

VERBAL AND NONVERBAL COMMUNICATION CODES

ВЕРБАЛЬНІ ТА НЕВЕРБАЛЬНІ КОДИ КОМУНІКАЦІЇ

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EMOTIVE SPEECH ACTS IN CROSS-CULTURAL COMMUNICATION: A COMPREHENSIVE ANALYSIS AND EXPERIMENTAL STUDY

Мета дослідження – визначити роль емоційних мовленнєвих актів у міжкультурному середовищі оволодіння мовою, виявивши складну взаємодію між універсальними емоційними маркерами та культурно-специфічними моделями вираження. У ході дослідження були застосовані *методи* аналізу даних (акустичний аналіз, аналіз виразу обличчя з використанням системи кодування мімічних дій (FACS), лексичний аналіз, кореляційний і регресійний аналіз).

Завдяки всебічному аналізу акустичних особливостей, міміки і лексичних моделей дослідження демонструє, що вираження емоцій відбувається за двома схемами: універсальні елементи залишаються незмінними в різних мовах, в той час як інші піддаються значній культурній адаптації. Результати показують, що ті, хто вивчають мову, розвивають “емоційний міжмовний контакт”, який синтезує стратегії вираження рідної мови з нормами мови, що вивчається. Ті, хто вивчає іспанську мову, демонстрували більшу виразність обличчя при вираженні щастя, що свідчить про прийняття правил більш відкритого прояву емоцій є **цільовий культурі. Гнів був більш чітко виражений у всіх групах**, які вивчають мову, що свідчить про використання різними емоціями різних каналів вираження. Аналіз основних компонентів та ієрархічна кластеризація виявили окремі профілі вираження емоцій у різних мовних групах, тоді як моделі множинної регресії виявили прогностичні взаємозв'язки між рівнем володіння мовою, культурним досвідом та емоційною адаптацією.

Наші результати підтверджують теоретичну модель, яка об'єднує універсалістські та релятивістські погляди на вираження емоцій, припускаючи, що ті, хто вивчають мову, орієнтуються в динамічному просторі між цими полюсами. Дослідження підтверджує, що деякі аспекти вираження емоцій, такі як підвищена інтенсивність голосу при гніві та зниження темпу мови при смутку, залишаються відносно незмінними в різних мовних групах, що підтверджує гіпотезу універсальності. Однак інші аспекти — зокрема, виразність обличчя, що виражає щастя, і вибір лексики, що відображає емоційні стани, — демонструють значну адаптацію до норм мови перекладу, що підтверджує точку зору культурної відносності.

Наші дані показують, що ті, хто вивчає мову, розвивають те, що можна назвати “емоційною міжмовою” — динамічною системою вираження емоцій, яка включає елементи як з рідного емоційного репертуару, так і з культурних норм мови, що вивчається. Ця емоційна взаємодія розвивається в міру підвищення рівня володіння мовою і залучення до культури, але процес адаптації варіюється залежно від різних каналів емоційного вираження і різних емоцій. Висновок про те, що знайомство з культурою опосередковує взаємозв'язок між рівнем володіння мовою і емоційною виразністю, дозволяє припустити, що емоційна адаптація при вивченні мови є не лише функцією лінгвістичних знань, але вимагає більш глибокого вивчення культури і залученості в неї.

Ключові слова: емоційна мова, міжкультурна комунікація, емоційний тон, вплив культурних відмінностей, інтенсивність голосу, лексичний вибір, міжкультурна емоційна адаптація.

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Introduction

Language serves not only as a tool for transmitting information but also as a powerful medium for expressing emotions. Emotive speech acts, which convey affective states through linguistic and extralinguistic means, play a crucial role in interpersonal communication, helping individuals express gratitude, anger, sadness, and other emotional states. Understanding how emotions are conveyed and perceived in different linguistic and cultural contexts is essential for improving language acquisition, intercultural communication, and the development of artificial intelligence systems capable of recognizing and responding to human emotions.

The study of emotive speech acts is particularly significant in today's globalized world, where intercultural communication plays a pivotal role in both personal and professional contexts. Despite growing research on emotive speech acts, much of the existing literature has focused on native speakers, leaving a significant gap in understanding how non-native speakers perceive and produce emotional expressions in a second or foreign language.

The extent to which these features influence emotional perception among non-native speakers remains underexplored. Cultural background and linguistic exposure may shape how learners interpret emotional tones in a second language. This research seeks to address these gaps by analyzing how Azerbaijani students learning American English, British English, and Spanish perceive and respond to emotive speech acts, utilizing advanced quantitative and qualitative methodologies including spectral analysis, facial action coding (FACS), corpus linguistics techniques, principal component analysis, hierarchical clustering, and multiple regression modeling.

The findings have implications for linguistics, psycholinguistics, second language acquisition, and artificial intelligence. By understanding how non-native speakers process emotional speech, educators can develop more effective language teaching strategies, while AI developers can enhance emotion recognition systems to improve human-computer interactions.

Theoretical Review

The study of emotive speech acts is rooted in several key theoretical frameworks spanning multiple disciplines, including linguistics, psychology, sociology, and philosophy.

Speech Act Theory [Austin, 1962; Searle, 1969] provides a foundational framework for understanding emotive speech acts. Austin introduced the concept of performative utterances, which are statements that not only describe reality but also perform actions. Emotive speech acts, such as expressions of gratitude or anger, fall under this category of performative utterances.

Searle expanded on Austin's work by classifying speech acts into distinct categories, including expressives, which are employed to convey the speaker's emotional state. Examples of expressives include "I'm sorry," "I'm delighted," and "I'm disappointed" [Austin, 1962]. Searle emphasized that emotive speech acts are deeply tied to the speaker's intentions and the context in which they are uttered.

Grice's Theory of Implicature [Grice, 1975] provides another valuable lens for understanding emotive speech acts. Grice proposed that speakers often imply meanings beyond the literal content of their words, a phenomenon known as implicature. This is particularly relevant for emotive speech acts, such as irony or sarcasm, which rely on implicature to convey emotions indirectly.

Cognitive linguistics offers [Lakoff, 1987; Langacker, 1987] a unique perspective on emotive speech acts by emphasizing the role of mental representations in shaping language. Lakoff introduced the concept of conceptual metaphors, which allow abstract emotions to be expressed through concrete images. For example, the metaphor "anger is fire" is reflected in expressions like "She was burning with rage" [Lakoff, 1987].

The sociocultural approach [Bakhtin, 1981; Goffman, 1959; Fairclough, 1992] highlights the role of social context in shaping emotive speech acts. Bakhtin viewed language as a dialogical process, where emotions are expressed through interaction with others [Bakhtin, 1981]. Goffman introduced the concept of "face," which refers to the public self-image that individuals strive to maintain in social interactions. According to Goffman, emotive speech acts are often employed to manage face and navigate social relationships [Goffman, 1959]. Brown and Levinson [1987] emphasize the importance of considering the context and audience in understanding the meaning of utterances.

Theory of Cross-Cultural Emotion Recognition [Ekman, 1972; Sauter, Eisner, Ekman, Scott, 2010; Elfenbein, Ambady, 2002; Laukka, Elfenbein, 2020; Pell, Monetta, Paulmann, Kotz, 2009]. The universality of emotional expression has been a topic of significant debate. Ekman [1972] proposed that certain basic emotions are universally recognized and expressed through consistent facial and vocal cues. However, research by Elfenbein and Ambady [2002] has shown that cultural differences can influence the interpretation of emotional expressions, highlighting the importance of considering cultural context when studying emotive speech acts.

This research integrates Speech Act Theory's emphasis on speaker intention with advancements in cross-cultural emotion research. While traditional approaches like Ekman's universality theory provide a foundation for understanding basic emotional expressions, our framework extends this by examining how language acquisition contexts create unique spaces where universal emotional cues interact with target language cultural norms. We propose that language learners develop an *emotional interlanguage* — a dynamic system of emotional expression that reflects elements of both their native emotional patterns and those of their target language community.

Addressing the Research Questions

The experimental findings provide valuable insights into the three key research questions that guided this study:

1. *What linguistic features contribute to the conveyance of emotional tone in everyday conversation?*

The results of our acoustic and linguistic analyses reveal several key features that play significant roles in conveying emotional tone:

Vocal Intensity: Our findings demonstrate that vocal intensity serves as a powerful indicator of emotional states, particularly for anger. The significantly higher intensity levels observed for anger ($M = 85$ dB, $SD = 5$ dB) compared to happiness ($M = 75$ dB, $SD = 7$ dB) and sadness ($M = 65$ dB, $SD = 6$ dB) suggest that speakers modulate their vocal energy to communicate emotional intensity. This aligns with Juslin and Laukka's [2003] research highlighting the universal role of vocal intensity in emotional expression.

Pitch Variations: The data reveal that pitch variations function as crucial markers for distinguishing between positive and negative emotions. The significantly higher mean pitch observed for happiness ($M = 220$ Hz, $SD = 18$ Hz), compared to sadness ($M = 200$ Hz, $SD = 18$ Hz), supports the established theory that higher pitch correlates with positive emotional states [Ladd, 1996]. This pitch-emotion relationship appears consistent across all three language learning groups, suggesting its potential universality as an emotional cue.

Lexical Choices: The keyword analysis revealed significant patterns in the linguistic expression of emotions. Particularly noteworthy was the finding that participants used significantly more anger-related keywords (e.g., "furious," "mad") in anger scenarios ($M = 3.2$, $SD = 1.1$) compared to happiness scenarios ($M = 0.8$, $SD = 0.5$). This indicates that certain emotions, particularly anger, are more explicitly verbalized through specific lexical choices. Interestingly, sadness showed less distinct lexical marking, suggesting that different emotions may rely on different channels of expression.

Facial Expressions: The facial expression analysis demonstrated that specific Action Units (AUs), such as AU12 (lip corner puller) for happiness, serve as important visual cues for emotional communication. The activation patterns of these AUs varied across language groups, highlighting the interaction between universal emotional expressions and cultural influences.

These findings collectively demonstrate that emotional tone in conversation is conveyed through a complex interplay of prosodic features (intensity and pitch), lexical choices, and facial expressions, with different emotions utilizing these channels to varying degrees.

2. *How do cultural differences influence the use of emotive speech acts across languages?*

Our comparison of Azerbaijani students learning American English, British English, and Spanish revealed several important cultural differences in emotional expression:

Language-Specific Facial Expressions: The analysis of facial expressions showed that Azerbaijani students learning Spanish exhibited significantly greater activation of AU12 (lip corner puller) for happiness ($M = 0.40$, $SD = 09$) compared to those learning American English (M

= 0.45, SD = 10). This finding suggests that students may adopt different patterns of emotional expression depending on the target language they are learning, supporting the cultural adaptability hypothesis proposed by Fernández-Dols [See: Fernández-Dols, Carrera, Crivelli, 2012; Fernández-Dols, Russell, 2017].

Variation in Emotional Prosody: While all language learning groups showed similar patterns in using pitch and intensity to mark emotions, subtle differences were observed in how these features were combined. For instance, students learning Spanish showed a tendency to use more dynamic pitch variations within utterances when expressing happiness compared to those learning American and British English, although this difference did not reach statistical significance in our sample. This suggests that language learning influences prosodic patterns of emotional expression.

Target Language Influence on Emotional Expression: The keyword analysis revealed differences in how explicitly emotions were verbalized across language learning groups. Students learning Spanish used more emotion-specific terms when describing both positive and negative emotional states than those learning American and British English. This indicates that the linguistic and cultural features of the target language may influence how learners express emotions verbally.

Cross-Cultural Emotional Adaptation: Our findings suggest that Azerbaijani students adapt their emotional expression patterns when communicating in different target languages, displaying evidence of what could be called “emotional code-switching.” This phenomenon indicates that language learners develop awareness of the cultural norms governing emotional expression in their target languages.

These findings collectively demonstrate that while certain aspects of emotional expression appear universal, the target language and its associated cultural context significantly influence how language learners express and interpret emotions – highlighting the importance of cultural awareness in second language acquisition.

3. *What is the relationship between cognitive processes and emotive speech acts?*

The experimental results provide insights into the cognitive dimensions of emotive speech acts:

Conceptual Metaphors in Emotional Expression: The linguistic analysis revealed patterns consistent with Lakoff’s [1987] theory of conceptual metaphors. For instance, participants across all language learning groups used spatial metaphors when describing emotional states (e.g., “feeling down” for sadness, “rising anger”), suggesting that abstract emotional concepts are cognitively structured through concrete spatial experiences. However, the specific metaphorical mappings showed some variation across target languages, supporting the cognitive-linguistic integration model.

Cognitive Processing of Emotional Cues: The integration of multiple channels (vocal, verbal, facial) in emotional communication suggests a complex cognitive processing system that weights and combines various cues. Our findings indicate that when linguistic and paralinguistic cues conflicted, Azerbaijani students relied more heavily on prosodic features than lexical content for emotional judgments, suggesting a cognitive prioritization of certain emotional channels.

Emotion Recognition Strategies: The data suggest that participants employed different cognitive strategies when interpreting emotions across languages. When processing emotions in their native Azerbaijani language (in control tasks), participants showed faster recognition times and reported relying on a holistic impression. In contrast, when processing emotions in their target languages (American English, British English, or Spanish), participants reported more analytical approaches, focusing on specific cues such as keywords or pronounced prosodic features.

First Language to Target Language Transfer: The experiment revealed evidence of L1-to-L2 transfer in emotional processing, as participants sometimes applied emotion recognition strategies from their native Azerbaijani language when interpreting emotions in their target languages. This suggests that cognitive frameworks for emotional processing may be partially transferred from first to second languages while also adapting to new linguistic contexts.

These findings support a dynamic relationship between cognitive processes and emotive speech acts, where universal cognitive mechanisms interact with language-specific

conceptualizations, and cultural frameworks to shape how emotions are expressed and understood in second language acquisition and cross-cultural communication.

The experimental results thus provide substantive answers to our three research questions, demonstrating that emotive speech acts involve a complex interplay of universal linguistic features, target language cultural influences, and cognitive processes in language learning contexts. These insights have significant implications for language pedagogy, intercultural communication, and the development of emotion recognition technologies that are sensitive to both universal patterns and cultural variations.

By addressing these questions through a mixed-methods approach combining acoustic analysis, facial expression coding, and keyword analysis among Azerbaijani students learning three different languages, this study aims to provide a comprehensive understanding of emotive speech acts in language learning contexts, bridging the gap between theoretical frameworks and practical applications in second language acquisition.

Research Methodology

Research Hypothesis and Theoretical Foundation

We hypothesize that Azerbaijani students will exhibit distinctive emotional expression patterns across their target languages, reflecting both universal features and culturally modulated adaptations. Specifically, while certain prosodic features such as increased vocal intensity for anger are expected to remain consistent, facial expressiveness and lexical usage will vary according to target language norms. These differences will correlate with cultural familiarity and immersion.

This hypothesis is grounded in two complementary theoretical frameworks:

1. The Universality-Relativity Paradigm [Ekman, 1972; Elfenbein, Ambady, 2002]: This framework suggests that while certain basic emotions are universally recognized through consistent facial and vocal cues, cultural differences can significantly influence how these emotions are expressed and interpreted.

2. The Cognitive-Linguistic Integration Model [Lakoff, 1987; Langacker, 1987]: This model posits that emotional concepts are structured through conceptual metaphors and cognitive schemas that vary across languages and cultures.

Participants

One hundred participants were recruited from the Azerbaijan University of Languages in Baku, Azerbaijan. These participants were third-year students studying American English ($n = 34$), British English ($n = 33$), and Spanish ($n = 33$). Participants ranged in age from 19 to 22 years ($M = 20.5$, $SD = 1.2$), with a gender distribution of 80% female and 20% male. Inclusion criteria were fluency in their target language (minimum B2 level according to the Common European Framework of Reference for Languages) and no reported history of speech or hearing disorders. All participants provided informed consent before participating in the study.

Experimental Design and Materials

We employed a mixed-methods design incorporating both between-subjects (language group: American English, British English, Spanish) and within-subjects (emotion type: gratitude, anger, sadness) variables. The experimental stimuli consisted of three emotionally evocative scenarios designed to elicit authentic emotional responses rather than acted performances.

Emotion Elicitation Scenarios

Gratitude Scenario: "Imagine a close friend has helped you move to a new apartment during a particularly difficult time in your life. This assistance required them to rearrange their schedule and travel a considerable distance. How would you express your gratitude to them?"

Anger Scenario: "Imagine you discover that someone in your university class has falsely accused you of stealing their research notes and has reported this to your professor, potentially affecting your academic standing. How would you respond to this accusation when confronting the accuser?"

Sadness Scenario: "Imagine you have just received a phone call informing you that a loved one has passed away unexpectedly. How would you express your grief when sharing this news with a close friend?"

The scenario presentation order was counterbalanced using a Latin square design to control for order effects. Participants were given two minutes to formulate their responses for each scenario, which were then recorded for analysis.

Each scenario was pre-tested with 45 participants for emotional specificity and cultural neutrality. All selected scenarios scored above 4.2 (out of 5) in emotional clarity and above 4.0 in cross-cultural applicability.

Acoustic features: Mean F0 (Hz), intensity range (dB), speech rate (syllables/sec), pitch variation.

Facial expression metrics: AU12 (happiness), AU4 (anger), AU15 (sadness) intensity scores.

Lexical cues: Frequency of emotion-related keywords (e.g., "furious," "crushed"), metaphor use, presence of intensifiers.

Data Collection Procedures

Participants were tested individually in a soundproof laboratory. After reading each scenario, they were instructed to respond naturally, as if experiencing the situation in real life. Audio recordings were made using a Shure SM58 microphone connected to a Zoom H6 digital recorder (sampling rate: 44.1 kHz, 16-bit depth). Simultaneously, video recordings were captured using a Sony FDR-AX53 camera (1080p resolution, 30 fps) positioned 1.5 meters from the participant to capture facial expressions.

Data Analysis Methods

Acoustic Analysis

We conducted spectral and temporal analysis using Praat software (version 6.1.42) to examine the following variables:

Fundamental Frequency (F0) Analysis:

- a) Mean pitch (Hz)
- b) Pitch range (calculated as the difference between maximum and minimum F0)
- c) Pitch variation (coefficient of variation = standard deviation of F0 divided by mean F0)
- d) F0 contour shape (rising, falling, or flat trajectories)

Intensity Analysis:

- a) Mean intensity (dB)
- b) Intensity range (maximum minus minimum intensity)
- c) Intensity fluctuation (standard deviation of intensity)
- d) Energy distribution across frequency bands (using Fast Fourier Transform)

Temporal Features:

- a) Speech rate (syllables per second)
- b) Articulation rate (excluding pauses)
- c) Pause duration (mean length of silent intervals > 200ms)
- d) Pause frequency (number of pauses per utterance)
- e) Rhythmic patterns (using normalized Pairwise Variability Index)

Voice Quality Measures:

- a) Jitter (cycle-to-cycle variation in fundamental frequency)
- b) Shimmer (cycle-to-cycle variation in amplitude)
- c) Harmonics-to-noise ratio (measure of voice clarity)
- d) Spectral tilt (distribution of energy across the spectrum)

Statistical analyses of acoustic data included:

Multivariate Analysis of Variance (MANOVA): To examine differences in acoustic parameters across emotions and language groups.

Univariate ANOVAs: Follow-up tests for significant MANOVA effects.

Tukey's HSD post-hoc tests: For pairwise comparisons between specific emotions and language groups.

Principal Component Analysis (PCA): To identify clusters of correlated acoustic features that explain variance in emotional expressions.

Linear Discriminant Analysis (LDA): To determine which acoustic features best discriminate between emotional states.

Effect sizes were calculated using partial eta-squared (η^2) for ANOVA results and Cohen's d for pairwise comparisons.

Facial Expression Analysis

Facial expressions were analyzed using the Facial Action Coding System (FACS) developed by Ekman and Friesen (1978). Two certified FACS coders, blind to the experimental conditions, coded the following action units (AUs):

Happiness:

- AU6 (cheek raiser)
- AU12 (lip corner puller)
- AU25 (lips part)
- AU26 (jaw drop)

Anger:

- AU4 (brow lowerer)
- AU5 (upper lid raiser)
- AU7 (lid tightener)
- AU23 (lip tightener)
- AU24 (lip presser)

Sadness:

- AU1 (inner brow raiser)
- AU15 (lip corner depressor)
- AU17 (chin raiser)
- AU43 (eyes closed)

The intensity of each AU was rated on a 5-point scale (0 = absent, 5 = maximum intensity).

For each facial recording, coders analyzed:

Temporal dynamics: Onset, apex, and offset times for each AU

Co-occurrence patterns: Frequency of specific AU combinations

Asymmetry indices: Differences between left and right facial activation

Cultural display variations: Culturally specific expressions not captured by standard FACS

Statistical analyses of facial expression data included:

Inter-coder reliability: Assessed using Cohen's kappa ($\kappa = .85$)

Two-way ANOVA: To examine the effects of language group and emotion type on facial expression patterns

Hierarchical Cluster Analysis: To identify patterns of AU co-occurrence across emotions and language groups

Multidimensional Scaling (MDS): To visualize similarities and differences in facial expression patterns across languages

Lexical Analysis

Transcribed responses were analyzed using corpus linguistics techniques implemented in NVivo software (version 14):

Keyword Extraction:

- a) Frequency analysis of emotion-specific terms
- b) Keyness calculations using log-likelihood tests
- c) Collocation analysis (words commonly appearing near emotion terms)
- d) Concordance analysis (contextual examination of key terms)

Lexical Density and Diversity:

- a) Type-token ratio (TTR) (number of unique words divided by total words)
- b) Brunet's index ($W = N^V - 0.165$, where N = text length and V = vocabulary size)
- c) Moving-average type-token ratio (MATTR) to control for text length effects
- d) Measure of textual lexical diversity (MTLD) to assess lexical variation

Sentiment Analysis:

a) LIWC (Linguistic Inquiry and Word Count) software for quantification of positive/negative emotion words

- b) Valence analysis using SentiWordNet
- c) Emotional intensity scoring based on lexical semantic resources
- d) Domain-specific emotional lexicon development and application

Discourse Structure Analysis:

- a) Transitivity analysis (types of processes used to express emotions)
- b) Modality patterns (expressions of certainty/uncertainty)
- c) Discourse markers related to emotional states
- d) Metaphor identification and analysis (using the Metaphor Identification Procedure)

Statistical analyses of lexical data included:

Inter-coder reliability: Assessed using Krippendorff’s alpha ($\alpha = .82$) for manual coding aspects

MANOVA: To compare lexical features across language groups

Bonferroni-corrected follow-up tests: For multiple comparisons

Correspondence Analysis: To visualize relationships between linguistic features and emotional expressions

Semantic Network Analysis: To map conceptual relationships between emotion terms across languages

Correlational and Regression Analysis

To examine relationships between linguistic and non-linguistic features, we conducted:

Hierarchical Multiple Regression Analysis:

- a) Dependent variables: Measures of emotional expressiveness
- b) Independent variables: Language proficiency (entered at step 1), years of study (step 2), gender (step 3), cultural familiarity measures (step 4)
- c) Interaction terms: To test moderating effects of language group on relationships between predictors and outcomes

Pearson’s Correlation Analysis:

- a) Bivariate correlations between acoustic parameters and facial expressions
- b) Partial correlations controlling for language proficiency
- c) Cross-correlations to examine time-lagged relationships between modalities

Machine Learning Models:

- a) Random Forest classification to identify the most important predictors of emotional expression
- b) Support Vector Machines (SVM) to model the relationship between linguistic features and emotional categories
- c) K-means clustering to identify patterns of multimodal emotional expression across language groups

Structural Equation Modeling (SEM):

- a) Path analysis to test theoretical relationships between language exposure, cultural awareness, and emotional competence
- b) Latent variable analysis to examine the underlying structure of emotional expression abilities

These comprehensive analytical approaches allowed us to examine the complex relationships between linguistic, cultural, and emotional variables, providing a multifaceted understanding of how emotions are expressed across language groups.

Table 1

Statistical Techniques Summary

Modality	Analysis Method	Purpose
Acoustic	MANOVA, PCA, LDA	Group comparisons, feature reduction, emotion classification
Facial	ANOVA, Cluster Analysis, MDS, Cohen’s kappa	Identify emotion patterns, cultural variation, inter-coder agreement
Lexical	MANOVA, Correspondence Analysis, PCA, Correlation Analysis	Lexical feature comparison, marker detection, emotion correlation
Correlational	Pearson, Partial Correlation	Intermodal relationships
Predictive	Multiple Regression, Random Forest, SVM	Identify predictors of emotional adaptation
Integration	Canonical Correlation Analysis	Multimodal coordination assessment

Results

Acoustic Analysis

The analysis of acoustic features revealed significant patterns in the expression of emotions through vocal characteristics. A multivariate analysis of variance (MANOVA) with emotion type and language group as independent variables showed a significant main effect of emotion on acoustic parameters, Wilks' $\lambda = 42$, $F(8, 184) = 13.26$, $p < .001$, partial $\eta^2 = .37$.

Follow-up univariate ANOVAs revealed a statistically significant effect of emotion on vocal intensity, $F(2, 94) = 25.32$, $p < .001$, partial $\eta^2 = .35$. Post-hoc comparisons using Tukey's HSD test showed that expressions of anger were characterized by significantly higher vocal intensity compared to expressions of happiness ($p < .001$, $d = 1.28$) and sadness ($p < .001$, $d = 1.65$). This demonstrates that participants consistently modulated their vocal energy to communicate anger, regardless of their target language, suggesting that intensity may serve as a universal marker for this emotion.

Analysis of pitch variations revealed that happiness was associated with higher mean fundamental frequency (F0) compared to sadness, $F(2, 94) = 18.72$, $p < .001$, partial $\eta^2 = .28$. The mean F0 for happiness expressions ($M = 242.3$ Hz, $SD = 32.6$) was significantly higher than for sadness ($M = 187.4$ Hz, $SD = 24.8$), $p < .001$, $d = 1.89$, and moderately higher than for anger ($M = 223.1$ Hz, $SD = 29.7$), $p = .038$, $d = 0.62$. This finding indicates that participants employed pitch modulation as a key strategy for differentiating positive from negative emotions, consistent with previous research on emotional prosody.

Speech rate analysis showed that sadness was expressed with significantly slower articulation rates ($M = 3.2$ syllables/second, $SD = 0.5$) compared to both happiness ($M = 5.1$ syllables/second, $SD = 0.7$), $p < .001$, $d = 3.08$, and anger ($M = 4.8$ syllables/second, $SD = 0.8$), $F(2, 94) = 14.89$, $p < .001$, partial $\eta^2 = .24$. This demonstrates that temporal features contribute meaningfully to emotional differentiation in speech.

Principal Component Analysis (PCA) of acoustic features revealed three components accounting for 78.3% of the variance in emotional expressions. The first component (37.2% of variance) was primarily associated with intensity and energy distribution, the second component (26.5%) with pitch variations, and the third component (14.6%) with temporal features. The PCA revealed that American English and British English learners clustered together in their acoustic profiles, while Spanish learners formed a separate cluster, particularly for expressions of happiness.

Linear Discriminant Analysis (LDA) successfully classified emotional expressions based on acoustic features with an overall classification accuracy of 76.4% (compared to a chance level of 33.3%). Classification accuracy was highest for anger (84.2%), followed by sadness (78.9%) and happiness (66.1%), indicating that anger had the most distinctive acoustic profile.

Facial Expression Analysis

The analysis of facial expressions using FACS revealed both universal patterns and cultural variations in how emotions were expressed across language groups. A two-way ANOVA examining the effects of language group and emotion type on AU12 activation (lip corner puller, associated with happiness) showed a significant main effect of language group, $F(2, 94) = 4.25$, $p = .016$, partial $\eta^2 = .07$.

Post-hoc analyses indicated that Azerbaijani students learning Spanish exhibited significantly greater AU12 activation when expressing happiness ($M = 3.8$, $SD = 0.9$) compared to those learning American English ($M = 3.1$, $SD = 1.2$), $p = .042$, $d = 0.62$. This finding suggests that students may adapt their facial expression patterns to match the cultural norms of their target language, as Spanish culture typically encourages more overt expressions of positive emotions.

The analysis of anger-related action units (AU4, AU5, AU7, AU23) showed consistent activation patterns across all language groups, suggesting that facial expressions of anger may be more universally encoded. However, the intensity of these expressions varied, with Spanish learners showing somewhat stronger activations ($M = 3.6$, $SD = 0.8$) compared to American English ($M = 3.2$, $SD = 0.9$) and British English learners ($M = 3.3$, $SD = 0.7$), though these differences did not reach statistical significance ($p = .09$).

Hierarchical cluster analysis of facial action units revealed three distinct clusters corresponding to the three target emotions, with a cophenetic correlation coefficient of 0.82, indicating a strong fit between the original distances and the clustered distances. However, the analysis also revealed subclusters within each emotion category that corresponded to language groups, particularly for happiness expressions. This suggests that, while the basic configuration of facial expressions for each emotion is consistent across language groups, cultural influences modulate the intensity and specific patterns of these expressions.

Multidimensional scaling (MDS) analysis of facial expression patterns produced a two-dimensional solution with a stress value of 0.14, indicating a reasonably good fit. The first dimension separated positive from negative emotions, while the second dimension appeared to reflect intensity of expression. Spanish learners' expressions of happiness were positioned farther along the intensity dimension compared to American and British English learners, providing further evidence for cultural adaptation in facial expressiveness.

Temporal analysis of facial expressions revealed that Spanish learners maintained happiness expressions for longer durations ($M = 4.2$ seconds, $SD = 1.1$) compared to American English learners ($M = 3.1$ seconds, $SD = 0.9$), $p < .01$, $d = 1.08$, and British English learners ($M = 3.3$ seconds, $SD = 0.8$), $p < .01$, $d = 0.91$. This suggests that duration of expression, in addition to intensity, may be a culturally adapted aspect of emotional expression.

Lexical Analysis

Lexical analysis revealed significant patterns in how emotions were verbalized across different scenarios and language groups. Paired samples t-tests comparing the use of emotion-specific vocabulary across scenarios indicated that participants employed significantly more anger-related keywords in anger scenarios ($M = 7.3$ terms per response, $SD = 2.4$), compared to happiness scenarios ($M = 3.1$ terms per response, $SD = 1.2$), $t(118) = 10.25$, $p < .001$, $d = 2.50$.

LIWC analysis showed that anger was more explicitly verbalized through specific lexical choices (e.g., "furious," "mad," "outraged") than sadness, which was often expressed through more general negative emotion terms. The proportion of explicit emotion terms (as a percentage of total words) was significantly higher for anger expressions ($M = 8.2\%$, $SD = 2.1\%$) than sadness expressions ($M = 5.6\%$, $SD = 1.8\%$), $p < .001$, $d = 1.42$. This suggests that different emotions elicit distinct patterns of lexical selection, with anger prompting more precise emotional vocabulary.

Cross-linguistic analysis revealed that Spanish learners used a greater variety of emotion terms (higher type-token ratio, $M = 0.68$, $SD = 0.07$) when expressing both positive and negative emotions compared to learners of American English ($M = 0.61$, $SD = 0.08$), $p = .003$, $d = 0.74$, and British English ($M = 0.63$, $SD = 0.07$), $p = .012$, $d = 0.65$. This finding suggests that the linguistic and cultural features of the target language may influence how learners express emotions verbally.

Collocation analysis revealed distinctive patterns in the words that co-occurred with emotion terms across language groups. Spanish learners more frequently combined emotion terms with intensifiers, (e.g., "extremely happy," "deeply sad") compared to English learners. American English learners showed a greater tendency to use metaphorical expressions for emotions (e.g., "over the moon," "crushed"), while British English learners more frequently employed understatement, (e.g., "a bit upset," "rather disappointed").

The Measure of Textual Lexical Diversity (MTLD) analysis, which controls for text length effects, confirmed that Spanish learners exhibited greater lexical diversity in emotional expression ($M = 82.4$, $SD = 12.1$) compared to American English learners ($M = 73.6$, $SD = 11.8$), $p = .007$, $d = 0.71$, and British English learners ($M = 75.2$, $SD = 10.9$), $p = .017$, $d = 0.64$. This pattern was consistent across all three emotions, suggesting a general effect of language on lexical diversity rather than an emotion-specific effect.

Semantic network analysis of emotion terms revealed distinct conceptual structures across language groups. Spanish learners' emotion networks showed greater interconnectedness between positive and negative emotion concepts, while English learners maintained clearer boundaries between positive and negative emotion lexicons. This suggests that language learning may influence not only the vocabulary of emotional expression but also the underlying conceptual organization of emotions.

Correlational and Regression Analysis

Hierarchical multiple regression analyses revealed that years of study in the target language significantly predicted participants' use of emotion-specific vocabulary, $\beta = .34$, $p = .003$, and pitch variation, $\beta = .29$, $p = .008$, when controlling for gender and baseline language proficiency. This suggests that longer exposure to the target language enhances learners' ability to express emotions through both lexical and prosodic channels.

The regression model for facial expressiveness revealed that cultural familiarity, (measured by time spent in target language countries and engagement with target culture media), was a significant predictor of AU12 activation during happiness expressions, $\beta = .41$, $p < .001$, beyond the effects of language proficiency ($\beta = .23$, $p = .042$). This finding supports the hypothesis that cultural exposure plays a crucial role in the adaptation of emotional expression patterns.

A moderation analysis revealed a significant interaction between language group and cultural familiarity in predicting facial expressiveness, $F(2, 94) = 5.67$, $p = .004$, partial $\eta^2 = .11$. The relationship between cultural familiarity and facial expressiveness was stronger for Spanish learners ($r = .56$, $p < .001$) compared to American English learners ($r = .31$, $p = .023$) and British English learners ($r = .28$, $p = .034$), suggesting that Spanish culture may place greater emphasis on overt emotional expression.

Pearson correlation analysis showed significant associations between facial expressions and acoustic features, with AU12 activation positively correlating with pitch variation ($r = .38$, $p < .001$) during happiness expressions. This finding suggests an integrated multimodal approach to emotional expression, where facial and vocal cues work in concert to communicate affective states.

Path analysis using Structural Equation Modeling (SEM) revealed a complex relationship between language proficiency, cultural familiarity, and emotional expressiveness. Cultural familiarity mediated the relationship between language proficiency and emotional expressiveness, particularly for facial expressions (indirect effect = $.18$, 95% CI $[.08, .29]$). This suggests that language proficiency enables cultural learning, which in turn facilitates the adaptation of emotional expression patterns.

Random Forest classification identified cultural familiarity, years of study, and frequency of target language use as the most important predictors of emotional expression patterns, collectively accounting for 68.7% of the variance in classification accuracy. Language proficiency, while important, was a less powerful predictor than measures of cultural engagement, suggesting that emotional adaptation may depend more on cultural learning than on linguistic competence per se.

K-means clustering of multimodal emotional expression patterns (combining acoustic, facial, and lexical features) revealed three distinct clusters that corresponded closely to the three language groups (adjusted Rand index = $.76$), with the clearest separation for happiness expressions. This provides further evidence that emotional expression patterns may be shaped by target language and culture, particularly for positive emotions.

Discussion

Interpretation of Findings

The findings of this study provide strong evidence of an interaction between universal emotional markers and culture-specific expression patterns in language learning contexts. The results highlight the importance of vocal intensity, pitch variations, and facial expressions in conveying emotions, while also revealing notable cultural adaptations in emotional expression.

The strong relationship between vocal intensity and anger expression across all language groups supports the universality hypothesis of basic emotional expressions proposed by Ekman [1972]. The consistent acoustic profile of anger, characterized by increased vocal intensity and moderately elevated pitch, suggests that certain aspects of emotional expression may be biologically determined and relatively resistant to cultural influence.

However, the cultural variations observed in facial expressions, particularly for happiness, align with the cultural relativity perspective advocated by Elfenbein and Ambady [2002]. The finding that Spanish learners exhibited greater facial expressiveness for happiness compared

to American English learners suggests that students may internalize the display rules of their target language culture. Spanish culture generally encourages more overt emotional expression compared to Anglo-American cultures, which tend to favor more restrained emotional displays. This cultural adaptation in facial expression supports the emotional acculturation hypothesis proposed by Matsumoto, Yoo, and Fontaine [2008], which suggests that individuals adjust their emotional expression patterns to match the norms of their host culture.

The observation that anger was more explicitly verbalized across all language groups suggests that certain emotions may be more readily expressed through language than others. This finding aligns with the cognitive linguistics perspective, which suggests that emotions are differentially accessible to conscious awareness and verbal expression based on their cognitive salience and cultural importance. The greater lexical diversity observed in Spanish learners' emotional expressions suggests that the linguistic resources available in a language may shape how emotions are conceptualized and expressed.

The correlational and regression analyses reveal a complex interplay between language proficiency, cultural familiarity, and emotional expression. The finding that cultural familiarity mediates the relationship between language proficiency and emotional expressiveness suggests a developmental trajectory in which language learning enables cultural learning, which in turn facilitates emotional adaptation. This supports a model of emotional acculturation in which linguistic and cultural competence develop in parallel and mutually reinforce each other.

The multimodal nature of emotional expression is evident in the correlations between facial expressions and acoustic features. The positive correlation between AU12 activation and pitch variation during happiness expressions suggests that different channels of emotional expression (facial, vocal, verbal) are coordinated to convey a coherent emotional message. This coordination may become more culture-specific as language learners internalize the display rules of their target culture.

Theoretical Implications

The findings contribute to several theoretical frameworks in linguistics, psychology, and communication studies. First, they support the integration of universalist and relativist perspectives on emotion, suggesting that while certain aspects of emotional expression (such as increased vocal intensity for anger) may be universal [Scherer, 2003], others (such as facial expressiveness for happiness) are shaped by cultural context.

Second, the results advance our understanding of Second Language Acquisition (SLA) by highlighting the development of what might be called “emotional interlanguage” — a dynamic system of emotional expression that incorporates elements from both the native and target language cultures. This concept extends traditional interlanguage theory beyond grammatical and lexical domains to encompass pragmatic and emotional competence.

Third, the findings support the embodied cognition perspective, which suggests that emotional understanding is grounded in bodily experience. The consistent correlation between facial expressions and vocal features across language groups indicates that emotional expression involves coordinated multimodal systems that may share underlying cognitive and neural mechanisms.

Fourth, the results contribute to the cultural neuroscience of emotion by suggesting that language learning may reshape neural circuits involved in emotional processing. The cultural adaptation observed in facial expressions and lexical choices suggests that emotional expression is not only a matter of learning display rules but may involve deeper changes in how emotions are experienced and processed.

Finally, the findings support a dynamic systems perspective on emotional development in second language learning. The complex interactions between language proficiency, cultural familiarity, and emotional expression suggest that emotional acculturation is not a linear process but involves multiple feedback loops and developmental trajectories.

Practical Applications

The findings have important practical implications for language teaching, intercultural communication, and artificial intelligence development.

For language educators, the results suggest that effective language teaching should include explicit instruction in the cultural norms governing emotional expression. This might involve awareness-raising activities that highlight differences in emotional display rules across cultures, as well as practice in recognizing and producing culturally appropriate emotional expressions. The finding that cultural familiarity predicts emotional adaptability suggests that immersive cultural experiences and authentic language exposure may be particularly valuable for developing emotional competence in a second language.

For intercultural communication trainers, the findings underscore the importance of developing emotional competence as a key component of intercultural competence. Training programs could incorporate exercises that help individuals recognize and adapt to different cultural patterns of emotional expression, reducing the potential for miscommunication in cross-cultural encounters. The multimodal nature of emotional expression observed in our findings has important implications for intercultural communication training. Since emotions are conveyed through coordinated facial, vocal, and lexical channels, effective communication requires an awareness of how these channels may operate differently across cultures. Training programs could benefit from incorporating exercises that enhance sensitivity to cultural differences in emotional expression patterns, particularly for individuals working in multicultural environments.

For developers of artificial intelligence systems, the results highlight the importance of incorporating cultural context in emotion recognition algorithms. Current AI systems often rely on universal models of emotional expression that may not capture the cultural variations documented in this study. More sophisticated algorithms that account for cultural differences in emotional expression could improve human-computer interactions, particularly in multilingual and multicultural contexts.

Limitations and Future Research Directions

Methodological Limitations

While this study provides valuable insights into emotive speech acts across different language learning contexts, several methodological limitations should be acknowledged.

First, the use of hypothetical scenarios to elicit emotional responses may not fully capture the complexity of authentic emotional expressions in real-world contexts. Although we designed the scenarios to evoke genuine emotional reactions, participants' awareness of the experimental setting may have affected the naturalness of their responses. Future research could employ more naturalistic methods, such as diary studies or ecological momentary assessment, to capture emotional expressions in everyday contexts.

Second, the cross-sectional design of the study limits our ability to draw conclusions about the developmental trajectory of emotional expression in language learning. A longitudinal approach tracking changes in emotional expression patterns over time would provide more robust evidence for the process of emotional acculturation.

Third, while our sample size was adequate for the statistical analyses conducted, the predominantly female composition (80%) limits the generalizability of findings, particularly given potential gender differences in emotional expression. Future studies should aim for more balanced gender representation to examine possible interaction effects between gender, culture, and emotional expression.

Fourth, our focus on three target languages (American English, British English, and Spanish) provides a limited perspective on cultural variations in emotional expression. Expanding the research to include more linguistically and culturally diverse languages, such as tonal languages (e.g., Mandarin Chinese) or languages from non-Western cultures (e.g., Japanese, Arabic), would provide a more comprehensive understanding of the relationship between language, culture, and emotion.

Theoretical Limitations

From a theoretical perspective, our study primarily focused on behavioral manifestations of emotional expression without directly examining underlying cognitive and neural processes. This limits our ability to draw conclusions about how language learning might reshape the

cognitive architecture of emotion processing. Future research could incorporate neuroimaging techniques (such as fMRI or EEG) to investigate the potential neural reorganization associated with emotional adaptation in second language learning.

Additionally, our research framework conceptualized emotions primarily in terms of discrete categories (happiness, anger, sadness). While this approach has strong precedent in the emotion literature, dimensional models that conceptualize emotions along continuous dimensions (e.g., valence, arousal, dominance) offer complementary perspectives. Future research could integrate discrete and dimensional approaches to provide a more nuanced understanding of emotional expression across cultures.

Future Research Directions

Building on the findings and limitations of this study, several promising directions for future research emerge:

Longitudinal Studies of Emotional Acculturation: Tracking changes in emotional expression patterns over the course of language learning would provide valuable insights into the developmental trajectory of emotional acculturation. Such studies could examine whether different aspects of emotional expression (e.g., facial, vocal, lexical) adapt to target culture norms at different rates or through different mechanisms.

Neural Correlates of Emotional Adaptation: Neuroimaging studies could investigate whether extended exposure to a second language and culture leads to neural reorganization in brain regions associated with emotional processing. This would contribute to our understanding of the neuroplasticity of emotion systems and their relationship to language learning.

Cultural Frame Switching: Research on bicultural individuals suggests that they may “switch” between cultural frames depending on contextual cues. Future studies could examine whether advanced language learners demonstrate similar frame-switching in their emotional expression patterns when moving between their native and target languages.

Technology-Mediated Emotional Communication: As intercultural communication increasingly occurs through digital channels (e.g., video calls, text messages), future research could investigate how digital mediation affects the expression and perception of emotions across cultures. This would have important implications for designing culturally sensitive communication technologies.

Pedagogical Interventions for Emotional Competence: Experimental studies could test the effectiveness of different pedagogical approaches for developing emotional competence in a second language. Such research would help language educators integrate emotional aspects of communication more effectively into their teaching practices.

Emotional Intelligence and Cross-Cultural Adaptation: Future research could examine the relationship between emotional intelligence and the ability to adapt emotional expression patterns to a new cultural context. This would contribute to our understanding of individual differences in cross-cultural adaptation.

Advanced Data Analysis Techniques

Classification and Clustering Analysis

To extend our understanding of emotional expression patterns across language groups, we employed additional advanced classification and clustering techniques beyond those initially reported.

Random Forest Classification

We implemented a Random Forest classification algorithm to identify the most discriminative features for distinguishing between emotional states across language groups. The model was trained on a combined feature set including acoustic parameters, facial action units, and lexical features, with a 70/30 train-test split and 10-fold cross-validation.

The Random Forest classifier achieved an overall accuracy of 82.4% in distinguishing across emotional expressions (happiness, anger, sadness) across all language groups. Feature importance analysis revealed that the five most salient features were:

- a) Vocal intensity range (relative importance: 0.86)

- b) AU12 activation intensity (relative importance: 0.79)
- c) Speech rate (relative importance: 0.75)
- d) Proportion of emotion-specific lexical items (relative importance: 0.71)
- e) Fundamental frequency variation (relative importance: 0.68)

Notably, the model performed best for Spanish learners (accuracy: 87.2%), followed by American English learners (accuracy: 81.3%) and British English learners (78.6%). This suggests that Spanish learners may exhibit more distinctive emotional expression patterns across modalities, potentially reflecting greater expressiveness in Spanish culture.

The confusion matrix revealed that misclassifications were most common between anger and sadness expressions, particularly for British English learners. This finding suggests that negative emotions may be expressed more similarly in British English culture, potentially reflecting cultural norms that encourage restraint in emotional expression.

Hierarchical Clustering with Dynamic Time Warping

To account for the temporal dynamics of emotional expressions, we applied hierarchical clustering with Dynamic Time Warping (DTW) distance measures to time-series data of acoustic and facial features. This approach allows for alignment of temporal sequences with different durations and rates, providing a more nuanced understanding of emotional expression dynamics.

The DTW-based clustering revealed three primary temporal patterns in emotional expressions:

Rapid Onset-Extended Plateau: Characterized by quick activation of emotional markers followed by sustained expression. This pattern was most common for happiness expressions among Spanish learners (observed in 68% of cases).

Gradual Buildup-Rapid Decline: Featured a progressive intensification of emotional markers followed by quick deactivation. This pattern was predominant for anger expressions across all language groups (observed in 57% of cases).

Delayed Peak-Extended Decline: Showed a later emotional intensity peak followed by gradual deactivation. This pattern was most common for sadness expressions (observed in 72% of cases).

Interestingly, while the general temporal patterns were consistent across language groups, the timing parameters varied significantly. Spanish learners exhibited faster onset times for happiness expressions ($M = 0.8$ seconds, $SD = 0.3$) compared to American English learners ($M = 1.2$ seconds, $SD = 0.4$), $p = .003$, $d = 1.13$, suggesting quicker emotional activation in line with more expressive cultural norms.

Multimodal Integration Analysis

To better understand how different channels of emotional expression interact, we conducted a series of analyses examining the integration of acoustic, facial, and lexical features.

Canonical Correlation Analysis

Canonical Correlation Analysis (CCA) was applied to examine relationships between sets of variables across modalities (acoustic-facial, acoustic-lexical, facial-lexical). The analysis revealed significant canonical correlations between:

Acoustic and Facial Features: First canonical correlation $r = .76$, $p < .001$, suggesting strong coordination between vocal and facial channels of emotional expression.

Acoustic and Lexical Features: First canonical correlation $r = .64$, $p < .001$, indicating moderate-to-strong relationships between vocal characteristics and verbal emotion expression.

Facial and Lexical Features: First canonical correlation $r = .58$, $p < .001$, showing moderate relationships between facial expressions and verbal emotion communication.

Importantly, the strength of these canonical correlations varied by language group. Spanish learners showed stronger acoustic-facial integration ($r = .82$) compared to American English ($r = .72$) and British English learners ($r = .74$), $p = .021$, suggesting more coherent multimodal expression patterns. This finding aligns with research suggesting that more emotionally expressive cultures may exhibit stronger coordination between different channels of emotional communication.

Sequential Analysis

To examine the temporal coordination between different modalities, we conducted sequential analysis using time-aligned data for acoustic, facial, and lexical features. Cross-

recurrence quantification analysis revealed significant patterns in the temporal coordination of emotional cues:

1. For happiness expressions, facial cues typically preceded vocal cues by an average of 0.4 seconds ($SD = 0.2$), with lexical elements following shortly after. This sequential pattern was consistent across language groups, suggesting a potentially universal aspect of multimodal happiness expression.

2. For anger expressions, vocal intensity increases typically preceded facial action unit activation by an average of 0.3 seconds ($SD = 0.2$), with anger-specific lexical items appearing concurrently with peak vocal intensity. This pattern was most pronounced in American English learners.

3. For sadness expressions, all three modalities showed more synchronous activation without clear leading channels, potentially reflecting the more subdued and integrated nature of sadness expression.

These findings suggest that different emotions may have characteristic sequential patterns in multimodal expression, with possible cultural variations in the precise timing and coordination of these patterns.

Regression and Predictive Modeling

To identify predictors of emotional expression adaptation across language groups, we implemented multiple regression modeling and machine learning approaches.

Multiple Regression Models

We developed separate multiple regression models for each emotional expression channel (acoustic, facial, lexical) to identify significant predictors of adaptation to target language norms. For each model, we calculated a “cultural approximation index” representing how closely participants’ expressions matched typical patterns in the target language culture (based on published normative data).

For facial expressiveness in happiness expressions, the regression model explained 68.3% of variance (adjusted $R^2 = .683$, $F(6, 93) = 35.26$, $p < .001$). Significant predictors included:

- a) Cultural immersion duration ($\beta = .42$, $p < .001$)
- b) Target culture media consumption ($\beta = .38$, $p < .001$)
- c) Self-reported cultural affinity ($\beta = .27$, $p = .003$)
- d) Language proficiency ($\beta = .24$, $p = .018$)
- e) Gender ($\beta = .19$, $p = .032$)

Importantly, language proficiency alone explained only 27.4% of variance, with cultural exposure variables accounting for an additional 39.1%. This supports the hypothesis that emotional expression adaptation is driven more by cultural learning than by linguistic competence per se.

For acoustic features of emotional expression, a similar pattern emerged, with cultural immersion ($\beta = .37$, $p < .001$) and media exposure ($\beta = .35$, $p < .001$) emerging as stronger predictors than language proficiency ($\beta = .29$, $p = .006$).

Support Vector Machine (SVM) Predictive Modeling

To develop a predictive model of emotional adaptation, we implemented an SVM with Radial Basis Function kernel, using participants’ background variables to predict their emotional expression patterns. The model achieved an overall accuracy of 79.6% in predicting whether participants’ emotional expressions would closely match target language norms (defined as falling within 1 standard deviation of published normative data for native speakers).

The most influential features in the model, as determined by recursive feature elimination, were:

- a) Years of target culture immersion
- b) Frequency of interaction with native speakers
- c) Consumption of target culture media (hours per week)
- d) Self-reported cultural identification
- e) Formal language study duration

These findings further support the critical role of cultural exposure and engagement in facilitating emotional expression adaptation, beyond formal language study.

In-Depth Analysis of Testing Instruments and Stimuli

Detailed Analysis of Elicitation Scenarios

The emotion elicitation scenarios were developed through a rigorous process to ensure cross-cultural validity and emotional specificity. Prior to the main study, we conducted a validation phase with 45 participants (15 from each target language group) who rated potential scenarios on two dimensions: emotional specificity (how clearly the scenario evoked the target emotion) and cultural equivalence (how similarly the scenario would be interpreted across cultures).

From an initial pool of 12 scenarios (four per emotion), we selected the three scenarios that demonstrated the highest ratings for both emotional specificity (mean rating > 4.2 on a 5-point scale) and cultural equivalence (mean rating > 4.0 on a 5-point scale).

The selected scenarios were further refined through cognitive interviews with bilingual consultants to ensure linguistic and cultural appropriateness across all three target languages. Particular attention was paid to ensuring that culturally specific references were either removed or adapted to maintain equivalent emotional salience.

In addition to the previously described scenarios, the testing instruments included follow-up questions to enhance emotional engagement and ensure consistent interpretation:

Happiness Scenario Enhancement: “How would this help make you feel better during your difficult time? What specific words would you use to let your friend know how much their help means to you?”

Anger Scenario Enhancement: “How has this false accusation affected your reputation? What exact words would you use to confront the person who made the accusation?”

Sadness Scenario Enhancement: “How close were you to this person? What specific memories would you share with your friend when telling them about the loss?”

These follow-up questions served to deepen emotional engagement while providing structured guidance for responses, ensuring greater consistency across participants.

Acoustic Analysis Instruments and Procedures

The acoustic analysis involved a more sophisticated set of procedures than initially described. In addition to the primary Praat software analysis, we employed multiple complementary approaches:

Acoustic Feature Extraction: We utilized the openSMILE acoustic feature extraction toolkit (version 2.3.0) to compute the extended Geneva Minimalistic Acoustic Parameter Set (eGeMAPS), which includes 88 standardized acoustic parameters relevant for affective computing.

Voice Quality Analysis: We applied the Glottal Enterprise algorithm to extract measures of vocal effort, including H1-H2 (the difference between the first and second harmonic amplitudes), Harmonic Noise Ratio (HNR), and Quasi-Open Quotient (QQQ). These measurements provide detailed information on the physiological aspects of emotional vocalization.

Prosodic Modeling: We implemented the MOMEL (MOdeling MELody) algorithm to model pitch contours, identifying anchor points that characterize the prosodic structure of emotional utterances. This was complemented by INTSINT (INternational Transcription System for INTonation) coding to standardize the description of intonation patterns across languages.

Emotional Acoustic Fingerprinting: We developed customized scripts to identify emotion-specific acoustic signatures by applying machine learning techniques, using the scikit-learn library in Python, to the extracted features. This approach allowed us to identify acoustic patterns characteristic of each emotion across language groups.

To ensure measurement reliability, all acoustic analyses were conducted by two independent analysts, and inter-rater reliability was calculated using intraclass correlation coefficients (ICC). The mean ICC for acoustic measurements was .91 (range: .87-.95), indicating excellent reliability.

Facial Expression Analysis Instruments and Procedures

Facial expression analysis employed multiple complementary approaches beyond basic FACS coding:

Automated Facial Analysis: In addition to manual FACS coding, we utilized FaceReader software (version 8.0, Noldus Information Technology) to provide automated analysis of facial expressions. This system applies artificial intelligence algorithms to detect basic emotions and facial action units, providing an objective complement to manual coding.

Dimensional Affect Analysis: We implemented continuous measurement of valence and arousal dimensions using the FEELtrace system, which allows raters to continuously track the perceived emotional state of participants in a 2D affective space. This provided a dynamic record of emotional transitions that complemented the categorical FACS analysis.

Micro-Expression Detection: We employed high-speed video recording (200 frames per second) and specialized analysis techniques to identify micro-expressions — brief, involuntary facial expressions that may reveal emotional leakage. This analysis was particularly valuable for examining cultural display rules that may suppress certain emotional expressions.

Cultural Display Rule Assessment: We developed a custom coding scheme to identify culture-specific facial expression patterns not captured by standard FACS. This scheme was based on prior ethnographic research on display rules in American, British, and Spanish cultures, and was validated through consultation with cultural informants.

Facial expression data was analyzed by three certified FACS coders who were blind to experimental conditions, with a subset of recordings (30%) coded by all three coders to establish reliability. Krippendorff's alpha for inter-coder reliability was .88 for AU presence/absence and .84 for AU intensity ratings, indicating strong reliability.

Lexical Analysis Instruments and Procedures

Lexical analysis employed specialized techniques and software to examine emotional language use across language groups:

Custom Emotional Lexicon Development: We constructed language-specific emotional lexicons for each target language by combining existing resources (e.g., LIWC, NRC Emotion Lexicon) with corpus-derived emotion terms from contemporary sources. These lexicons were validated by native speakers, who rated terms for emotional specificity and intensity.

Computational Semantic Analysis: We employed latent semantic analysis (LSA) and word embeddings (Word2Vec) to examine semantic relationships between emotion terms across languages. This allowed us to identify cross-linguistic differences in how emotion concepts are structured and related.

Transitivity Analysis: Following Systemic Functional Linguistics framework, we analyzed transitivity patterns in emotional expressions, examining how participants construed emotions as processes, attributes, or entities across languages. This revealed important differences in how emotions were conceptualized grammatically.

Metaphor Identification and Analysis: We applied the Metaphor Identification Procedure Vrije Universiteit (MIPVU) to systematically identify metaphorical expressions related to emotions. This revealed cross-linguistic differences in the conceptual metaphors used to express emotional states (e.g., “anger is heat” vs “anger is pressure”).

To ensure coding reliability, all qualitative analyses were conducted by multiple coders with expertise in the respective target languages. Cohen's kappa for metaphor identification was .83, and Krippendorff's alpha for transitivity coding was .81, indicating strong reliability.

Results Extension and Integration

Cultural Adaptation Trajectories

To examine the developmental trajectory of emotional expression adaptation, we conducted additional analyses based on participants' language learning histories. Participants were categorized according to their duration of target language study (2–3 years, 4–5 years, 6+ years), we were able to identify distinct patterns in the adaptation of emotional expression.

A multivariate analysis of variance (MANOVA) with language exposure duration as the independent variable revealed a significant main effect on emotional expression measures, Wilks' $\lambda = .63$, $F(12, 184) = 4.08$, $p < .001$, partial $\eta^2 = .21$. Follow-up univariate ANOVAs showed significant effects of language exposure duration on:

Facial Expression Adaptation: $F(2, 97) = 8.46$, $p < .001$, partial $\eta^2 = .15$

Acoustic Adaptation: $F(2, 97) = 6.92$, $p = .002$, partial $\eta^2 = .13$

Lexical Adaptation: $F(2, 97) = 11.34$, $p < .001$, partial $\eta^2 = .19$

Post-hoc comparisons revealed that facial expression adaptation showed the earliest changes, with significant differences between the 2–3 year and 4–5 year groups ($p = .008$, $d = 0.72$). Acoustic features showed a more gradual adaptation, with significant differences emerging only between the 2–3 year and 6+ year groups ($p = .003$, $d = 0.85$). Lexical adaptation showed the most linear progression, with significant increments between each successive duration group.

These findings suggest a sequential pattern of emotional adaptation, with non-verbal channels (particularly facial expressions) adapting earlier than verbal (lexical) channels. This pattern was consistent across language groups but occurred at different rates, with Spanish learners showing more rapid adaptation in facial expressiveness compared to English learners.

Emotional Intelligence and Cultural Adaptation

To explore individual differences in emotional adaptation, we administered the Trait Emotional Intelligence Questionnaire-Short Form (TEIQue-SF) and the Cultural Intelligence Scale (CQS) across all participants. These measures were then correlated with indices of emotional expression adaptation.

Multiple regression analysis revealed that emotional intelligence (EI) and cultural intelligence (CQ) jointly accounted for 42.3% of the variance in overall emotional adaptation ($R^2 = .423$, $F(2, 97) = 35.62$, $p < .001$). However, the relative contribution of these factors varied by language group and emotional channel:

For American English learners, CQ was a stronger predictor of emotional adaptation ($\beta = .43$, $p < .001$) than EI ($\beta = .29$, $p = .018$), particularly for lexical choices.

For British English learners, a similar pattern emerged, with CQ ($\beta = .46$, $p < .001$) outweighing EI ($\beta = .26$, $p = .024$).

For Spanish learners, EI played a more prominent role ($\beta = .41$, $p < .001$) alongside CQ ($\beta = .39$, $p < .001$), particularly in relation to facial expression adaptation.

These findings suggest that while general emotional abilities contribute to emotional adaptation in language learning, cultural intelligence—the ability to function effectively in culturally diverse settings—may be equally or even more important, particularly for languages with more subtle or restrained emotional display rules.

Neurophysiological Correlates of Emotional Adaptation

A subset of participants ($n = 45$; 15 from each language group) completed additional testing involving physiological measures during emotional expression tasks. These measures included:

Galvanic Skin Response (GSR): Measuring autonomic arousal during emotional expression

Heart Rate Variability (HRV): Assessing parasympathetic nervous system activity

Facial Electromyography (EMG): Recording electrical activity in facial muscles (zygomaticus major and corrugator supercilii)

Repeated measures ANOVA revealed significant differences in physiological responses during emotional expressions in native versus target languages. When expressing emotions in their target language, participants showed:

Reduced GSR amplitude for anger expressions compared to native language expressions ($M_{\text{difference}} = 0.82 \mu\text{S}$, $p < .001$, $d = 0.94$), suggesting decreased autonomic arousal.

Increased HRV during sadness expressions ($M_{\text{difference}} = 8.3 \text{ ms}$, $p = .006$, $d = 0.78$), indicating greater regulatory control.

Differential facial EMG patterns, with happiness expressions showing comparable zygomaticus activity across languages, but anger expressions showing reduced corrugator activity in the target language ($M_{\text{difference}} = 11.6 \mu\text{V}$, $p < .001$, $d = 1.02$).

These physiological findings suggest that emotional expression in a second language may involve reduced emotional reactivity and increased regulatory control — a phenomenon consistent with the “foreign language effect” described in decision-making literature. However, this effect varied significantly across language group, with Spanish learners showing physiological responses more similar to those in their native language, particularly in expressions of happiness.

Conclusions

This study examined the complex interplay between universal emotional markers and culturally specific expression patterns in language learning environments. Our findings support a nuanced theoretical model that integrates universalist and relativist perspectives on emotional expression, suggesting that language learners navigate a dynamic space between these poles.

The research confirms that certain aspects of emotional expression — such as increased vocal intensity for anger and decreased speech rate for sadness — remain relatively consistent across language groups, supporting the universality hypothesis. However, other aspects — particularly facial expressiveness in happiness and lexical choices for emotional states — show significant adaptation to target language norms, supporting the cultural relativity perspective.

Our data reveals that language learners develop what might be termed an “emotional interlanguage” — a dynamic system of emotional expression that incorporates elements from both their native emotional repertoire and the target language’s cultural norms. This emotional interlanguage evolves with increased language proficiency and cultural exposure, but the adaptation process varies across different channels of emotional expression and across emotions.

The finding that cultural familiarity mediates the relationship between language proficiency and emotional expressiveness suggests that emotional adaptation in language learning is not simply a function of linguistic knowledge, but requires deeper cultural learning and engagement. This supports a model of language acquisition that places greater emphasis on pragmatic and cultural competence alongside the traditional focus on grammatical and lexical knowledge.

This study contributes to our understanding of the complex relationship between language, culture, and emotion — a relationship that lies at the heart of human communication and social connection. By examining how language learners navigate the challenges of expressing emotions across linguistic and cultural boundaries, we gain insights not only into second language acquisition, but also into the fundamental nature of emotional expression.

The “emotional interlanguage” concept proposed in this study offers a promising framework for understanding how individuals adapt their emotional expression patterns as they move between cultural contexts. This concept may have relevance beyond language learning, with potential applications for research on migration, intercultural relationships, and globalization.

As global mobility increases and intercultural interactions become more common, the ability to effectively communicate emotions across cultural boundaries gains importance. Our findings suggest that this ability is not simply acquired through language learning but requires deeper cultural engagement and awareness. This highlights the need for educational and training approaches that address the cultural dimensions of emotional communication alongside linguistic aspects.

Looking forward, the increasing integration of artificial intelligence into communication systems raises new questions about how machines might recognize, interpret, and express emotions across cultural contexts. The insights from this study regarding cultural variations in emotional expression patterns could inform the development of more culturally sensitive AI systems.

In conclusion, this research illuminates the intricate play between universal and culture-specific aspects of emotional expression, revealing how language learners progressively adapt their emotional communication patterns to new cultural contexts. This adaptation process represents a fundamental aspect of becoming communicatively competent in a second language — one that deserves greater attention in language education, intercultural training, and communication technology development.

Adherence to Ethical Standards

The study adhered to strict ethical guidelines to ensure the well-being and confidentiality of participants. All participants provided informed consent before participating in the study, and they were informed of their right to withdraw at any time without penalty. The audio and video recordings were anonymized, and participant identities were kept confidential. The study protocol was approved by the institutional review board of the Azerbaijan University of Languages.

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EMOTIVE SPEECH ACTS IN CROSS-CULTURAL COMMUNICATION: A COMPREHENSIVE ANALYSIS AND EXPERIMENTAL STUDY

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The aim of the study is to determine the role of emotive speech acts in cross-cultural language learning environments, revealing the complex interplay between universal emotional markers and culturally specific expression patterns. In the course of the research, data analysis *methods* were applied (acoustic analysis, facial expression analysis using the Facial Action Coding System (FACS), lexical analysis, correlational and regression analysis).

Through comprehensive analysis of acoustic features, facial expressions, and lexical patterns, the research demonstrates that emotional expression follows dual patterns: universal elements remain consistent across languages while others undergo significant cultural adaptation. Results indicate that language learners develop an “emotional interlanguage” that synthesizes native expression strategies with target language norms. Spanish learners exhibited greater facial expressiveness when expressing happiness, suggesting adoption of the target culture’s more overt emotional display rules. Anger was more explicitly verbalized across all language learning groups, indicating that different emotions utilize distinct channels of expression. Principal component analysis and hierarchical clustering revealed discrete emotional expression profiles across language groups, while multiple regression models identified predictive relationships between linguistic proficiency, cultural exposure, and emotional adaptation. Our findings support a nuanced theoretical model that integrates universalist and relativist perspectives on emotional expression, suggesting that language learners navigate a dynamic space between these poles.

The research confirms that certain aspects of emotional expression — such as increased vocal intensity for anger and decreased speech rate for sadness — remain relatively consistent across language groups, supporting the universality hypothesis. However, other aspects — particularly facial expressiveness for happiness and lexical choices for emotional states — show significant adaptation to target language norms, supporting the cultural relativity perspective.

Our data reveals that language learners develop what might be termed an “emotional interlanguage” — a dynamic system of emotional expression that incorporates elements from both their native emotional repertoire and the target language’s cultural norms. This emotional interlanguage evolves with increased language proficiency and cultural exposure, but the adaptation process varies across different channels of emotional expression and across different emotions.

The finding that cultural familiarity mediates the relationship between language proficiency and emotional expressiveness suggests that emotional adaptation in language learning is not simply a function of linguistic knowledge, but requires deeper cultural learning and engagement.

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