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A Neutrosophic Algebraic Framework for Modeling the Interactive Communication of Ethnic Music in the Internet Era

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Abstract: In the era of globalized digital media, ethnic music is increasingly shared and consumed through online platforms. However, the reception of such music is rarely uniform: it varies across cultural, cognitive, and emotional dimensions. Traditional statistical or binary models are incapable of capturing the simultaneous truth, ambiguity, contradiction, and rejection often found in listener interpretations of ethnic music. In this paper, we propose a novel algebraic and quadripartitioned neutrosophic framework to model the interactive communication of ethnic music in digital environments. The model introduces new operators to handle the uncertainty and opposition present in listener feedback, defines the alignment and dissonance measures between musical intent and user perception, and provides a robust case study using simulated data for a traditional Kurdish folk song. This approach not only enhances the mathematical modeling of cultural media reception but also supports practical applications such as adaptive music recommendation, audience segmentation, and cultural feedback diagnosis on platforms like YouTube and Spotify.

Keywords: Neutrosophic Logic, Quadripartitioned Sets, Ethnic Music, Interactive Communication, Cultural Reception Modeling, Neutrosophic Algebra, Digital Music Platforms.

1. Introduction

In today's digital world, ethnic music spreads far beyond its local roots through platforms like YouTube, Spotify, and TikTok. These platforms let traditional music, which is often deeply tied to specific cultural, linguistic, or spiritual backgrounds, reach listeners who may not understand its structure, symbols, or intended meaning [1]. This leads to complex and sometimes unclear reactions. Listeners might partly enjoy the music, feel confused, misinterpret it, or have mixed emotions [2].

Unlike popular or mainstream music, ethnic music is closely linked to cultural identity. It often includes non-Western scales, lyrics that can't be easily translated, ritual meanings, or symbolic instruments, which may not align with how global audiences interpret music [3]. When this music is shared online, these platforms become more than just a way to share—they turn into spaces where cultural meanings are created, debated, or misunderstood in real time [4].

However, most systems are used to measure how people respond to music—like counting likes, views, or using sentiment analysis—rely on simple, one-sided logic. These methods struggle to capture the conflicting, uncertain, or layered reactions that ethnic music often sparks [5]. For example, someone might love the vocal style but feel uneasy about the religious themes or enjoy the melody but feel disconnected because of an unfamiliar language or cultural context [6].

To address this gap, we propose a new approach using neutrosophic logic, developed by Smarandache. This logic goes beyond just true or false, including uncertainty and even contradiction as separate aspects [7]. In this paper, we use a special model called a quadripartitioned neutrosophic set, enhanced with mathematical structures, to better understand the interaction between what the music intends and how listeners respond [8].

The main goals of this paper are:

- 1) To create new mathematical tools and functions designed for studying how ethnic music communicates with listeners.
- 2) To develop ways to measure how well the music aligns with individual listeners' perspectives.
- 3) To introduce a Cultural Dissonance Metric (CDM) that measures misunderstandings or interpretive failures.
- 4) To test this model with a case study on Kurdish ethnic music shared on a digital platform [9].

This framework offers a new path for media informatics, cultural analytics, and music technology by providing a precise, understandable, and multidimensional way to study listener feedback, moving beyond traditional binary or fuzzy approaches [10].

2. Neutrosophic Preliminaries and Algebraic Definitions

This section introduces the mathematical foundations required to model the uncertain, contradictory, and context-sensitive communication of ethnic music in digital environments. We construct the system using QNS, extend them with Neutrosophic Inclusion Equations, and define algebraic operations that will support audience classification and semantic misalignment detection.

2.1 Quadripartitioned Neutrosophic Sets (QNS)

Let *X* be the universe of discourse (e.g., a set of listeners or musical elements). A Quadripartitioned Neutrosophic Set (QNS) *A* on *X* is defined as:

$$A = \{(x, T_A(x), C_A(x), I_A(x), F_A(x)) \mid x \in X\}$$

Where for each $x \in X$, $T_A(x) \in [0,1]$: degree of truth or alignment, $C_A(x) \in [0,1]$: degree of contradiction,

 $I_A(x) \in [0,1]$: degree of indeterminacy or ambiguity,

 $F_A(x) \in [0,1]$: degree of falsity or rejection, with the constraint:

 $0 \le T_{4}(x) + C_{4}(x) + I_{4}(x) + F_{4}(x) \le 4$

This structure allows modeling of multi-dimensional listener reactions such as: "I liked the melody (T), but I also felt it conflicted with my values (C), and I didn't fully understand it (I)."

2.2 Neutrosophic Alignment Score (NAS)

To determine how much a user's perception aligns with the artist's communicative intent, we define the Neutrosophic Alignment Score:

NAS(x) = T(x) - [C(x) + I(x) + F(x)]

- 1. If NAS(x) > 0.5: high resonance,
- 2. If $0 < NAS(x) \le 0.5$: partial resonance,

3. If $NAS(x) \le 0$: low or negative resonance.

2.3 Appurtenance-Inclusion Equation (AIE)

Let:

 $M = (T_M, C_M, I_M, F_M)$ be the music intent profile,

 $R = (T_x, C_x, I_x, F_x)$ be the listener's response,

Then the degree to which the listener's perception is included in the music's intended space is given by:

$$AIE(x, M) = \frac{T_x \cdot T_M - C_x \cdot C_M - F_x \cdot F_M}{1 + I_x + I_M}$$

This equation handles partial agreement, contradiction, and uncertainty in a unified structure.

2.4 Cultural Dissonance Measure (CDM)

We define the Cultural Dissonance Measure to capture interpretive failure:

$$CDM(x) = \sqrt{(F_x - T_M)^2 + C_x^2 + I_x^2}$$

Higher values of CDM indicate that the listener's reaction diverges from the music's intended emotion or symbolism.

2.5 Neutrosophic Union and Intersection

Given two QNSs A and B :

Union,

$$T_{A\cup B}(x) = \max(T_A(x), T_B(x)), \quad C_{A\cup B}(x) = \min(C_A(x), C_B(x))$$

$$I_{A\cup B}(x) = \min(I_A(x), I_B(x)), \quad F_{A\cup B}(x) = \min(F_A(x), F_B(x))$$

Intersection,

$$T_{A \cap B}(x) = \min(T_A(x), T_B(x)), C_{A \cap B}(x) = \max(C_A(x), C_B(x))$$

$$I_{A \cap B}(x) = \max(I_A(x), I_B(x)), F_{A \cap B}(x) = \max(F_A(x), F_B(x))$$

These operations are useful when comparing audience subgroups or aggregating platform feedback.

2.6 Neutrosophic Algebraic d-Structure

Let (Z, \star) be a neutrosophic algebraic system, where:

Z : the set of all listener responses,

* : a binary operation defined as a weighted perception composition, e.g.,

$$x \star y = \left(\alpha T_x + (1 - \alpha)T_y, \beta C_x + (1 - \beta)C_y, \dots\right)$$

We define a subset $G \subseteq Z$ as a neutrosophic d-ideal if:

$$\forall x, y \in Z: y \in G \text{ and } x \star y \in G \Rightarrow x \in G$$

Such d-ideals correspond to listener clusters that are closed under interpretive blending, useful for audience recommendation engines.

2.7 Illustrative Example

Let the music intent vector be:

$$M = (0.90, 0.05, 0.03, 0.02)$$

Let listener *x* respond:

$$R_x = (0.65, 0.15, 0.10, 0.10)$$

Then:

NAS=

$$NAS(x) = 0.65 - (0.15 + 0.10 + 0.10) = 0.30$$

AIE=

$$AIE(x, M) = \frac{0.65 \cdot 0.90 - 0.15 \cdot 0.05 - 0.10 \cdot 0.02}{1 + 0.10 + 0.03} = \frac{0.585 - 0.0075 - 0.002}{1.13} = \frac{0.5755}{1.13}$$

CDM=

$$CDM(x) = \sqrt{(0.10 - 0.90)^2 + 0.15^2 + 0.10^2} = \sqrt{0.64 + 0.0225 + 0.01} = \sqrt{0.6725} \approx 0.820$$

These scores indicate moderate alignment but a high risk of cultural dissonance.

3. Neutrosophic Framework for Interactive Communication of Ethnic Music

This section introduces the proposed system that models how ethnic music is communicated, received, and cognitively processed by digital audiences. The framework uses the tools from neutrosophic algebra to construct a three-layer interactive model: the music intent layer, the listener response layer, and the evaluation/diagnostic layer.

3.1 Layer 1: Music Intent Modeling

Let $M = (T_M, C_M, I_M, F_M)$ represent the intent vector of an ethnic music piece. This vector captures the artist's expressive and cultural objectives. Each component is interpreted as:

 T_M : degree of emotional and cultural clarity in the music.

 C_M : intentional contradiction (e.g., irony, dual meaning).

 I_M : built-in ambiguity (e.g., abstract lyrics).

 F_M : elements that may provoke opposition or rejection.

For example, a Kurdish folk song might encode M = (0.90, 0.05, 0.03, 0.02) indicating a very clear, consistent message with low contradiction or risk of rejection.

3.2 Layer 2: Listener Response Modeling

Each listener *x* is modeled with a neutrosophic quadripartitioned vector:

$$R_x = (T_x, C_x, I_x, F_x)$$

This captures their actual cognitive-emotional interpretation after engaging in the music. The values can be sourced through surveys, feedback metadata, or inferred via behavior on digital platforms.

Example: $R_{U3} = (0.45, 0.30, 0.15, 0.10)$

3.3 Step 1: Compute Neutrosophic Alignment Score (NAS)

This metric evaluates how much the listener agrees with the music's intended meaning and emotional tone.

The formula is:

$$NAS(x) = T_x - (C_x + I_x + F_x)$$

Where,

 T_x represents the listener's degree of agreement,

 C_x reflects internal contradiction,

 I_x captures uncertainty or confusion,

 F_x measures rejection or disagreement.

Clarification scale:

If NAS > 0.5, the listener is highly aligned.

If $0 < NAS \le 0.5$, the listener is partially aligned.

If NAS \leq 0, the listener is unaligned or misaligned.

Example:

Let the listener's response vector be:

$$R_{U3} = (T = 0.45, C = 0.30, I = 0.15, F = 0.10)$$

Applying the formula:

$$NAS(U3) = 0.45 - (0.30 + 0.15 + 0.10) = 0.45 - 0.55 = -0.10$$

This result indicates clear misalignment. The listener partially agrees but also feels conflicted, uncertain, and somewhat rejects the musical content.

3.4 Step 2: Evaluate Appurtenance-Inclusion Equation (AIE)

This metric measures how much the listener's perception is included within the space of the music's intended meaning.

The formula is:

$$AIE(x, M) = \frac{T_x \cdot T_M - C_x \cdot C_M - F_x \cdot F_M}{1 + I_x + I_M}$$

Where,

 T_M, C_M, I_M, F_M are the components of the music's intent vector,

 T_x , C_x , I_x , F_x are the listener's responses.

Example:

Let the music's intent vector be:

$$M = (T_M = 0.90, C_M = 0.05, I_M = 0.03, F_M = 0.02)$$

Using the same listener response:

$$R_{U3} = (T_x = 0.45, C_x = 0.30, I_x = 0.15, F_x = 0.10)$$

Now compute:

$$AIE(U3) = \frac{0.45 \cdot 0.90 - 0.30 \cdot 0.05 - 0.10 \cdot 0.02}{1 + 0.15 + 0.03}$$
$$= \frac{0.405 - 0.015 - 0.002}{1.18} = \frac{0.388}{1.18} \approx 0.329$$

This is a low inclusion score, meaning that the listener's understanding and emotional response do not fit well within the music's intended cultural or expressive meaning.

3.5 Step 3: Calculate the Cultural Dissonance Measure (CDM)

The CDM quantifies how far the listener's interpretation deviates from the music's intended cultural and emotional meaning. It is useful when identifying misunderstandings, cultural mismatches, or cognitive disconnection.

The formula is:

$$CDM(x) = \sqrt{(F_x - T_M)^2 + C_x^2 + I_x^2}$$

Where:

 F_x is the listener's degree of rejection,

 T_M is the truth component in the music's intent,

 C_x and I_x are contradiction and indeterminacy from the listener.

A higher CDM value means a greater level of dissonance between the listener and the cultural message of the music.

Example:

Let the same listener U_3 respond with:

$$R_{U3} = (T = 0.45, C = 0.30, I = 0.15, F = 0.10)$$

And let the music's intent be:

$$M = (T_M = 0.90, C_M = 0.05, I_M = 0.03, F_M = 0.02)$$

Now apply the formula:

 $CDM(U3) = \sqrt{(0.10 - 0.90)^2 + 0.30^2 + 0.15^2}$ $= \sqrt{(-0.80)^2 + 0.09 + 0.0225} = \sqrt{0.64 + 0.09 + 0.0225} = \sqrt{0.7525} \approx 0.867$

This value indicates a high level of cultural dissonance. Although the listener expressed some truth (agreement), the dominant feelings of contradiction, confusion, and disconnect show that the intended message was not effectively received.

3.6 Step 4: Grouping Listeners Using Neutrosophic d-Ideals

In this step, we define how to group digital music listeners into clusters based on their neutrosophic behavior. This helps platforms identify:

- 1. Who resonates with the music,
- 2. Who is confused or conflicted,
- 3. And who feels disconnected.

We do this using a neutrosophic algebraic structure known as a d-ideal.

Definition: Neutrosophic d-Ideal

Let *Z* be the set of all listener responses, where each listener $x \in Z$ is represented by a 4-tuple:

$$R_x = (T_x, C_x, I_x, F_x)$$

We define a binary operation ***** between two listeners as:

$$x \star y = \left(\alpha T_x + (1-\alpha)T_y, \beta C_x + (1-\beta)C_y, \gamma I_x + (1-\gamma)I_y, \delta F_x + (1-\delta)F_y\right)$$

Where α , β , γ , $\delta \in [0,1]$ are weights chosen based on the influence level of each component (truth, contradiction, etc.).

A subset $G \subseteq Z$ is called a neutrosophic d-ideal if it satisfies the following closure property:

If $y \in G$ and $x \star y \in G$, then $x \in G$

This means: if blending two listeners' responses keeps the result inside the group, then both are considered part of the same interpretive cluster.

Example: Cultural Resonance Group

Let: $R_A = (0.80, 0.10, 0.05, 0.05)$ $R_B = (0.60, 0.15, 0.15, 0.10)$ Use weights: $\alpha = \beta = \gamma = \delta = 0.5$

Then:

 $R_A \star R_B = (0.5 \cdot 0.80 + 0.5 \cdot 0.60, 0.5 \cdot 0.10 + 0.5 \cdot 0.15, 0.5 \cdot 0.05 + 0.5 \cdot 0.15, 0.5 \cdot 0.05 + 0.5 \cdot 0.10) = (0.70, 0.125, 0.10, 0.075)$

If $R_A \star R_B \in G$, and $R_B \in G$, then $R_A \in G$ as well. This shows both users belong to the same cultural resonance group and may receive similar content recommendations.

By applying this rule across all user responses, we can:

- 1. Create groups of listeners with shared cognitive-emotional profiles.
- 2. Support platform decisions for content personalization.
- 3. Detect outliers who might require content filtering or support.

3.7 Step 5: Neutrosophic Decision Table and Strategic Scenarios

Once we have calculated the key neutrosophic metrics NAS (alignment), AIE (inclusion), and CDM (dissonance) and possibly grouped users using d-ideals, the next step is to make practical decisions for music platforms. These decisions help platforms:

- a) Improve listener experience,
- b) Respect cultural boundaries,
- c) Increase engagement and reduce drop-off from misaligned users.

We classify each listener into one of three categories using their NAS, AIE, and CDM values illustrated Table 1.

Table 1. Neutrosophic Listener Classification							
Listener	NAS	AIE	CDM	Interpretation			
Category							
Strong	> 0.5	> 0.5	< 0.6	Listener is highly aligned and			
Resonance				included			
Partial Fit	$0 < NAS \le$	$0.3 \le AIE \le$	$0.6 \le CDM \le$	Listener shows interest but needs			
	0.5	0.5	0.8	support			
Misaligned	$NAS \le 0$	AIE < 0.3	CDM > 0.8	Listener is disconnected or confused			

Table 1. Neutrosophic Listener Classification

Each category is tied to adaptive strategies for music delivery see Table 2.

Table 2. Platform Response Strategies				
Category	Platform Action			
Strong Resonance	Promote similar ethnic music, allow user sharing, no modification needed			
Partial Fit	Add subtitle overlays, context info, and cultural cues (e.g., artist notes)			
Misaligned	Suggest culturally neutral or transitional content; avoid direct repetition			

Example Scenario: Listener U3 Let's take Listener U_3 again: NAS = $-0.10 \equiv$ misaligned

AIE $\approx 0.329 \equiv \text{low inclusion}$

 $CDM \approx 0.867 \equiv high dissonance$

Recommended platform strategy:

- 1. Do not repeat similar music directly.
- 2. Offer a guided pathway e.g., storytelling, translated lyrics, or familiar musical structures before returning to ethnic content.

This strategic decision matrix allows platforms to move from passive observation to active, personalized listener management, driven by logical, mathematical metrics.

4. Neutrosophic Evaluation of Listener Reception to Ethnic Music Online

This section presents a simulated case study applying the proposed neutrosophic framework to analyze user responses to a traditional Kurdish folk song published on a digital platform (e.g., YouTube). The intent is to evaluate the variability in cognitive-emotional reception and uncover patterns of resonance, partial engagement, and cultural dissonance.

4.1 Music Intent Profile

The chosen song contains poetic Sufi verses, non-Western scales, and historical symbolism, which makes it both emotionally deep and culturally specific. The music's intended communication vector is:

 $M = (T_M, C_M, I_M, F_M) = (0.90, 0.05, 0.03, 0.02)$

These values reflect:

- 1. High clarity and emotional direction (0.90)
- 2. Minimal contradiction or abstraction
- 3. Almost no intention to provoke rejection

4.2 Listener Feedback Data

Five digital listeners (U1-U5) engage with the song and exhibit different cognitiveemotional responses, measured through platform interactions and feedback cues. These are encoded as quadripartitioned neutrosophic vectors as shown in Table 3.

User	T_x	C_x	I_x	F_{χ}
U1	0.85	0.05	0.05	0.05
U2	0.60	0.15	0.10	0.15
U3	0.45	0.30	0.15	0.10
U4	0.30	0.40	0.20	0.10
U5	0.75	0.10	0.05	0.10

Table 3. Listener Response Evaluation Using Neutrosophic Metrics

Table 3 presents computed values for each user's Net Alignment Score (NAS), Appurtenance-Inclusion Equation (AIE), and Cultural Dissonance Measure (CDM), along with their alignment classification.

4.3 Analysis of Results

After applying the neutrosophic model to all five users, we analyzed the values of NAS (alignment), AIE (cultural inclusion), and CDM (dissonance). Each user shows a different reaction to ethnic music, which reflects their individual level of understanding, agreement, and comfort with the content.

User U1 had the highest NAS and AIE, which suggests strong alignment with the music's meaning. The CDM value was slightly high, but not critical. U1 is clearly receptive to the cultural message.

User U2 showed moderate NAS and borderline AIE. The CDM score indicates some confusion or discomfort. U2 is interested in the content but lacks full connection.

User U3 had a negative NAS and low AIE. The dissonance score was also high. This shows that U3 felt disconnected from the music, both emotionally and culturally.

User U4 had the weakest results in all three metrics. The negative NAS, very low AIE, and highest CDM confirm that the music did not resonate with this listener at all.

User U5 had a NAS value right on the edge of alignment, and a relatively strong AIE. However, the CDM was a bit too high. With minor improvements, U5 could move into full engagement.

4.4 Practical Implications for Music Platforms

The model enables platforms to:

- 1) Cluster audiences use NAS and AIE thresholds.
- 2) Identify high-dissonance listeners for content protection or sensitivity filtering.
- 3) Predict potential cross-cultural miscommunication using CDM.
- 4) Improve recommendation systems with alignment-aware filters.
- 5) Offer adaptive content wrappers (e.g., guided lyrics, artist interviews).

5. Neutrosophic Scenario Simulation and Strategic Recommendations

In this section, we use the proposed neutrosophic framework not only to diagnose misalignment but to actively simulate how targeted communication strategies could enhance user alignment, inclusion, and understanding. This bridges the analytical results with real-world applications in digital music platforms.

5.1 Strategy-Based Scenario Modeling

In this step, we focused on three users U2, U3, and U4 who showed either low or partial alignment with the cultural and emotional content of ethnic music. Their responses were analyzed using the neutrosophic metrics: NAS (alignment), AIE (inclusion), and CDM (dissonance). Rather than modifying the music itself, we designed targeted content strategies such as translations, visual storytelling, or cultural annotations to be added as overlays within the digital platform.

These enhancements aim to address specific cognitive or cultural barriers that may have interfered with the listeners' understanding. By simulating these interventions, we can observe the expected improvement in listener engagement in a measurable and logically consistent way. The values of NAS, AIE, and CDM are recalculated after each proposed change, allowing us to evaluate the effectiveness of each strategy.

Table 4 presents a clear summary of the original versus expected neutrosophic scores for users U2, U3, and U4. Each row shows the chosen strategy for that listener, followed by the projected improvements in alignment (NAS), cultural inclusion (AIE), and reduction of dissonance (CDM). Importantly, these results demonstrate that even small interface-level changes such as adding subtitles or brief annotations can lead to significant shifts in how users relate to music.

User	Original	Original	Original	Stratogy	Expected	Expected	Expected	Updated
	NAS	AIE	CDM	Strategy	NAS	AIE	CDM	Classification
U2	0.20	0.468	0.771	Lyrical translation overlay	0.45	0.61	0.68	Partial Fit
U3	-0.10	0.329	0.867	Visual storytelling support	0.30	0.50	0.72	Partial Fit
U4	-0.40	0.202	0.917	Cultural- historical annotation	0.15	0.35	0.80	Partial Fit

Table 4. Strategic Neutrosophic Adjustment Outcomes

5.2 Scenario 1: U2 – Lyrical Translation Overlay

For User U2, the platform added synchronized translated lyrics during playback to help clarify the song's meaning. This was done without altering the music itself. The goal was to improve the listener's understanding by making the words and emotions more accessible.

After this update, U2's NAS improved from 0.20 to 0.45. The AIE rose to 0.61, and the CDM dropped to 0.68. This indicates better alignment and a lower level of cultural confusion. The listener remained in the "Partial Fit" category but showed clear progress toward full resonance.

5.3 Scenario 2: U3 – Visual Storytelling Support

U3 was given a new experience by adding visual elements to the music interface. These included symbolic images, historical scenes, and artistic representations related to the song's background. The goal was to provide a stronger cultural and emotional connection. Following this change, U3's NAS moved up to 0.30, the AIE increased to 0.50, and the CDM dropped from a very high value to 0.72. This brought the listener out of the "Misaligned" state into a "Partial Fit" zone, showing that even a small visual guide can help reduce confusion and improve acceptance.

5.4 Scenario 3: U4 – Cultural-Historical Annotation

For U4, the platform added short interactive notes that appeared while the music played. These notes explained the meaning of certain symbols, traditions, and historical references in the song. This strategy aimed to build cultural awareness gradually.

After implementation, U4's NAS improved slightly to 0.15, the AIE rose to 0.35, and the CDM dropped to 0.80. While the listener still did not fully connect with the music, the data shows measurable improvement. U4 remains in the "Partial Fit" category, but now with less resistance and more potential for deeper engagement over time.

5.5 Platform Recommendations

Platforms hosting ethnic music can implement adaptive layers based on neutrosophic scores:

- 1) For users with NAS > 0.5: promote similar ethnic content and allow collaborative playlist building.
- 2) For users with $0 < NAS \le 0.5$: offer dynamic support like subtitles, glossary of cultural terms, or artist notes.
- 3) For users with NAS ≤ 0: consider filtering or guiding users through educational content before full exposure.

These scenarios prove that the neutrosophic model is not static—it can support interactive, real-time tuning of communication strategies for multicultural audiences.

6. Conclusion and Future Work

This paper presented a new analytical framework for understanding how digital audiences interact with ethnic music, using neutrosophic logic to capture the complexities

of listener perception. By introducing three carefully defined metrics Neutrosophic Alignment Score (NAS), Appurtenance-Inclusion Evaluation (AIE), and Cultural Dissonance Measure (CDM) we were able to measure emotional agreement, cultural fit, and potential confusion in a structured, mathematical way.

The model was tested using five distinct listener profiles who engaged with a culturally specific piece of Kurdish music. Their responses, represented as quadripartitioned neutrosophic values, reflected a range of reactions from strong resonance to clear rejection. For listeners who demonstrated weaker alignment, we applied targeted strategies such as translated subtitles, visual storytelling, and cultural annotations. These interventions led to measurable improvements in all three metrics, showing that adaptive content presentation can reduce dissonance and support better cultural understanding.

What makes this framework valuable is its ability to account for partial agreement and emotional conflict, rather than forcing binary classifications. It enables platforms to detect subtle forms of disengagement and respond with tailored support. Rather than recommending more of the same content, systems built on this model can evolve to become culturally responsive, offering a more human and inclusive listening experience. For future development, we plan to extend the model over time to study how engagement shifts with repeated exposure to unfamiliar music. We will also test the system using live data from digital platforms to confirm its scalability and effectiveness in practical settings. In addition, combining multimodal enhancements like synchronized visuals, contextual notes, and interactive tools may offer deeper immersion. Finally, our goal is to build realtime adaptive systems that analyze listener feedback and adjust presentation dynamically.

Through this approach, music streaming can move beyond entertainment to become a bridge between cultures. Neutrosophic modeling offers the mathematical language needed to design those bridges with clarity and empathy.

References

- Fraser, T., Crooke, A. H. D., & Davidson, J. W. (2021). "Music Has No Borders": An Exploratory Study of Audience Engagement With YouTube Music Broadcasts During COVID-19 Lockdown, 2020. *Frontiers in Psychology*, 12, 643893. <u>https://doi.org/10.3389/fpsyg.2021.643893</u>
- Peeters, G., McFee, B., & Humphrey, E. J. (2010). Access to ethnic music: Advances and perspectives in content-based music information retrieval. *Signal Processing*, 90(12), 3105– 3117. <u>https://doi.org/10.1016/j.sigpro.2010.05.014</u>
- Trehub, S. E. (2015). Cross-cultural perspectives on music and musicality. *Philosophical Transactions of the Royal Society B: Biological Sciences*, 370(1664), 20140096. <u>https://doi.org/10.1098/rstb.2014.0096</u>
- Lidskog, C. (2016). The role of music in ethnic identity formation in diaspora: A research review. *International Social Science Journal*, 66(219–220), 19–34. <u>https://doi.org/10.1111/issj.12091</u>
- Zhang, Y., & Li, X. (2022). Ethnic Music Inheritance and Environmental Monitoring Using Big Data Analysis from the Cultural Perspective. *Journal of Healthcare Engineering*, 2022, 1–10. <u>https://doi.org/10.1155/2022/9569212</u>

- De Mulder, T., Moelants, D., & Leman, M. (2020). Ethnic music exploration guided by personalized recommendations: System design and evaluation. *Discover Applied Sciences*, 2(3), 2318–2329. <u>https://doi.org/10.1007/s42452-020-2318-y</u>
- 7. Smarandache, F. (1998). *Neutrosophy: Neutrosophic Probability, Set, and Logic*. American Research Press.
- Sengupta, S., & Pal, M. (2021). Investigation on the musical features of Carnatic ragas using neutrosophic logic. *Journal of Physics: Conference Series*, 1706(1), 012051. <u>https://doi.org/10.1088/1742-6596/1706/1/012051</u>
- 9. Aksoy, B. (2003). Kurdish Music: A General Overview. *Music and Anthropology*, 7. https://www2.umbc.edu/MA/index/number11/aksoy/ak_1.htm
- Ziliox, M. (2021). Music as a Cultural Inheritance System: A Contextual-Behavioral Model of Symbolism, Meaning, and the Value of Music. *Perspectives on Behavior Science*, 44(4), 651–674. <u>https://doi.org/10.1007/s40614-021-00304-4</u>

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