets is main points in the paper as a generalization of the similarity introduced by Omar [10] and Sugeno [13]. We calculate the results by the program Maple 7. Finally we give an applications to know if the patient is suffering from the diseases or not and introduce the main building in a topology by using the same the set. In future research, similarity measures
Table 17: Then the similarity measures $C_I(A,B)$ and $\rho_I(A,B)$. 

<table>
<thead>
<tr>
<th></th>
<th>$t_1$</th>
<th>$t_2$</th>
<th>$t_3$</th>
<th>$t_4$</th>
<th>$t_5$</th>
<th>$t_6$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$x_1$</td>
<td>$C_I(A,B) = -0.2723$</td>
<td>$C_I(A,B) = -0.0010$</td>
<td>$C_I(A,B) = -0.0023$</td>
<td>$C_I(A,B) = -0.0006$</td>
<td>$C_I(A,B) = -0.006$</td>
<td>$C_I(A,B) = -0.006$</td>
</tr>
<tr>
<td></td>
<td>$\rho_I(A,B) = -1.278$</td>
<td>$\rho_I(A,B) = -1.900$</td>
<td>$\rho_I(A,B) = -2.800$</td>
<td>$\rho_I(A,B) = -1.000$</td>
<td>$\rho_I(A,B) = 0.0707$</td>
<td>$\rho_I(A,B) = -1.000$</td>
</tr>
<tr>
<td>$x_2$</td>
<td>$C_I(A,B) = 0.0246$</td>
<td>$C_I(A,B) = -0.0272$</td>
<td>$C_I(A,B) = -0.0277$</td>
<td>$C_I(A,B) = -0.0004$</td>
<td>$C_I(A,B) = 0.0113$</td>
<td>$C_I(A,B) = -0.0045$</td>
</tr>
<tr>
<td></td>
<td>$\rho_I(A,B) = 0.3818$</td>
<td>$\rho_I(A,B) = 0.1144$</td>
<td>$\rho_I(A,B) = 0.2548$</td>
<td>$\rho_I(A,B) = 0.0051$</td>
<td>$\rho_I(A,B) = -0.0044$</td>
<td>$\rho_I(A,B) = 0.0007$</td>
</tr>
<tr>
<td>$x_3$</td>
<td>$C_I(A,B) = -0.3044$</td>
<td>$C_I(A,B) = 0.113$</td>
<td>$C_I(A,B) = -0.0128$</td>
<td>$C_I(A,B) = 0.0051$</td>
<td>$C_I(A,B) = 0.011$</td>
<td>$C_I(A,B) = 0.0007$</td>
</tr>
<tr>
<td></td>
<td>$\rho_I(A,B) = -0.3041$</td>
<td>$\rho_I(A,B) = -0.0133$</td>
<td>$\rho_I(A,B) = -0.133$</td>
<td>$\rho_I(A,B) = -0.9997$</td>
<td>$\rho_I(A,B) = -0.7297$</td>
<td>$\rho_I(A,B) = -0.7300$</td>
</tr>
<tr>
<td>$x_4$</td>
<td>$C_I(A,B) = 0.0265$</td>
<td>$C_I(A,B) = 0.0048$</td>
<td>$C_I(A,B) = -0.0044$</td>
<td>$C_I(A,B) = -0.0008$</td>
<td>$C_I(A,B) = -1.000$</td>
<td>$C_I(A,B) = -0.006$</td>
</tr>
<tr>
<td></td>
<td>$\rho_I(A,B) = 0.4110$</td>
<td>$\rho_I(A,B) = -0.6393$</td>
<td>$\rho_I(A,B) = -0.6148$</td>
<td>$\rho_I(A,B) = -0.4198$</td>
<td>$\rho_I(A,B) = -1.000$</td>
<td>$\rho_I(A,B) = -1.000$</td>
</tr>
<tr>
<td>$x_5$</td>
<td>$C_I(A,B) = -0.0265$</td>
<td>$C_I(A,B) = 0.0048$</td>
<td>$C_I(A,B) = -0.0044$</td>
<td>$C_I(A,B) = -0.0008$</td>
<td>$C_I(A,B) = 0.03$</td>
<td>$C_I(A,B) = 0.003$</td>
</tr>
<tr>
<td></td>
<td>$\rho_I(A,B) = -1.300$</td>
<td>$\rho_I(A,B) = -0.5500$</td>
<td>$\rho_I(A,B) = -3.800$</td>
<td>$\rho_I(A,B) = -0.6519$</td>
<td>$\rho_I(A,B) = -5.88$</td>
<td>$\rho_I(A,B) = 0.266$</td>
</tr>
</tbody>
</table>
between temporal complex multifuzzy soft and applications in engineering, medical, physics, and automobiles will be studied.

**Data Availability**

No data were used to support this study.

**Conflicts of Interest**

The author declares no conflicts of interest.

**Authors’ Contributions**

All authors contributed equally.

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**References**


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**Table 18: Standard case symptoms of one of diseases \( L \).**

<table>
<thead>
<tr>
<th>( x_1 )</th>
<th>( t_1 )</th>
<th>( (0.2i, 0.3i) )</th>
<th>( t_2 )</th>
<th>( (-0.1i, -0.3i) )</th>
<th>( t_3 )</th>
<th>( (0.3i, 0.5i) )</th>
<th>( t_4 )</th>
<th>( (0.2i, 0.1i) )</th>
<th>( t_5 )</th>
<th>( (0.2i, 0.1i) )</th>
<th>( t_6 )</th>
<th>( (0.2i, 0.1i) )</th>
</tr>
</thead>
</table>

**Table 19: The symptoms of case \( L_1 \).**

<table>
<thead>
<tr>
<th>( x_1 )</th>
<th>( t_1 )</th>
<th>( (0.1i, 0.1i) )</th>
<th>( t_2 )</th>
<th>( (-0.1i, -0.6i) )</th>
<th>( t_3 )</th>
<th>( (0.1i, 0.5i) )</th>
<th>( t_4 )</th>
<th>( (0.2i, 0.1i) )</th>
<th>( t_5 )</th>
<th>( (0.2i, 0.1i) )</th>
<th>( t_6 )</th>
<th>( (0.2i, 0.1i) )</th>
</tr>
</thead>
</table>

**Table 20: Similarity measures \( C_T(L, L_1) \).**

<table>
<thead>
<tr>
<th>( x_1 )</th>
<th>( C_T(L, L_1) = -0.02723 )</th>
<th>( C_T(L, L_1) = 0.001 )</th>
<th>( C_T(L, L_1) = 0.0023 )</th>
<th>( C_T(L, L_1) = -0.0006 )</th>
<th>( C_T(L, L_1) = -0.006 )</th>
<th>( C_T(L, L_1) = -0.006 )</th>
</tr>
</thead>
</table>

<table>
<thead>
<tr>
<th>( x_2 )</th>
<th>( C_T(L, L_1) = 0.0246 )</th>
<th>( C_T(L, L_1) = -0.0045 )</th>
<th>( C_T(L, L_1) = -0.1144 )</th>
<th>( C_T(L, L_1) = 0.00 )</th>
<th>( C_T(L, L_1) = 0.003 )</th>
<th>( C_T(L, L_1) = 0.004 )</th>
</tr>
</thead>
</table>

<table>
<thead>
<tr>
<th>( x_3 )</th>
<th>( C_T(L, L_1) = 0.0277 )</th>
<th>( C_T(L, L_1) = -0.0007 )</th>
<th>( C_T(L, L_1) = -0.0004 )</th>
<th>( C_T(L, L_1) = 0.00 )</th>
<th>( C_T(L, L_1) = -0.014 )</th>
<th>( C_T(L, L_1) = -0.001 )</th>
</tr>
</thead>
</table>

<table>
<thead>
<tr>
<th>( x_4 )</th>
<th>( C_T(L, L_1) = 0.0265 )</th>
<th>( C_T(L, L_1) = 0.0048 )</th>
<th>( C_T(L, L_1) = -0.0044 )</th>
<th>( C_T(L, L_1) = -0.0008 )</th>
<th>( C_T(L, L_1) = -0.006 )</th>
<th>( C_T(L, L_1) = -0.002 )</th>
</tr>
</thead>
</table>

| \( x_5 \) | \( C_T(L, L_1) = -0.0265 \) | \( C_T(L, L_1) = 0.0048 \) | \( C_T(L, L_1) = -0.0044 \) | \( C_T(L, L_1) = -0.0008 \) | \( C_T(L, L_1) = 0.003 \) | \( C_T(L, L_1) = -0.003 \) |


