NC-Cross Entropy Based MADM Strategy in Neutrosophic Cubic Set Environment

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Received: 29 March 2018; Accepted: 26 April 2018; Published: 29 April 2018

Abstract: The objective of the paper is to introduce a new cross entropy measure in a neutrosophic cubic set (NCS) environment, which we call NC-cross entropy measure. We prove its basic properties. We also propose weighted NC-cross entropy and investigate its basic properties. We develop a novel multi attribute decision-making (MADM) strategy based on a weighted NC-cross entropy measure. To show the feasibility and applicability of the proposed multi attribute decision-making strategy, we solve an illustrative example of the multi attribute decision-making problem.

Keywords: single valued neutrosophic set (SVNS); interval neutrosophic set (INS); neutrosophic cubic set (NCS); multi attribute decision-making (MADM); NC-cross entropy measure

1. Introduction

In 1998, Smarandache [1] introduced a neutrosophic set by considering membership (truth), indeterminacy, non-membership (falsity) functions as independent components to uncertain, inconsistent and incomplete information. In 2010, Wang et al. [2] defined a single valued neutrosophic set (SVNS), a subclass of neutrosophic sets to deal with real and scientific and engineering applications. In the medical domain, Ansari et al. [3] employed a neutrosophic set and neutrosophic inference to knowledge based systems. Several researchers applied neutrosophic sets effectively for image segmentation problems [4–9]. Neutrosophic sets are also applied for integrating geographic information system data [10] and for binary classification problems [11].

Pramanik and Chackrabarti [12] studied the problems faced by construction workers in West Bengal in order to find its solutions using neutrosophic cognitive maps [13]. Based on the experts’ opinion and the notion of indeterminacy, the authors formulated a neutrosophic cognitive map and studied the effect of two instantaneous state vectors separately on a connection matrix and neutrosophic adjacency matrix. Mondal and Pramanik [14] identified some of the problems of Hijras (third gender), namely, absence of social security, education problems, bad habits, health problems, stigma and discrimination, access to information and service problems, violence, issues of the Hijra community, and sexual behavior problems. Based on the experts’ opinion and the notion of indeterminacy, the authors formulated a neutrosophic cognitive map and presented the effect of two instantaneous state vectors separately on a connection matrix and neutrosophic adjacency matrix.

Pramanik and Roy [15] studied the game theoretic model [16] of the Jammu and Kashmir conflict between India and Pakistan in a SVNS environment. The authors examined the progress and the
status of the conflict, as well as the dynamics of the relationship by focusing on the influence of United States of America, India and China in crisis dynamics. The authors investigated the possible solutions. The authors also explored the possibilities and developed arguments for an application of the principle of neutrosophic game theory to present a standard $2 \times 2$ zero-sum game theoretic model to identify an optimal solution.

Maria Sodenkamp applied the concept of SVNSs in multi attribute decision-making (MADM) in her Ph. D. thesis in 2013 [17]. In 2018, Pranab Biswas [18] studied various strategies for MADM in SVNS environment in his Ph. D. thesis. Kharal [19] presented a MADM strategy in a single valued neutrosophic environment and presented the application of the proposed strategy for the evaluation of university professors for tenure and promotions. Mondal and Pramanik [20] extended the teacher selection strategy [21] in SVNS environments. Mondal and Pramanik [22] also presented MADM strategy to school choice problems in SVNS environments. Mondal and Pramanik [23] and presented an MADM decision-making model for clay-brick selection in a construction field based on grey relational analysis in SVNS environments. Biswas et al. [24–26] presented several MADM strategies in single valued neutrosophic environments such as technique for order of preference by similarity to ideal solution (TOPSIS) [24], grey relational analysis [25], and entropy based MADM [26]. Several studies [27–30]—using similarity measures based MADM—have been proposed in SVNS environments. Several studies enrich the study of MADM in SVNS environments such as projection and bidirectional projection measure based MADM [31], maximizing deviation method [32], Frank prioritized Bonferroni mean operator based MADM [33], biparametric distance measures based MADM [34], prospect theory based MADM [35], multi-objective optimization by ratio analysis plus the full multiplicative form (MULTIMOORA) [36], weighted aggregated sum product assessment (WASPAS) [37,38], complex proportional assessment (COPRAS) [39], TODIM [40], projection based TODIM [41], outranking [42], analytic hierarchy process (AHP) [43], and Vlsekriterijumska optimizacija i Kompromisno Resenje (VIKOR) [44].

In 2005, Wang et al. [45] introduced the interval neutrosophic set (INS) by considering membership function, non-membership function and indeterminacy function as independent functions that assume the values in interval form. Pramanik and Mondal [46] extended the single valued neutrosophic grey relational analysis method to interval neutrosophic environments and applied it to an MADM problem. The authors employed an information entropy method, which is used to obtain the unknown attribute weights and presented a numerical example. Dey et al. [47] investigated an extended grey relational analysis strategy for MADM problems in uncertain interval neutrosophic linguistic environments. The authors solved a numerical example and compared the obtained results with results obtained from the other existing strategies in the literature. Dey et al. [48] developed two MADM strategies in INS environment based on the combination of angle cosine and projection method. The authors presented an illustrative numerical example in the Khadi institution to demonstrate the effectiveness of the proposed MADM strategies. Several studies enrich the development of MADM in INS environments such as VIKOR [49], TOPSIS [50, 51], outranking strategy [52], similarity measure [29, 53–55], weighted correlation coefficient based MADM strategy [56], and generalized weighted aggregation operator based MADM strategy [57]. The study in recent trends in neutrosophic theory and applications can be found in [58].

Ali et al. [59] proposed the neutrosophic cubic set (NCS) by hybridizing NS and INS. Banerjee et al. [60] developed the grey relational analysis based MADM strategy in NCS environments. Pramanik et al. [61] presented an Extended TOPSIS strategy for MADM in NCS environments with neutrosophic cubic information. Zhan et al. [62] developed an MADM strategy based on two weighted average operators in NCS environments. Lu and Ye [63] presented three cosine measures between NCSs and established three MADM strategies in NCS environments. Shi and Ye [64] introduced Dombi aggregation operators of NCSs and applied for an MADM problem. Ye [65] presents operations and an aggregation method of neutrosophic cubic numbers for MADM. For multi attribute group decision-making (MAGDM), Pramanik et al. [66] defined a similarity measure for NCSs and proved some of its basic properties and developed a new MAGDM strategy with linguistic variables in NCS environments. To develop TODIM, Pramanik et al. [67]
proposed the score and accuracy functions for NCSs and proved their basic properties, and developed a strategy for ranking of neutrosophic cubic numbers based on the score and accuracy functions. In the same study, Pramanik et al. [67] presented a numerical example and presented a comparison analysis. In the same study, Pramanik et al. [67] conducted a sensitivity analysis to reflect the impact of ranking order of the alternatives for different values of attenuation factor of losses for MAGDM strategies.

Pramanik et al. [68] proposed a new VIKOR strategy for MAGDM in NCS environments. The authors also presented sensitivity analysis to reflect the impact of different values of the decision-making mechanism coefficient on ranking order of the alternatives.

Cross entropy measure is an important measure to calculate the divergence of any variable from prior one variable. In 1968, Zadeh [69] first proposed fuzzy entropy to deal with the divergence of two fuzzy variables. Thereafter, Deluca and Termini [70] introduced some axioms of fuzzy entropy involving Shannon’s function [71]. Szmidt and Kacprzyk [72] proposed an entropy measure for intuitionistic fuzzy sets (IFs) by employing a geometric interpretation of IFs. Majumder and Samanta [73] introduced a SVNS based entropy measure and similarity measure. Furthermore, Aydogdu [74] proposed an entropy measure for INSs. Ye and Du [75] proposed some entropy measures for INS based on the distances as the extension of the entropy measures of interval valued IFs. Shang and Jiang [76] defined cross entropy in a fuzzy environment between two fuzzy variables. Vlachos and Sergiadis [77] defined intuitionistic fuzzy cross-entropy. Ye [78] employed intuitionistic fuzzy cross entropy to multi criteria decision-making (MCDM). Maheshwari and Srivastava [79] proposed new cross entropy in intuitionistic fuzzy environment and employed it to medical diagnosis. Zhang et al. [80] defined entropy and cross entropy measure for interval-intuitionistic sets and discuss their properties. Ye [81] proposed an interval intuitionistic fuzzy cross entropy measure and employed it to solve MCDM problems.

Ye [82] defined cross entropy for SVNSs and employed it solve to MCDM problems. To remove the drawbacks of cross entropy [82], Ye [83] proposed another cross entropy for SVNSs. In the same study, Ye [83] also proposed new cross entropy for INSs. Tian et al. [84] proposed a cross entropy for INS environments and employed it to MCDM problems. Sahin [85] proposed an interval neutrosophic cross entropy measure based on fuzzy cross entropy and single valued neutrosophic cross entropy measures and applied it to MCDM problems. Recently, Pramanik et al. [86] proposed a novel cross entropy, namely, NS-cross entropy in SVNS environments and proved its basic properties. In the same research, Pramanik et al. [86] also proposed weighted NS-cross entropy and employed it to MAGDM problem. Furthermore, Dalapati et al. [87] extended NS-cross entropy in INS environments and employed it for solving MADM problems. Pramanik et al. [88] developed two new MADM strategies based on cross entropy measures in bipolar neutrosophic sets (BNSs) and interval BNS environments.

1.1. Research Gap: NC-Cross Entropy-Based MADM Strategy in NCS Environments

This study answers the following research questions:

1. Is it possible to introduce an NC-cross entropy measure in NCS environments?
2. Is it possible to introduce a weighted cross entropy measure in NCS environments?
3. Is it possible to develop a novel MADM strategy based on weighted NC-cross entropy?

1.2. Motivation

The studies [59–68] reveal that cross entropy measure is not proposed in NCS environments. Since MADM strategy is not studied in the literature, we move to propose a comprehensive NC-cross entropy-based strategy for tackling MADM in the NCS environment. This study develops a novel NC-cross entropy-based MADM strategy.

The objectives of the paper are:

1. To introduce a NC-cross entropy measure and establish its basic properties in an NCS environment.
2. To introduce a weighted NC-cross measure and establish its basic properties in NCS environments.
3. To develop a novel MADM strategy based on weighted NC-cross entropy measure in NCS environments.

To fill the research gap, we propose NC-cross entropy-based MADM.

The remainder of the paper is presented as follows: in Section 2, we describe the basic definitions and operation of SVNSs, INSs, and NCSs. In Section 3, we propose an NC-cross entropy measure and a weighted NC-cross entropy measure and establish their basic properties. Section 4 is devoted to developing MADM strategy using NC-cross entropy. Section 5 provides an illustrative numerical example to show the applicability and validity of the proposed strategy in NCS environments. Section 6 presents briefly the contribution of the paper. Section 7 offers conclusions and the future scope of research.

2. Preliminaries

In this section, some basic concepts and definitions of SVNS, INS and NCS are presented that will be utilized to develop the paper.

**Definition 1.** Single valued neutrosophic set (SVNS)
Assume that $U$ is a space of points (objects) with generic elements $u \in U$. A SVNS [2] $H$ in $U$ is characterized by a truth-membership function $T_H(u)$, an indeterminacy-membership function $I_H(u)$, and a falsity-membership function $F_H(u)$, where $T_H(u)$, $I_H(u)$, $F_H(u) \in [0,1]$ for each point $u$ in $U$. Therefore, a SVNS $A$ is expressed as

$$H = \{u, T_H(u), I_H(u), F_H(u) | u \in U\}, \quad (1)$$

whereas the sum of $T_H(u)$, $I_H(u)$ and $F_H(u)$ satisfies the condition:

$$0 \leq T_H(u) + I_H(u) + F_H(u) \leq 3. \quad (2)$$

The order triplet $<T, I, F>$ is called a single valued neutrosophic number (SVNN).

**Example 1.** Let $H$ be any SVNS in $U$; then, $H$ can be expressed as: $H = \{u, (0.7, 0, 3, 0.5) | u \in U\}$ and SVNN presented $H = <0.7, 0.3, 0.5>$.

**Definition 2.** Inclusion of SVNS
The inclusion of any two SVNSs [2] $H_1$ and $H_2$ in $U$ is denoted by $H_1 \subseteq H_2$ and defined as follows:

$$H_1 \subseteq H_2 \text{ iff } T_{H_1}(u) \leq T_{H_2}(u), I_{H_1}(u) \geq I_{H_2}(u), F_{H_1}(u) \geq F_{H_2}(u) \text{ for all } u \in U. \quad (3)$$

**Example 2.** Let $H_1$ and $H_2$ be any two SVNNs in $U$ presented as follows: $H_1 = <(0, 0.7, 0.3, 0.5)>$ and $H_2 = <(0.8, 0.2, 0.4)>$ for all $u \in U$. Using the property of inclusion of two SVNNs, we conclude that $H_1 \subseteq H_2$.

**Definition 3.** Equality of two SVNS
The equality of any two SVNSs [2] $H_1$ and $H_2$ in $U$ denoted by $H_1 = H_2$ is defined as:

$$T_{H_1}(u) = T_{H_2}(u), I_{H_1}(u) = I_{H_2}(u), F_{H_1}(u) = F_{H_2}(u) \text{ for all } u \in U. \quad (4)$$

**Definition 4.** Complement of any SVNS
The complement of any SVNS [2] $H$ in $U$ denoted by $H^c$ and defined as follows:
\[ H^c = \{ u, 1 - T_H(u), 1 - I_H(u), 1 - F_H(u) \mid u \in U \}. \] (5)

**Example 3.** Let \( H \) be any SVNN in \( U \) presented as follows:
\[ H = <0.7, 0.3, 0.5>. \] Then, the compliment of \( H \) is obtained as \( H^c = <0.3, 0.7, 0.5> \).

**Definition 5.** Union of two SVNSs

The union of two SVNSs \([2]\) \( H_1 \) and \( H_2 \) is a neutrosophic set \( H_3 \) (say) written as \( H_3 = H_1 \cup H_2 \).

Here,
\[
\begin{align*}
T_{H_3}(u) &= \max \{T_{H_1}(u), T_{H_2}(u)\}, \\
I_{H_3}(u) &= \min \{I_{H_1}(u), I_{H_2}(u)\}, \\
F_{H_3}(u) &= \min \{F_{H_1}(u), F_{H_2}(u)\}, \forall u \in U.
\end{align*}
\] (6)

**Example 4.** Let \( H_1 \) and \( H_2 \) be two SVNNs in \( U \) presented as follows:
\[ H_1 = <4.0, 3.0, 6.0> \text{ and } H_2 = <6.0, 3.0, 7.0>. \] Then, union of them is obtained using Equation (6) as follows:
\[ H_1 \cup H_2 = <6.0, 3.0, 6.0>. \]

**Definition 6.** Intersection of any two SVNSs

The intersection of two SVNSs \([2]\) \( H_1 \) and \( H_2 \) denoted by \( H_4 \) and defined as \( H_4 = H_1 \cap H_2 \).

Here,
\[
\begin{align*}
T_{H_4}(u) &= \min \{T_{H_1}(u), T_{H_2}(u)\}, \\
I_{H_4}(u) &= \max \{I_{H_1}(u), I_{H_2}(u)\}, \\
F_{H_4}(u) &= \max \{F_{H_1}(u), F_{H_2}(u)\}, \forall u \in U.
\end{align*}
\] (7)

**Example 5.** Let \( H_1 \) and \( H_2 \) be two SVNNs in \( U \) presented as follows:
\[ H_1 = <6.0, 3.0, 7.0> \text{ and } H_2 = <4.0, 3.0, 6.0>. \] Then, the intersection of them is obtained using Equation (7) as follows:
\[ H_1 \cap H_2 = <0.6, 0.3, 0.6>. \]

**Definition 7.** Some operations of SVNS

Let \( H_1 \) and \( H_2 \) be any two SVNSs \([2]\). Then, addition and multiplication are defined as:
\[
\begin{align*}
H_1 \oplus H_2 &= <T_{H_1}(u) + T_{H_2}(u), T_{H_1}(u) - T_{H_2}(u), I_{H_1}(u), I_{H_2}(u), F_{H_1}(u), F_{H_2}(u)>, \forall u \in U. \\
H_1 \otimes H_2 &= <T_{H_1}(u) - T_{H_2}(u), I_{H_1}(u), I_{H_2}(u) + I_{H_2}(u) - I_{H_1}(u), F_{H_1}(u) + F_{H_2}(u) - F_{H_1}(u), F_{H_2}(u)>, \forall u \in U.
\end{align*}
\] (8)

\(9\)

**Example 6.** Let \( H_1 \) and \( H_2 \) be two SVNNs in \( U \) presented as follows:
\[ H_1 = <0.6, 0.3, 0.4> \text{ and } H_2 = <0.7, 0.3, 0.6>. \] Then, using Equations (8) and (9), we obtained \( H_1 \oplus H_2 \) and \( H_1 \otimes H_2 \) as follows:
1. \( H_1 \oplus H_2 = <0.88, 0.09, 0.24> \).
2. \( H_1 \otimes H_2 = <0.42, 0.51, 0.76> \).

**Definition 8.** Interval neutrosophic set (INS)

Assume that \( U \) is a space of points (objects) with generic element \( u \in U \). An INS \([45]\) \( J \) in \( U \) is characterized by a truth-membership function \( T_J(u) \), an indeterminacy-membership function \( I_J(u) \), and a falsity-membership function \( F_J(u) \), where \( T_J(u) = [T^+_J(u), T^-_J(u)] \), \( I_J(u) = [I^+_J(u), I^-_J(u)] \), \( F_J(u) = [F^+_J(u), F^-_J(u)] \) for each point \( u \) in \( U \). Therefore, a INSs \( J \) can be expressed as
\[ J = \{ u, [T^+_J(u), T^-_J(u)], [I^+_J(u), I^-_J(u)], [F^+_J(u), F^-_J(u)] \mid u \in U \}, \] (10)
where \( T_j^u, T_j^+_u, I_j^u, I_j^+_u, F_j^u, F_j^+_u \subseteq [0, 1]. \)

**Definition 9.** Inclusion of two INSs

Let \( J_1 = \{ u, [T_{j_1}^u(u), T_{j_1}^+_u(u)], [I_{j_1}^u(u), I_{j_1}^+_u(u)], [F_{j_1}^u(u), F_{j_1}^+_u(u)] \} \subseteq U \) and
\[
J_2 = \{ u, [T_{j_2}^u(u), T_{j_2}^+_u(u)], [I_{j_2}^u(u), I_{j_2}^+_u(u)], [F_{j_2}^u(u), F_{j_2}^+_u(u)] \} \subseteq U \)
be any two INSs \([45]\)

in \( U \), then \( J_1 \subseteq J_2 \) iff
\[
T_{j_1}^u(u) \leq T_{j_2}^u(u), \quad I_{j_1}^u(u) \leq I_{j_2}^u(u), \quad F_{j_1}^u(u) \leq F_{j_2}^u(u) \]
for all \( u \in U \).

**Definition 10.** Complement of an INS

The complement \( J^c \) of an INS \([45]\) \( J = \{ u, [T_j^u(u), T_j^+_u(u)], [I_j^u(u), I_j^+_u(u)], [F_j^u(u), F_j^+_u(u)] \} \subseteq U \) is defined as follows:
\[
J^c = \{ u, [1-T_j^u(u), 1-T_j^+_u(u)], [1-I_j^u(u), 1-I_j^+_u(u)], [1-F_j^u(u), 1-F_j^+_u(u)] \} \subseteq U \}
\]

**Definition 11.** Equality of two INSs

Let \( J_1 = \{ u, [T_{j_1}^u(u), T_{j_1}^+_u(u)], [I_{j_1}^u(u), I_{j_1}^+_u(u)], [F_{j_1}^u(u), F_{j_1}^+_u(u)] \} \subseteq U \) and
\[
J_2 = \{ u, [T_{j_2}^u(u), T_{j_2}^+_u(u)], [I_{j_2}^u(u), I_{j_2}^+_u(u)], [F_{j_2}^u(u), F_{j_2}^+_u(u)] \} \subseteq U \)
be any two INSs \([45]\)

in \( U \), then \( J_1 = J_2 \) iff
\[
T_{j_1}^u(u) = T_{j_2}^u(u), \quad I_{j_1}^u(u) = I_{j_2}^u(u), \quad F_{j_1}^u(u) = F_{j_2}^u(u) \]
for all \( u \in U \).

**Definition 12.** Neutrosophic cubic set (NCS)

Assume that \( U \) is a space of points (objects) with generic elements \( u_i \in U \). A NCS \([59]\) \( Q \) in \( U \) is a hybrid structure of INS and SVNS that can be expressed as follows:
\[
Q = \{ u, [T_{q_0}(u), T_{q_0}^+(u), I_{q_0}(u), I_{q_0}^+(u)], [F_{q_0}(u), F_{q_0}^+(u)], (T_{q_0}(u), I_{q_0}(u), F_{q_0}(u)) \} \}
\]

Here, \( (T_{q_0}(u), I_{q_0}(u), F_{q_0}(u)) \) and \( (T_{q_0}(u), I_{q_0}(u), F_{q_0}(u)) \) are INSs and SVNSs, respectively, in \( U \). NCS can be simply presented as
\[
<T_{q_0}(u), I_{q_0}(u), F_{q_0}(u), T_{q_0}(u), I_{q_0}(u), F_{q_0}(u)>.
\]

We call Equation (15) a neutrosophic cubic number (NCN).

**Definition 13.** Inclusion of two NCSs

Let \( Q_1 = \{ u, [T_{q_1}(u), T_{q_1}^+(u), I_{q_1}(u), I_{q_1}^+(u)], [F_{q_1}(u), F_{q_1}^+(u)], (T_{q_1}(u), I_{q_1}(u), F_{q_1}(u)) \} \subseteq U \) and
\[
Q_2 = \{ u, [T_{q_2}(u), T_{q_2}^+(u), I_{q_2}(u), I_{q_2}^+(u)], [F_{q_2}(u), F_{q_2}^+(u)], (T_{q_2}(u), I_{q_2}(u), F_{q_2}(u)) \} \subseteq U \}
be any two NCSs \([59]\) in \( U \). Then, \( Q_1 \subseteq Q_2 \) iff
\[
T_{q_1}(u) \leq T_{q_2}(u), \quad I_{q_1}(u) \leq I_{q_2}(u), \quad F_{q_1}(u) \leq F_{q_2}(u)
\]
\[ I_0(u) \geq I_0(u), \quad I_0(u) \geq I_0(u), \quad F_0(u) \geq F_0(u), \quad F_0(u) \geq F_0(u) \quad \text{and} \]
\[ T_0(u) \leq T_0(u), \quad I_0(u) \leq I_0(u), \quad F_0(u) \leq F_0(u) \quad \text{for all } u \in U. \]

**Definition 14. Equality of two NCSs**

Let \( Q_1 = \{ u, [T_0(u), T_0(u)], [I_0(u), I_0(u)], [F_0(u), F_0(u)], (T_0(u), I_0(u), F_0(u)) > |u_i| \in U \} \)

and

\[ Q_2 = \{ u, [T_0^+(u), T_0^+(u)], [I_0^+(u), I_0^+(u)], [F_0^+(u), F_0^+(u)], (T_0^+(u), I_0^+(u), F_0^+(u)) > |u_i| \in U \}

be any two NCSs [59] in \( U \). Then \( Q_1 = Q_2 \) iff

\[ I_0(u) = I_0(u), \quad I_0(u) = I_0(u), \quad F_0(u) = F_0(u), \quad F_0(u) = F_0(u) \quad \text{and} \]
\[ T_0(u) = T_0(u), I_0(u) = I_0(u), F_0(u) = F_0(u) \quad \text{for all } u \in U. \]  

**Definition 15. Complement of an NCS**

Let \( Q = \{ u, [T_0(u), T_0(u)], [I_0(u), I_0(u)], [F_0(u), F_0(u)], (T_0(u), I_0(u), F_0(u)) > |u_i| \in U \} \) be any NCS [59] in \( U \). Then, complement \( Q^c \) of \( Q \) is defined as follows:

\[ Q^c = \{ u, [1-T_0(u), 1-T_0(u)], [1-I_0(u), 1-I_0(u)], [1-F_0(u), 1-F_0(u)], (1-T_0(u), 1-I_0(u), 1-F_0(u)) > |u_i| \in U \}. \]

3. NC-Cross-entropy measure in NCS environment

**Definition 16. NC-cross entropy measure**

Let \( Q_1 \) and \( Q_2 \) be any two NCSs in \( U = \{ u_1, u_2, ..., u_n \} \). Then, neutrosophic cubic cross-entropy measure of \( Q_1 \) and \( Q_2 \) is denoted by \( CE_{NC}(Q_1, Q_2) \) and defined as follows:
Theorem 1. Let \( Q_1, Q_2 \) be any two NCSs in \( U \). The NC-cross entropy measure \( \text{CE}_{\text{NC}}(Q_1, Q_2) \) satisfies the following properties:

(i) \( \text{CE}_{\text{NC}}(Q_1, Q_2) \geq 0 \).

(ii) \( \text{CE}_{\text{NC}}(Q_1, Q_2) = 0 \) iff \( T_{Q_1}(u_i) = T_{Q_2}(u_i) \), \( T_{Q_1}(u_i) = T_{Q_2}(u_i) \), \( I_{Q_1}(u_i) = I_{Q_2}(u_i) \), \( I_{Q_1}(u_i) = I_{Q_2}(u_i) \), \( F_{Q_1}(u_i) = F_{Q_2}(u_i) \), \( F_{Q_1}(u_i) = F_{Q_2}(u_i) \), and \( Q_1 = Q_2 \) for all \( u_i \in U \).

(iii) \( \text{CE}_{\text{NC}}(Q_1, Q_2) = \text{CE}_{\text{NC}}(Q_2', Q_2') \).

(iv) \( \text{CE}_{\text{NC}}(Q_1, Q_2) = \text{CE}_{\text{NC}}(Q_2, Q_1) \).

Proof of Theorem 1.

(i)
For all values of $u_i \in U$, $|\Gamma_{01}(u)| \geq 0$, $|\Gamma_{02}(u)| \geq 0$, $|\Gamma_{01}(u) - \Gamma_{02}(u)| \geq 0$, $\sqrt{1 + |\Gamma_{01}(u)|^2} \geq 0$, $\sqrt{1 + |\Gamma_{02}(u)|^2} \geq 0$, $|1 - \Gamma_{01}(u)| \geq 0$, $|1 - \Gamma_{02}(u)| \geq 0$, $|1 - \Gamma_{01}(u)| - (1 - \Gamma_{02}(u)) \geq 0$, $\sqrt{1 + |1 - \Gamma_{01}(u)|^2} \geq 0$, $\sqrt{1 + |1 - \Gamma_{02}(u)|^2} \geq 0$. Then,

$$\begin{bmatrix}
2|\Gamma_{01}(u)| - |\Gamma_{02}(u)| \\
\sqrt{1 + |\Gamma_{01}(u)|^2} + \sqrt{1 + |\Gamma_{02}(u)|^2}
\end{bmatrix} \geq 0 \quad (20)
$$

Similarly,

$$\begin{bmatrix}
2|\Gamma_{01}(u)| - |\Gamma_{02}(u)| \\
\sqrt{1 + |\Gamma_{01}(u)|^2} + \sqrt{1 + |\Gamma_{02}(u)|^2}
\end{bmatrix} \geq 0 \quad (21)
$$

and

$$\begin{bmatrix}
2|\Gamma_{01}(u)| - |\Gamma_{02}(u)| \\
\sqrt{1 + |\Gamma_{01}(u)|^2} + \sqrt{1 + |\Gamma_{02}(u)|^2}
\end{bmatrix} \geq 0 \quad (22)
$$

Again,

For all values of $u_i \in U$, $|\Gamma_{01}(u)| \geq 0$, $|\Gamma_{02}(u)| \geq 0$, $|\Gamma_{01}(u) - \Gamma_{02}(u)| \geq 0$, $\sqrt{1 + |\Gamma_{01}(u)|^2} \geq 0$, $\sqrt{1 + |\Gamma_{02}(u)|^2} \geq 0$, $|1 - \Gamma_{01}(u)| \geq 0$, $|1 - \Gamma_{02}(u)| \geq 0$, $|1 - \Gamma_{01}(u)| - (1 - \Gamma_{02}(u)) \geq 0$, $\sqrt{1 + |1 - \Gamma_{01}(u)|^2} \geq 0$, $\sqrt{1 + |1 - \Gamma_{02}(u)|^2} \geq 0$. Then,

$$\begin{bmatrix}
2|\Gamma_{01}(u)| - |\Gamma_{02}(u)| \\
\sqrt{1 + |\Gamma_{01}(u)|^2} + \sqrt{1 + |\Gamma_{02}(u)|^2}
\end{bmatrix} \geq 0 \quad (23)
$$

and

$$\begin{bmatrix}
2|\Gamma_{01}(u)| - |\Gamma_{02}(u)| \\
\sqrt{1 + |\Gamma_{01}(u)|^2} + \sqrt{1 + |\Gamma_{02}(u)|^2}
\end{bmatrix} \geq 0 \quad (24)
$$

Similarly, we can show that

$$\begin{bmatrix}
2|\Gamma_{01}(u)| - |\Gamma_{02}(u)| \\
\sqrt{1 + |\Gamma_{01}(u)|^2} + \sqrt{1 + |\Gamma_{02}(u)|^2}
\end{bmatrix} \geq 0 \quad (25)$$
\[
\begin{align*}
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] \geq 0 \quad (26) \\
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] \geq 0 \quad (27)
\end{align*}
\]

and
\[
\begin{align*}
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] \geq 0 \quad (28)
\end{align*}
\]

Adding Equation (20) to Equation (28), we obtain \( CE_{nc} (Q_1, Q_2) > 0 \).

(ii).
\[
\begin{align*}
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] = 0, \quad (29) \\
&\Rightarrow T_{0_1}(u) = T_{0_2}(u)
\end{align*}
\]

\[
\begin{align*}
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] = 0, \quad (30) \\
&\Rightarrow I_0(u) = I_0(u)
\end{align*}
\]

\[
\begin{align*}
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] = 0, \quad (31) \\
&\Rightarrow F_0(u) = F_0(u), \text{ For all values of } u_i \in U.
\end{align*}
\]

Again,
\[
\begin{align*}
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] = 0, \quad (32) \\
&\Rightarrow T_{0_1}(u) = T_{0_2}(u),
\end{align*}
\]

\[
\begin{align*}
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] = 0, \quad (33) \\
&\Rightarrow T_{0_1}(u) = T_{0_2}(u),
\end{align*}
\]

\[
\begin{align*}
&\left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] + \left[ \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} + \frac{2}{\sqrt{1+I_0(u)^2} + \sqrt{1+I_0(u)^2}} \right] = 0, \quad (34) \\
&\Rightarrow I_0(u) = I_0(u),
\end{align*}
\]
$$\begin{align*}
\mathcal{C}_{\text{nc}}(Q_1, Q_2) &= \frac{1}{8} \sum_{i=1}^{m} \left[ \frac{2|I_{Q_1}^-(u) - I_{Q_2}^-(u)|}{\sqrt{1 + |I_{Q_1}^- (u)|^2} + \sqrt{1 + |I_{Q_2}^- (u)|^2}} + \frac{2|1 - I_{Q_1}^-(u) - (1 - I_{Q_2}^-(u))|}{\sqrt{1 + (1 - I_{Q_1}^- (u))^2} + \sqrt{1 + (1 - I_{Q_2}^- (u))^2}} \right] \\
&+ \frac{2|F_{Q_1}(u) - F_{Q_2}(u)|}{\sqrt{1 + |F_{Q_1}^+ (u)|^2} + \sqrt{1 + |F_{Q_2}^+ (u)|^2}} + \frac{2|1 - F_{Q_1}(u) - (1 - F_{Q_2}(u))|}{\sqrt{1 + (1 - F_{Q_1}^- (u))^2} + \sqrt{1 + (1 - F_{Q_2}^- (u))^2}} \\
&+ \frac{2|I_{Q_1}^-(u) - I_{Q_2}^-(u)|}{\sqrt{1 + |I_{Q_1}^- (u)|^2} + \sqrt{1 + |I_{Q_2}^- (u)|^2}} + \frac{2|1 - I_{Q_1}^-(u) - (1 - I_{Q_2}^-(u))|}{\sqrt{1 + (1 - I_{Q_1}^- (u))^2} + \sqrt{1 + (1 - I_{Q_2}^- (u))^2}} \\
&+ \frac{2|F_{Q_1}(u) - F_{Q_2}(u)|}{\sqrt{1 + |F_{Q_1}^+ (u)|^2} + \sqrt{1 + |F_{Q_2}^+ (u)|^2}} + \frac{2|1 - F_{Q_1}(u) - (1 - F_{Q_2}(u))|}{\sqrt{1 + (1 - F_{Q_1}^- (u))^2} + \sqrt{1 + (1 - F_{Q_2}^- (u))^2}} \\
&+ \frac{2|I_{Q_1}^-(u) - I_{Q_2}^-(u)|}{\sqrt{1 + |I_{Q_1}^- (u)|^2} + \sqrt{1 + |I_{Q_2}^- (u)|^2}} + \frac{2|1 - I_{Q_1}^-(u) - (1 - I_{Q_2}^-(u))|}{\sqrt{1 + (1 - I_{Q_1}^- (u))^2} + \sqrt{1 + (1 - I_{Q_2}^- (u))^2}} \\
&+ \frac{2|F_{Q_1}(u) - F_{Q_2}(u)|}{\sqrt{1 + |F_{Q_1}^+ (u)|^2} + \sqrt{1 + |F_{Q_2}^+ (u)|^2}} + \frac{2|1 - F_{Q_1}(u) - (1 - F_{Q_2}(u))|}{\sqrt{1 + (1 - F_{Q_1}^- (u))^2} + \sqrt{1 + (1 - F_{Q_2}^- (u))^2}} \\
&+ \frac{2|I_{Q_1}^-(u) - I_{Q_2}^-(u)|}{\sqrt{1 + |I_{Q_1}^- (u)|^2} + \sqrt{1 + |I_{Q_2}^- (u)|^2}} + \frac{2|1 - I_{Q_1}^-(u) - (1 - I_{Q_2}^-(u))|}{\sqrt{1 + (1 - I_{Q_1}^- (u))^2} + \sqrt{1 + (1 - I_{Q_2}^- (u))^2}} \\
&+ \frac{2|F_{Q_1}(u) - F_{Q_2}(u)|}{\sqrt{1 + |F_{Q_1}^+ (u)|^2} + \sqrt{1 + |F_{Q_2}^+ (u)|^2}} + \frac{2|1 - F_{Q_1}(u) - (1 - F_{Q_2}(u))|}{\sqrt{1 + (1 - F_{Q_1}^- (u))^2} + \sqrt{1 + (1 - F_{Q_2}^- (u))^2}} .
\end{align*}$$

(35)
\[
\sum_{i=1}^{3} \left( \begin{array}{c}
\frac{2\left(F_{\Omega_{1}}^{+}(u)-F_{\Omega_{2}}^{+}(u)\right)}{\sqrt{1+F_{\Omega_{1}}^{+}(u)} + \sqrt{1+F_{\Omega_{2}}^{+}(u)}} + \frac{2\left(1-F_{\Omega_{1}}^{-}(u))-(1-F_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(1-F_{\Omega_{1}}^{-}(u))\right)} + \frac{2\left(1-F_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(1-F_{\Omega_{2}}^{-}(u))\right)}} \\
\frac{2\left(1-T_{\Omega_{1}}^{+}(u))-(1-T_{\Omega_{2}}^{+}(u))\right)}{\sqrt{1+\left(1-T_{\Omega_{1}}^{+}(u))\right)} + \frac{2\left(1-T_{\Omega_{2}}^{+}(u))\right)}{\sqrt{1+\left(1-T_{\Omega_{2}}^{+}(u))\right)}} \\
\frac{2\left(I_{\Omega_{1}}^{+}(u))-(1-I_{\Omega_{2}}^{+}(u))\right)}{\sqrt{1+\left(I_{\Omega_{1}}^{+}(u))\right)} + \frac{2\left(1-I_{\Omega_{2}}^{+}(u))\right)}{\sqrt{1+\left(1-I_{\Omega_{2}}^{+}(u))\right)}} \\
\frac{2\left(I_{\Omega_{1}}^{-}(u))-(1-I_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(I_{\Omega_{1}}^{-}(u))\right)} + \frac{2\left(1-I_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(1-I_{\Omega_{2}}^{-}(u))\right)}} \\
\frac{2\left(1-F_{\Omega_{1}}(u))-(1-F_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-F_{\Omega_{1}}(u))\right)} + \frac{2\left(1-F_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-F_{\Omega_{2}}(u))\right)}} \\
\frac{2\left(1-T_{\Omega_{1}}(u))-(1-T_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-T_{\Omega_{1}}(u))\right)} + \frac{2\left(1-T_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-T_{\Omega_{2}}(u))\right)}} \\
\frac{2\left(I_{\Omega_{1}}(u))-(1-I_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(I_{\Omega_{1}}(u))\right)} + \frac{2\left(1-I_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-I_{\Omega_{2}}(u))\right)}} \\
\frac{2\left(1-F_{\Omega_{1}}^{-}(u))-(1-F_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(1-F_{\Omega_{1}}^{-}(u))\right)} + \frac{2\left(1-F_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(1-F_{\Omega_{2}}^{-}(u))\right)}} \\
\frac{2\left(1-T_{\Omega_{1}}^{-}(u))-(1-T_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(1-T_{\Omega_{1}}^{-}(u))\right)} + \frac{2\left(1-T_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(1-T_{\Omega_{2}}^{-}(u))\right)}} \\
\frac{2\left(I_{\Omega_{1}}^{-}(u))-(1-I_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(I_{\Omega_{1}}^{-}(u))\right)} + \frac{2\left(1-I_{\Omega_{2}}^{-}(u))\right)}{\sqrt{1+\left(1-I_{\Omega_{2}}^{-}(u))\right)}} \\
\frac{2\left(1-F_{\Omega_{1}}(u))-(1-F_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-F_{\Omega_{1}}(u))\right)} + \frac{2\left(1-F_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-F_{\Omega_{2}}(u))\right)}} \\
\frac{2\left(1-T_{\Omega_{1}}(u))-(1-T_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-T_{\Omega_{1}}(u))\right)} + \frac{2\left(1-T_{\Omega_{2}}(u))\right)}{\sqrt{1+\left(1-T_{\Omega_{2}}(u))\right)}}
\end{array}\right)
\right)
\]
\[\begin{align*}
&= \sum_{i=1}^{n} \left( \frac{2\left[1 - T_{0i}(u_i) - (1 - T_{02}(u_i)) \right]}{\sqrt{1 + \left| T_{0i}(u_i) \right|^2 + \left| T_{02}(u_i) \right|^2}} + \frac{2\left[1 - F_{0i}(u_i) - (1 - F_{02}(u_i)) \right]}{\sqrt{1 + \left| F_{0i}(u_i) \right|^2 + \left| F_{02}(u_i) \right|^2}} \right) \\
& \quad + \sum_{i=1}^{n} \left( \frac{2\left[1 - T_{0i}(u_i) - (1 - T_{02}(u_i)) \right]}{\sqrt{1 + \left| T_{0i}(u_i) \right|^2 + \left| T_{02}(u_i) \right|^2}} + \frac{2\left[1 - T_{0i}(u_i) - (1 - T_{02}(u_i)) \right]}{\sqrt{1 + \left| T_{0i}(u_i) \right|^2 + \left| T_{02}(u_i) \right|^2}} \right) \\
& \quad + \sum_{i=1}^{n} \left( \frac{2\left[1 - I_{0i}(u_i) - (1 - I_{02}(u_i)) \right]}{\sqrt{1 + \left| I_{0i}(u_i) \right|^2 + \left| I_{02}(u_i) \right|^2}} + \frac{2\left[1 - I_{0i}(u_i) - (1 - I_{02}(u_i)) \right]}{\sqrt{1 + \left| I_{0i}(u_i) \right|^2 + \left| I_{02}(u_i) \right|^2}} \right) \\
& \quad + \sum_{i=1}^{n} \left( \frac{2\left[1 - F_{0i}(u_i) - (1 - F_{02}(u_i)) \right]}{\sqrt{1 + \left| F_{0i}(u_i) \right|^2 + \left| F_{02}(u_i) \right|^2}} + \frac{2\left[1 - F_{0i}(u_i) - (1 - F_{02}(u_i)) \right]}{\sqrt{1 + \left| F_{0i}(u_i) \right|^2 + \left| F_{02}(u_i) \right|^2}} \right) \\
& = CE_{w}(Q, Q_{2})
\end{align*}\]

(iv).

Since \( \forall u_i \in U \), for a single valued part, we obtain:
\[\begin{align*}
&= 2\left[1 - T_{0i}(u_i) - (1 - T_{02}(u_i)) \right] + 2\left[1 - I_{0i}(u_i) - (1 - I_{02}(u_i)) \right] + 2\left[1 - F_{0i}(u_i) - (1 - F_{02}(u_i)) \right],
\end{align*}\]
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Then, we obtain
\[
\sqrt{1 + |T_{01}(u)|} + \sqrt{1 + |T_{02}(u)|} = \sqrt{1 + |T_{02}(u)|} + \sqrt{1 + |T_{01}(u)|},
\]
\[
\sqrt{1 + |I_{01}(u)|} + \sqrt{1 + |I_{02}(u)|} = \sqrt{1 + |I_{02}(u)|} + \sqrt{1 + |I_{01}(u)|},
\]
\[
\sqrt{1 + |F_{01}(u)|} + \sqrt{1 + |F_{02}(u)|} = \sqrt{1 + |F_{02}(u)|} + \sqrt{1 + |F_{01}(u)|},
\]
\[
\sqrt{1 + |(1 - T_{01}(u))|} + \sqrt{1 + |(1 - T_{02}(u))|} = \sqrt{1 + |(1 - T_{02}(u))|} + \sqrt{1 + |(1 - T_{01}(u))|},
\]
\[
\sqrt{1 + |(1 - I_{01}(u))|} + \sqrt{1 + |(1 - I_{02}(u))|} = \sqrt{1 + |(1 - I_{02}(u))|} + \sqrt{1 + |(1 - I_{01}(u))|},
\]
\[
\sqrt{1 + |(1 - F_{01}(u))|} + \sqrt{1 + |(1 - F_{02}(u))|} = \sqrt{1 + |(1 - F_{02}(u))|} + \sqrt{1 + |(1 - F_{01}(u))|},
\]
\[
\forall u_i \in U.
\]
For the interval neutrosophic part, we obtain
\[
|T_{01}(u) - T_{02}(u)| = |T_{02}(u) - T_{01}(u)|, 
\]
\[
|I_{01}(u) - I_{02}(u)| = |I_{02}(u) - I_{01}(u)|, 
\]
\[
|F_{01}(u) - F_{02}(u)| = |F_{02}(u) - F_{01}(u)|, 
\]
\[
|(1 - T_{01}(u)) - (1 - T_{02}(u))| = |(1 - T_{02}(u)) - (1 - T_{01}(u))|, 
\]
\[
|(1 - I_{01}(u)) - (1 - I_{02}(u))| = |(1 - I_{02}(u)) - (1 - I_{01}(u))|, 
\]
\[
|(1 - F_{01}(u)) - (1 - F_{02}(u))| = |(1 - F_{02}(u)) - (1 - F_{01}(u))|.
\]

Then, we obtain
\[
\sqrt{1 + |T_{01}(u)|} + \sqrt{1 + |T_{02}(u)|} = \sqrt{1 + |T_{02}(u)|} + \sqrt{1 + |T_{01}(u)|},
\]
\[
\sqrt{1 + |I_{01}(u)|} + \sqrt{1 + |I_{02}(u)|} = \sqrt{1 + |I_{02}(u)|} + \sqrt{1 + |I_{01}(u)|},
\]
\[
\sqrt{1 + |F_{01}(u)|} + \sqrt{1 + |F_{02}(u)|} = \sqrt{1 + |F_{02}(u)|} + \sqrt{1 + |F_{01}(u)|},
\]
\[
\sqrt{1 + |(1 - T_{01}(u))|} + \sqrt{1 + |(1 - T_{02}(u))|} = \sqrt{1 + |(1 - T_{02}(u))|} + \sqrt{1 + |(1 - T_{01}(u))|},
\]
\[
\sqrt{1 + |(1 - I_{01}(u))|} + \sqrt{1 + |(1 - I_{02}(u))|} = \sqrt{1 + |(1 - I_{02}(u))|} + \sqrt{1 + |(1 - I_{01}(u))|},
\]
\[
\sqrt{1 + |(1 - F_{01}(u))|} + \sqrt{1 + |(1 - F_{02}(u))|} = \sqrt{1 + |(1 - F_{02}(u))|} + \sqrt{1 + |(1 - F_{01}(u))|},
\]
\[
\forall u_i \in U.
\]
Similarly,
\[
|T_{01}(u) - T_{02}(u)| = |T_{02}(u) - T_{01}(u)|, 
\]
\[
|I_{01}(u) - I_{02}(u)| = |I_{02}(u) - I_{01}(u)|, 
\]
\[
|F_{01}(u) - F_{02}(u)| = |F_{02}(u) - F_{01}(u)|, 
\]
\[
|(1 - T_{01}(u)) - (1 - T_{02}(u))| = |(1 - T_{02}(u)) - (1 - T_{01}(u))|, 
\]
\[
|(1 - I_{01}(u)) - (1 - I_{02}(u))| = |(1 - I_{02}(u)) - (1 - I_{01}(u))|, 
\]
\[
|(1 - F_{01}(u)) - (1 - F_{02}(u))| = |(1 - F_{02}(u)) - (1 - F_{01}(u))|.
\]

Thus, \(\text{CE}_{\text{NC}}(Q_1, Q_2) = \text{CE}_{\text{NC}}(Q_2, Q_1)\).

**Definition 17.** Weighted NC-cross-entropy measure

We consider the weight \( w_i \) (\( i = 1, 2, 3, \ldots, n \)) of \( u_i \) (\( i = 1, 2, 3, \ldots, n \)) with \( w_i \in [0, 1] \) and \( \sum_{i=1}^{n} w_i = 1 \).

Then, a neutrosophic cubic weighted cross entropy measure between \( Q_1 \) and \( Q_2 \) can be defined as

\[
\|[1 - F_{01}(u)] - (1 - F_{02}(u))\| = \|[0 - F_{02}(u)] - (0 - F_{01}(u))\|.
\]
Theorem 2. Let \( Q_1, Q_2 \) be any two NCSs in \( U \). Then, weighted NC-cross entropy measure \( CE_{NC}^w(Q_1, Q_2) \) satisfies the following properties:

(i) \( CE_{NC}^w(Q_1, Q_2) \geq 0 \).

(ii) \( CE_{NC}^w(Q_1, Q_2) = 0 \) iff \( \frac{T_{01}(u)}{1+T_{01}(u)} = \frac{T_{02}(u)}{1+T_{02}(u)} \), \( \frac{I_{01}(u)}{1+I_{01}(u)} = \frac{I_{02}(u)}{1+I_{02}(u)} \), \( \frac{F_{01}(u)}{1+F_{01}(u)} = \frac{F_{02}(u)}{1+F_{02}(u)} \) and \( T_{01}(u) = T_{02}(u), I_{01}(u) = I_{02}(u), F_{01}(u) = F_{02}(u) \) \( \forall u \in U \).

(iii) \( CE_{NC}^w(Q_1, Q_2) = CE_{NC}^w(Q_2, Q_1) \).

(iv) \( CE_{NC}^w(Q_1, Q_2) \neq CE_{NC}^w(Q_1, Q_3) \).

Proof of Theorem 2.

(i).
For all values of $u_i \in U$, $|T_{Q_1}(u_i)| \geq 0$, $|T_{Q_2}(u_i)| \geq 0$, $|T_{Q_1}(u_i) - T_{Q_2}(u_i)| \geq 0$, $\sqrt{1 + |T_{Q_1}(u_i)|^2} \geq 0$, $\sqrt{1 + |1 - T_{Q_1}(u_i)|^2} \geq 0$, $\sqrt{1 + |1 - T_{Q_2}(u_i)|^2} \geq 0$.

Then,

$$\begin{bmatrix}
\frac{2I_1(u_i) - T_{Q_2}(u_i)}{\sqrt{1 + |I_1(u_i)|^2}} & \frac{2|1 - I_1(u_i)| - 0 - T_{Q_2}(u_i)}{\sqrt{1 + |1 - I_1(u_i)|^2}} \\
\frac{2|F_0(u_i) - T_{Q_2}(u_i)|}{\sqrt{1 + |F_0(u_i)|^2}} & \frac{2|1 - F_0(u_i)| - 0 - T_{Q_2}(u_i)}{\sqrt{1 + |1 - F_0(u_i)|^2}}
\end{bmatrix} \geq 0.$$  

Similarly,

$$\begin{bmatrix}
\frac{2|I_1(u_i) - I_{Q_2}(u_i)|}{\sqrt{1 + |I_1(u_i)|^2}} & \frac{2|1 - I_1(u_i)| - 0 - I_{Q_2}(u_i)}{\sqrt{1 + |1 - I_1(u_i)|^2}} \\
\frac{2|F_0(u_i) - F_{Q_2}(u_i)|}{\sqrt{1 + |F_0(u_i)|^2}} & \frac{2|1 - F_0(u_i)| - 0 - F_{Q_2}(u_i)}{\sqrt{1 + |1 - F_0(u_i)|^2}}
\end{bmatrix} \geq 0.$$  

and

$$\begin{bmatrix}
\frac{2|T_{Q_1}(u_i) - T_{Q_2}(u_i)|}{\sqrt{1 + |T_{Q_1}(u_i)|^2}} & \frac{2|1 - T_{Q_1}(u_i)| - 0 - T_{Q_2}(u_i)}{\sqrt{1 + |1 - T_{Q_1}(u_i)|^2}} \\
\frac{2|1 - T_{Q_1}(u_i)|}{\sqrt{1 + |1 - T_{Q_1}(u_i)|^2}} & \frac{2|1 - T_{Q_2}(u_i)|}{\sqrt{1 + |1 - T_{Q_2}(u_i)|^2}}
\end{bmatrix} \geq 0.$$  

Again, for all values of $u_i \in U$, $|T_{Q_1}(u_i)| \geq 0$, $|T_{Q_2}(u_i)| \geq 0$, $|T_{Q_1}(u_i) - T_{Q_2}(u_i)| \geq 0$, $\sqrt{1 + |T_{Q_1}(u_i)|^2} \geq 0$, $\sqrt{1 + |1 - T_{Q_1}(u_i)|^2} \geq 0$, $\sqrt{1 + |1 - T_{Q_2}(u_i)|^2} \geq 0$.

Then,

$$\begin{bmatrix}
\frac{2|T_{Q_1}(u_i) - T_{Q_2}(u_i)|}{\sqrt{1 + |T_{Q_1}(u_i)|^2}} & \frac{2|1 - T_{Q_1}(u_i)| - 0 - T_{Q_2}(u_i)}{\sqrt{1 + |1 - T_{Q_1}(u_i)|^2}} \\
\frac{2|1 - T_{Q_1}(u_i)|}{\sqrt{1 + |1 - T_{Q_1}(u_i)|^2}} & \frac{2|1 - T_{Q_2}(u_i)|}{\sqrt{1 + |1 - T_{Q_2}(u_i)|^2}}
\end{bmatrix} \geq 0.$$  

Similarly, we can show that

$$\begin{bmatrix}
\frac{2|I_1(u_i) - I_{Q_2}(u_i)|}{\sqrt{1 + |I_1(u_i)|^2}} & \frac{2|1 - I_1(u_i)| - 0 - I_{Q_2}(u_i)}{\sqrt{1 + |1 - I_1(u_i)|^2}} \\
\frac{2|F_0(u_i) - F_{Q_2}(u_i)|}{\sqrt{1 + |F_0(u_i)|^2}} & \frac{2|1 - F_0(u_i)| - 0 - F_{Q_2}(u_i)}{\sqrt{1 + |1 - F_0(u_i)|^2}}
\end{bmatrix} \geq 0.$$  

(40)

(41)

(42)

(43)
Adding Equation (39) to Equation (47), and using \( w_i \in [0, 1], \sum_{i=1}^{k} w_i = 1 \), we have \( CE_{\infty}^{n_c}(Q_1, Q_2) \geq 0 \). Hence, this completes the proof.

(ii).

\[
\begin{align*}
&\frac{2\left|T_{01}(u) - T_{02}(u)\right|}{\sqrt{1+\left|T_{01}(u)\right|^2} + \sqrt{1+\left|T_{02}(u)\right|^2}} + \frac{2\left|1-T_{01}(u)\right| - \left|1-T_{02}(u)\right|}{\sqrt{1+\left|1-T_{01}(u)\right|^2} + \sqrt{1+\left|1-T_{02}(u)\right|^2}} = 0, \\
&\therefore T_{01}(u) = T_{02}(u)
\end{align*}
\]

(48)

\[
\begin{align*}
&\frac{2\left|T_{01}(u) - T_{02}(u)\right|}{\sqrt{1+\left|T_{01}(u)\right|^2} + \sqrt{1+\left|T_{02}(u)\right|^2}} + \frac{2\left|1-T_{01}(u)\right| - \left|1-T_{02}(u)\right|}{\sqrt{1+\left|1-T_{01}(u)\right|^2} + \sqrt{1+\left|1-T_{02}(u)\right|^2}} = 0, \\
&\therefore T_{01}(u) = T_{02}(u)
\end{align*}
\]

(49)

and

\[
\begin{align*}
&\frac{2\left|F_{01}(u) - F_{02}(u)\right|}{\sqrt{1+\left|F_{01}(u)\right|^2} + \sqrt{1+\left|F_{02}(u)\right|^2}} + \frac{2\left|1-F_{01}(u)\right| - \left|1-F_{02}(u)\right|}{\sqrt{1+\left|1-F_{01}(u)\right|^2} + \sqrt{1+\left|1-F_{02}(u)\right|^2}} = 0, \\
&\therefore F_{01}(u) = F_{02}(u), \text{ For all values of } u_i \in U.
\end{align*}
\]

(50)

Again,

\[
\begin{align*}
&\frac{2\left|T_{01}(u) - T_{02}(u)\right|}{\sqrt{1+\left|T_{01}(u)\right|^2} + \sqrt{1+\left|T_{02}(u)\right|^2}} + \frac{2\left|1-T_{01}(u)\right| - \left|1-T_{02}(u)\right|}{\sqrt{1+\left|1-T_{01}(u)\right|^2} + \sqrt{1+\left|1-T_{02}(u)\right|^2}} = 0, \\
&\therefore T_{01}(u) = T_{02}(u)
\end{align*}
\]

(51)

\[
\begin{align*}
&\frac{2\left|T_{01}(u) - T_{02}(u)\right|}{\sqrt{1+\left|T_{01}(u)\right|^2} + \sqrt{1+\left|T_{02}(u)\right|^2}} + \frac{2\left|1-T_{01}(u)\right| - \left|1-T_{02}(u)\right|}{\sqrt{1+\left|1-T_{01}(u)\right|^2} + \sqrt{1+\left|1-T_{02}(u)\right|^2}} = 0, \\
&\therefore T_{01}(u) = T_{02}(u)
\end{align*}
\]

(52)
Using Equation (53) to Equation (56) and 
\[ w_i \in [0,1], \sum_{i=1}^{n} w_i = 1, \ w_i \geq 0, \] 
we have \( \text{CE}^\text{NC}_\gamma(Q_1, Q_2) = 0 \) iff \( T_{Q_1}(u_i) = T_{Q_2}(u_i) \), \( I_{Q_1}(u_i) = I_{Q_2}(u_i) \), \( I^*(u_i) = I^*_{Q_2}(u_i) \), \( F_{Q_1}(u_i) = F_{Q_2}(u_i) \), \( F^*_i(u_i) = F^*_{Q_2}(u_i) \) and \( T_{Q_1}(u_i) = T_{Q_2}(u_i) \), \( I_{Q_1}(u_i) = I_{Q_2}(u_i) \), \( F_{Q_1}(u_i) = F_{Q_2}(u_i) \), \( F^*_i(u_i) = F^*_{Q_2}(u_i) \) for all \( u_i \in U \).

(iii).

Using Definition (20), Definition (4), and Definition (10), we obtain the following expression:

\[
\text{CE}^\text{NC}_\gamma(Q', Q') = \frac{1}{8} \sum_{i=1}^{n} w_i \left[ \frac{2|T_{Q_1}(u_i) - T_{Q_2}(u_i)|}{\sqrt{1 + [T_{Q_1}(u_i)]^2 + [T_{Q_2}(u_i)]^2}} + \frac{2|1 - T_{Q_1}(u_i) - 1 - T_{Q_2}(u_i)|}{\sqrt{1 + [1 - T_{Q_1}(u_i)]^2 + [1 - T_{Q_2}(u_i)]^2}} \right] + \frac{2|I_{Q_1}(u_i) - I_{Q_2}(u_i)|}{\sqrt{1 + [I_{Q_1}(u_i)]^2 + [I_{Q_2}(u_i)]^2}} + \frac{2|1 - I_{Q_1}(u_i) - 1 - I_{Q_2}(u_i)|}{\sqrt{1 + [1 - I_{Q_1}(u_i)]^2 + [1 - I_{Q_2}(u_i)]^2}} + \frac{2|I^*_1(u_i) - I^*_2(u_i)|}{\sqrt{1 + [I^*_1(u_i)]^2 + [I^*_2(u_i)]^2}} + \frac{2|1 - I^*_1(u_i) - 1 - I^*_2(u_i)|}{\sqrt{1 + [1 - I^*_1(u_i)]^2 + [1 - I^*_2(u_i)]^2}} + \frac{2|F_{Q_1}(u_i) - F_{Q_2}(u_i)|}{\sqrt{1 + [F_{Q_1}(u_i)]^2 + [F_{Q_2}(u_i)]^2}} + \frac{2|1 - F_{Q_1}(u_i) - 1 - F_{Q_2}(u_i)|}{\sqrt{1 + [1 - F_{Q_1}(u_i)]^2 + [1 - F_{Q_2}(u_i)]^2}} \]

\[ + \frac{2|F^*_1(u_i) - F^*_2(u_i)|}{\sqrt{1 + [F^*_1(u_i)]^2 + [F^*_2(u_i)]^2}} + \frac{2|1 - F^*_1(u_i) - 1 - F^*_2(u_i)|}{\sqrt{1 + [1 - F^*_1(u_i)]^2 + [1 - F^*_2(u_i)]^2}} \]
\[
\mathcal{W}_{\mathcal{R}} \left( \frac{2}{8} \sum_{i=1}^{n} \left[ \frac{2(1-F_{0i}^+)(u_i) - 2(F_{0i}^-(u_i))}{\sqrt{1 + \|F_{0i}^-(u_i)\|^2} + \sqrt{1 + \|F_{0i}^+(u_i)\|^2}} + \frac{2(1-F_{0i}^-)(u_i) - 2(F_{0i}^+(u_i))}{\sqrt{1 + \|F_{0i}^+(u_i)\|^2} + \sqrt{1 + \|F_{0i}^-(u_i)\|^2}} \right] \right) = \frac{1}{8} \left[ \frac{2(1-T_{0i}^+)(u_i) - 2(T_{0i}^-)(u_i)}{\sqrt{1 + \|T_{0i}^-(u_i)\|^2} + \sqrt{1 + \|T_{0i}^+(u_i)\|^2}} + \frac{2(T_{0i}^-(u_i) - T_{0i}^+(u_i))}{\sqrt{1 + \|T_{0i}^+(u_i)\|^2} + \sqrt{1 + \|T_{0i}^-(u_i)\|^2}} \right] + \frac{2(1-T_{0i}^-)(u_i) - 2(T_{0i}^+(u_i))}{\sqrt{1 + \|T_{0i}^+(u_i)\|^2} + \sqrt{1 + \|T_{0i}^-(u_i)\|^2}} + \frac{2(T_{0i}^-(u_i) - T_{0i}^+(u_i))}{\sqrt{1 + \|T_{0i}^+(u_i)\|^2} + \sqrt{1 + \|T_{0i}^-(u_i)\|^2}} + \frac{2(1-F_{0i}^-)(u_i) - 2(F_{0i}^+(u_i))}{\sqrt{1 + \|F_{0i}^+(u_i)\|^2} + \sqrt{1 + \|F_{0i}^-(u_i)\|^2}} + \frac{2(F_{0i}^-(u_i) - F_{0i}^+(u_i))}{\sqrt{1 + \|F_{0i}^+(u_i)\|^2} + \sqrt{1 + \|F_{0i}^-(u_i)\|^2}} + \frac{2(1-F_{0i}^+)(u_i) - 2(F_{0i}^-)(u_i))}{\sqrt{1 + \|F_{0i}^-\|^2} + \sqrt{1 + \|F_{0i}^+\|^2}} + \frac{2(F_{0i}^-)(u_i) - F_{0i}^+(u_i))}{\sqrt{1 + \|F_{0i}^+\|^2} + \sqrt{1 + \|F_{0i}^-\|^2}} + \frac{2(1-T_{0i}^-)(u_i) - 2(T_{0i}^+(u_i))}{\sqrt{1 + \|T_{0i}^+(u_i)\|^2} + \sqrt{1 + \|T_{0i}^-(u_i)\|^2}} + \frac{2(T_{0i}^-)(u_i) - T_{0i}^+(u_i))}{\sqrt{1 + \|T_{0i}^+\|^2} + \sqrt{1 + \|T_{0i}^-\|^2}}. \right]
\]
\[
\begin{align*}
&= \frac{1}{8} \sum_{i=1}^{n} \left( \frac{2|I_1(u) - I_2(u)|}{1 + |I_1(u)|^2 + 1 + |I_2(u)|^2} + \frac{2|I_2(u) - I_2^*(u)|}{1 + |I_1(u)|^2 + 1 + |I_2(u)|^2} \\
&\quad + \frac{2|F_1(u) - F_2(u)|}{1 + |F_1(u)|^2 + 1 + |F_2(u)|^2} + \frac{2|F_2(u) - F_2^*(u)|}{1 + |F_1(u)|^2 + 1 + |F_2(u)|^2} \\
&\quad + \frac{2|I_1(u) - I_2(u)|}{1 + |I_1(u)|^2 + 1 + |I_2(u)|^2} + \frac{2|I_2(u) - I_2^*(u)|}{1 + |I_1(u)|^2 + 1 + |I_2(u)|^2} \\
&\quad + \frac{2|F_1(u) - F_2(u)|}{1 + |F_1(u)|^2 + 1 + |F_2(u)|^2} + \frac{2|F_2(u) - F_2^*(u)|}{1 + |F_1(u)|^2 + 1 + |F_2(u)|^2} \\
&\quad + \frac{2|I_1(u) - I_2(u)|}{1 + |I_1(u)|^2 + 1 + |I_2(u)|^2} + \frac{2|I_2(u) - I_2^*(u)|}{1 + |I_1(u)|^2 + 1 + |I_2(u)|^2} \\
&\quad + \frac{2|F_1(u) - F_2(u)|}{1 + |F_1(u)|^2 + 1 + |F_2(u)|^2} + \frac{2|F_2(u) - F_2^*(u)|}{1 + |F_1(u)|^2 + 1 + |F_2(u)|^2} \right) = CE_{\Phi}(Q_1, Q_2), \forall u \in U.
\end{align*}
\]

(iv).

Since \( \forall u \in U \), for single valued parts, we obtain:
\[
\begin{align*}
&\left| T_{O_1}(u) - T_{O_2}(u) \right| = \left| T_{O_2}(u) - T_{O_1}(u) \right| , \quad \left| I_{O_1}(u) - I_{O_2}(u) \right| = \left| I_{O_2}(u) - I_{O_1}(u) \right| , \quad \left| F_{O_1}(u) - F_{O_2}(u) \right| = \left| F_{O_2}(u) - F_{O_1}(u) \right| , \\
&\left| (1 - T_{O_1}(u)) - (1 - T_{O_2}(u)) \right| = \left| (1 - T_{O_2}(u)) - (1 - T_{O_1}(u)) \right| , \quad \left| (1 - I_{O_1}(u)) - (1 - I_{O_2}(u)) \right| = \left| (1 - I_{O_2}(u)) - (1 - I_{O_1}(u)) \right| , \\
&\left| (1 - F_{O_1}(u)) - (1 - F_{O_2}(u)) \right| = \left| (1 - F_{O_2}(u)) - (1 - F_{O_1}(u)) \right| .
\end{align*}
\]
Then, we obtain
\[
\sqrt{1 + |T_{O_1}(u)|^2} + \sqrt{1 + |T_{O_2}(u)|^2} = \sqrt{1 + |T_{O_2}(u)|^2} + \sqrt{1 + |T_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |I_{O_1}(u)|^2} + \sqrt{1 + |I_{O_2}(u)|^2} = \sqrt{1 + |I_{O_2}(u)|^2} + \sqrt{1 + |I_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |F_{O_1}(u)|^2} + \sqrt{1 + |F_{O_2}(u)|^2} = \sqrt{1 + |F_{O_2}(u)|^2} + \sqrt{1 + |F_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |(1-T_{O_1}(u))|^2} + \sqrt{1 + |(1-T_{O_2}(u))|^2} = \sqrt{1 + |(1-T_{O_2}(u))|^2} + \sqrt{1 + |(1-T_{O_1}(u))|^2},
\]
\[
\sqrt{1 + |(1-I_{O_1}(u))|^2} + \sqrt{1 + |(1-I_{O_2}(u))|^2} = \sqrt{1 + |(1-I_{O_2}(u))|^2} + \sqrt{1 + |(1-I_{O_1}(u))|^2},
\]
\[
\sqrt{1 + |(1-F_{O_1}(u))|^2} + \sqrt{1 + |(1-F_{O_2}(u))|^2} = \sqrt{1 + |(1-F_{O_2}(u))|^2} + \sqrt{1 + |(1-F_{O_1}(u))|^2},
\]
\[
\forall u_i \in U.
\]

For interval neutrosophic part, we have
\[
|T_{O_1}(u)| - |T_{O_2}(u)| = |T_{O_2}(u)| - |T_{O_1}(u)|,
\]
\[
|I_{O_1}(u)| - |I_{O_2}(u)| = |I_{O_2}(u)| - |I_{O_1}(u)|,
\]
\[
|F_{O_1}(u)| - |F_{O_2}(u)| = |F_{O_2}(u)| - |F_{O_1}(u)|,
\]
\[
|(1-T_{O_1}(u))| - |(1-T_{O_2}(u))| = |(1-T_{O_2}(u))| - |(1-T_{O_1}(u))|,
\]
\[
|(1-I_{O_1}(u))| - |(1-I_{O_2}(u))| = |(1-I_{O_2}(u))| - |(1-I_{O_1}(u))|,
\]
\[
|(1-F_{O_1}(u))| - |(1-F_{O_2}(u))| = |(1-F_{O_2}(u))| - |(1-F_{O_1}(u))|.
\]

Then, we obtain
\[
\sqrt{1 + |T_{O_1}(u)|^2} + \sqrt{1 + |T_{O_2}(u)|^2} = \sqrt{1 + |T_{O_2}(u)|^2} + \sqrt{1 + |T_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |I_{O_1}(u)|^2} + \sqrt{1 + |I_{O_2}(u)|^2} = \sqrt{1 + |I_{O_2}(u)|^2} + \sqrt{1 + |I_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |F_{O_1}(u)|^2} + \sqrt{1 + |F_{O_2}(u)|^2} = \sqrt{1 + |F_{O_2}(u)|^2} + \sqrt{1 + |F_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |(1-T_{O_1}(u))|^2} + \sqrt{1 + |(1-T_{O_2}(u))|^2} = \sqrt{1 + |(1-T_{O_2}(u))|^2} + \sqrt{1 + |(1-T_{O_1}(u))|^2},
\]
\[
\sqrt{1 + |(1-I_{O_1}(u))|^2} + \sqrt{1 + |(1-I_{O_2}(u))|^2} = \sqrt{1 + |(1-I_{O_2}(u))|^2} + \sqrt{1 + |(1-I_{O_1}(u))|^2},
\]
\[
\sqrt{1 + |(1-F_{O_1}(u))|^2} + \sqrt{1 + |(1-F_{O_2}(u))|^2} = \sqrt{1 + |(1-F_{O_2}(u))|^2} + \sqrt{1 + |(1-F_{O_1}(u))|^2},
\]
\[
\forall u_i \in U.
\]

Similarly,
\[
|T_{O_1}(u)| - |T_{O_2}(u)| = |T_{O_2}(u)| - |T_{O_1}(u)|,
\]
\[
|I_{O_1}(u)| - |I_{O_2}(u)| = |I_{O_2}(u)| - |I_{O_1}(u)|,
\]
\[
|F_{O_1}(u)| - |F_{O_2}(u)| = |F_{O_2}(u)| - |F_{O_1}(u)|,
\]
\[
|(1-T_{O_1}(u))| - |(1-T_{O_2}(u))| = |(1-T_{O_2}(u))| - |(1-T_{O_1}(u))|,
\]
\[
|(1-I_{O_1}(u))| - |(1-I_{O_2}(u))| = |(1-I_{O_2}(u))| - |(1-I_{O_1}(u))|,
\]
\[
|(1-F_{O_1}(u))| - |(1-F_{O_2}(u))| = |(1-F_{O_2}(u))| - |(1-F_{O_1}(u))|,
\]
then
\[
\sqrt{1 + |T_{O_1}(u)|^2} + \sqrt{1 + |T_{O_2}(u)|^2} = \sqrt{1 + |T_{O_2}(u)|^2} + \sqrt{1 + |T_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |I_{O_1}(u)|^2} + \sqrt{1 + |I_{O_2}(u)|^2} = \sqrt{1 + |I_{O_2}(u)|^2} + \sqrt{1 + |I_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |F_{O_1}(u)|^2} + \sqrt{1 + |F_{O_2}(u)|^2} = \sqrt{1 + |F_{O_2}(u)|^2} + \sqrt{1 + |F_{O_1}(u)|^2},
\]
\[
\sqrt{1 + |(1-T_{O_1}(u))|^2} + \sqrt{1 + |(1-T_{O_2}(u))|^2} = \sqrt{1 + |(1-T_{O_2}(u))|^2} + \sqrt{1 + |(1-T_{O_1}(u))|^2},
\]
\[
\sqrt{1 + |(1-I_{O_1}(u))|^2} + \sqrt{1 + |(1-I_{O_2}(u))|^2} = \sqrt{1 + |(1-I_{O_2}(u))|^2} + \sqrt{1 + |(1-I_{O_1}(u))|^2},
\]
\[
\sqrt{1 + |(1-F_{O_1}(u))|^2} + \sqrt{1 + |(1-F_{O_2}(u))|^2} = \sqrt{1 + |(1-F_{O_2}(u))|^2} + \sqrt{1 + |(1-F_{O_1}(u))|^2},
\]
\[
\forall u_i \in U.
\]

In addition, \( w_i \in [0, 1], \sum_{i=1}^{n} w_i = 1, w_i \geq 0. \)

Thus, \( CE^{\text{w}}_{\text{NC}} (Q_p, Q_q) = CE^{\text{w}}_{\text{NC}} (Q_q, Q_p). \)

hence completing the proof. \(\square\)

4. MADM Strategy Using Proposed NC-Cross Entropy Measure in the NCS Environment
In this section, we develop an MADM strategy using the proposed NC-cross entropy measure.

**Description of the MADM problem:**

Let \( A = \{A_1, A_2, A_3, \ldots, A_m\} \) and \( G = \{G_1, G_2, G_3, \ldots, G_n\} \) be the discrete set of alternatives and attribute, respectively. Let \( W = \{w_1, w_2, w_3, \ldots, w_n\} \) be the weight vector of attributes \( G_j \) (\( j = 1, 2, 3, \ldots, n \)), where \( w_j \geq 0 \) and \( \sum_{j=1}^{n} w_j = 1 \).

Now, we describe the steps of MADM strategy using NC-cross entropy measure.

**Step: 1. Formulate the decision matrices**

For MADM with neutrosophic cubic information, the rating values of the alternatives \( A_i \) (\( i = 1, 2, 3, \ldots, m \)) on the basis of criterion \( G_j \) (\( j = 1, 2, 3, \ldots, n \)) by the decision-maker can be expressed in NCN as \( a_{ij} = [T_{ij}^+, T_{ij}^0, [I_{ij}^+, I_{ij}^0], [F_{ij}^+, F_{ij}^0]] \) (\( T_{ij}^+, I_{ij}^+, F_{ij}^+ \) > (\( i = 1, 2, 3, \ldots, m; j = 1, 2, 3, \ldots, n \)). We present these rating values of alternatives provided by the decision-maker in matrix form as follows:

\[
M = \begin{pmatrix}
G_1 & G_2 & \ldots & G_n \\
A_1 & a_{11} & a_{12}^* & \ldots & a_{1n}^* \\
A_2 & a_{21} & a_{22} & \ldots & a_{2n} \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
A_m & a_{m1} & a_{m2}^* & \ldots & a_{mn}
\end{pmatrix}
\]  

(57)

**Step: 2. Formulate priori/ideal decision matrix**

In the MADM process, the priori decision matrix is used to select the best alternative from the set of feasible alternatives. In the decision-making situation, we use the following decision matrix as priori decision matrix.

\[
P^* = \begin{pmatrix}
G_1 & G_2 & \ldots & G_n \\
A_1 & a_{11}^* & a_{12} & \ldots & a_{1n} \\
A_2 & a_{21} & a_{22} & \ldots & a_{2n} \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
A_m & a_{m1} & a_{m2} & \ldots & a_{mn}^*
\end{pmatrix}
\]  

(58)

where, \( a_{ij}^* = [1, 1]_i, [0, 0], [0, 0] \) for benefit type attributes and \( a_{ij}^* = [0, 0], [1, 1], [1, 1] \) for cost type attributes, (\( i = 1, 2, 3, \ldots, m; j = 1, 2, 3, \ldots, n \)).

**Step: 3. Formulate the weighted NC-cross entropy matrix**

Using Equation (38), we calculate weighted NC-cross entropy values between decision matrix and priori matrix. The cross entropy value can be presented in matrix form as follows:

\[
M_{NC}^w = \begin{pmatrix}
CE_{NC}^w (A_1) \\
\vdots \\
CE_{NC}^w (A_2) \\
\vdots \\
\vdots \\
CE_{NC}^w (A_m)
\end{pmatrix}
\]  

(59)

**Step: 4. Rank the priority**

Smaller value of the cross entropy reflects that an alternative is closer to the ideal alternative. Therefore, the preference ranking order of all the alternatives can be determined according to the increasing order of the cross entropy values \( CE_{NC}^w (A_i) \) (\( i = 1, 2, 3, \ldots, m \)). The smallest cross entropy value reflects the best alternative and the greatest cross entropy value reflects the worst alternative.

A conceptual model of the proposed strategy is shown in Figure 1.
5. Illustrative Example

In this section, we solve an illustrative example of an MADM problem to reflect the feasibility and efficiency of our proposed strategy in NCSs environments.

Now, we use an example [89] for cultivation and analysis. A venture capital firm intends to make evaluation and selection to five enterprises with the investment potential:

1. Automobile company (A1)
2. Military manufacturing enterprise (A2)
3. TV media company (A3)
4. Food enterprises (A4)
5. Computer software company (A5)

On the basis of four attributes namely:

1. Social and political factor (G1)
2. The environmental factor (G2)
3. Investment risk factor (G3)
4. The enterprise growth factor (G4).

Weight vector of attributes is \( W = \{ .24, .25, .23, .28 \} \).

The steps of decision-making strategy to rank alternatives are presented as follows:

**Step 1. Formulate the decision matrix**
The decision-maker represents the rating values of alternative $A_i$ ($i = 1, 2, 3, 4, 5$) with respect to the attribute $G_j$ ($j = 1, 2, 3, 4$) in terms of NCNs and constructs the decision matrix $M$ as follows:

\[
M = \begin{bmatrix}
A_1 & <[6.8],[2.3],[3.4],[4.4],[5.4]>,<[6.8],[2.3],[4.4],[5.4]>,<[6.8],[2.3],[3.4],[4.4]>,<[6.8],[2.3],[4.4],[5.4]>,<[6.8],[2.3],[2.4],[4.4]>
A_2 & <[5.7],[2.3],[3.4],[4.4]>,<[5.7],[2.3],[5.3],[4.4]>,<[5.7],[2.3],[2.4],[4.4]>,<[5.7],[2.3],[4.4],[5.3]>,<[5.7],[2.3],[2.4],[4.4]>
A_3 & <[6.8],[2.3],[3.4],[4.4]>,<[6.8],[2.3],[4.4],[5.3]>,<[6.8],[2.3],[3.4],[4.4]>,<[6.8],[2.3],[2.4],[4.4]>,<[6.8],[2.3],[3.4],[4.4]>
A_4 & <[5.7],[2.3],[3.4],[4.4]>,<[5.7],[2.3],[5.3],[4.4]>,<[5.7],[2.3],[2.4],[4.4]>,<[5.7],[2.3],[4.4],[5.3]>,<[5.7],[2.3],[2.4],[4.4]>
A_5 & <[0.8],[2.3],[3.4],[4.4]>,<[0.8],[2.3],[4.4],[5.3]>,<[0.8],[2.3],[3.4],[4.4]>,<[0.8],[2.3],[2.4],[4.4]>,<[0.8],[2.3],[3.4],[4.4]>
\end{bmatrix}
\]

(60)

**Step: 3. Formulate priori/ideal decision matrix**

Priori/ideal decision matrix

\[
M^1 = \begin{bmatrix}
G_1 & G_2 & G_3 & G_4
\end{bmatrix}
\begin{bmatrix}
A_1 & <[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>
A_2 & <[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>
A_3 & <[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>
A_4 & <[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>
A_5 & <[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>,<[1,1],[0,0],[0,0],[1,0],0,0>
\end{bmatrix}
\]

(61)

**Step: 4. Calculate the weighted INS cross entropy matrix**

Using Equation (38), we calculate weighted NC-cross entropy values between ideal matrixes (61) and decision matrix (60):

\[
M^w_{CE, NC} = \begin{bmatrix}
0.66 \\
0.58 \\
0.60 \\
0.74 \\
0.71
\end{bmatrix}
\]

(62)

**Step: 5. Rank the priority**

The position of cross entropy values of alternatives arranging in increasing order is 0.58 < 0.60 < 0.66 < 0.71 < 0.74. Since the smallest values of cross entropy indicate that the alternative is closer to the ideal alternative, the ranking priority of alternatives is $A_2 > A_3 > A_1 > A_5 > A_4$. Hence, military manufacturing enterprise ($A_2$) is the best alternative for investment.

Graphical representation of alternatives versus cross entropy is shown in Figure 2. From the Figure 2, we see that $A_2$ is the best preference alternative and $A_5$ is the least preference alternative.

![Figure 2. Bar diagram of alternatives versus cross entropy values of alternatives.](image-url)
Figure 3 presents relation between cross entropy value and preference ranking of the alternative.

![Graphical representation of cross entropy values and ranking of alternatives.](image)

Figure 3. Graphical representation of cross entropy values and ranking of alternatives.

6. Contributions of the Paper

The contributions of the paper are summarized as follows:

1. We have introduced an NC-cross entropy measure and proved its basic properties in NCS environments.
2. We have introduced a weighted NC-cross entropy measure and proved its basic properties in NCS environments.
3. We have developed a novel MADM strategy based on weighted NC-cross entropy to solve MADM problems.
4. We solved an illustrative example of MADM problem using proposed strategies.

7. Conclusions

We have introduced NC-cross entropy measure in NCS environments. We have proved the basic properties of the proposed NC-cross entropy measure. We have also introduced weighted NC-cross entropy measure and established its basic properties. Using the weighted NC-cross entropy measure, we developed a novel MADM strategy. We have also solved an MADM problem to illustrate the proposed MADM strategy. The proposed NC-cross entropy based MADM strategy can be employed to solve a variety of problems such as logistics center selection [90,91], weaver selection [92], teacher selection [21], brick selection [93], renewable energy selection [94], etc. The proposed NC-cross entropy based MADM strategy can also be extended to MAGDM strategy using suitable aggregation operators.

Acknowledgments: The authors are very grateful to the anonymous reviewers for their insightful and constructive comments and suggestions that have led to an improved version of this paper.

Author Contributions: Surapati Pramanik conceived and designed the problem; Surapati Pramanik and Shyamal Dalapati solved the problem; Surapati Pramanik, Shariful Alam, Florentin Smarandache and Tapan Kumar Roy analyzed the results; Surapati Pramanik and Shyamal Dalapati wrote the paper.

Conflicts of Interest: The authors declare no conflicts of interest.
References


