Human Centered Evaluation of CLAP-Based Emotion Annotations for Therapeutic Music Using MTurk

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Motivation

- CLAP (Contrastive Language–Audio Pretraining) generates emotion tags by aligning audio & text embeddings.
- Enables large-scale music annotation without manual labeling.
- Outputs are weakly supervised and may not align with human emotional perception.
- Our work: Develops an MTurk-based framework to validate & refine CLAP-generated annotations.
- Can help recommendations for music based on desired emotional profiles

Research Questions

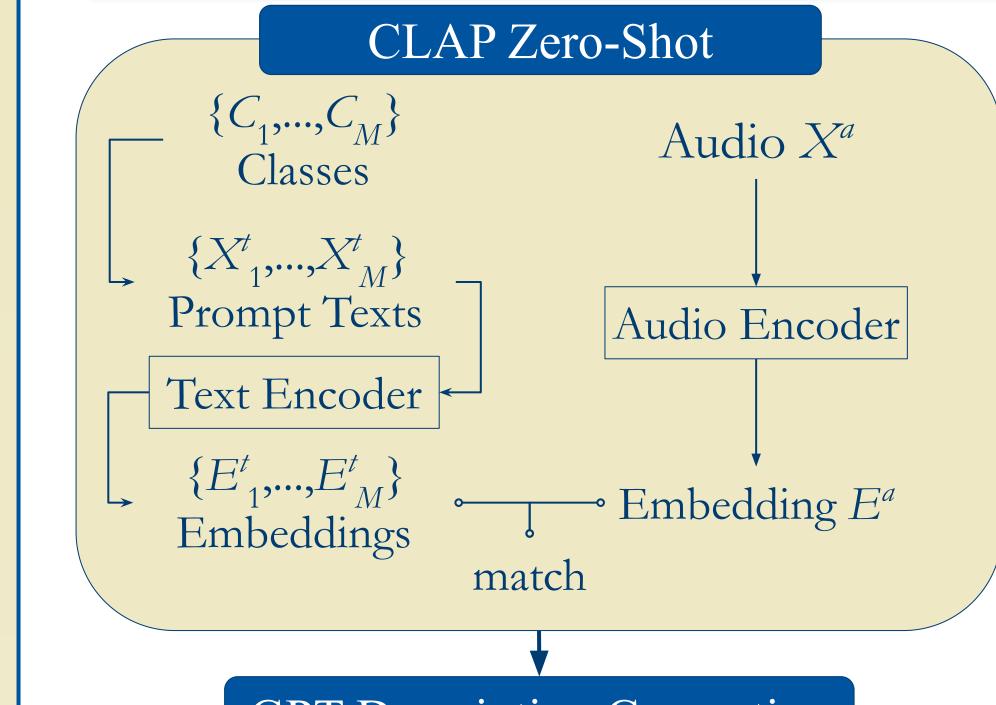
- ① **Can** contrastive learning models like CLAP reliably generate emotionally coherent tags for therapeutic music without explicit supervision?
- 2 How closely do CLAP-generated emotion annotations align with human emotional perception and interpretation?
- © Can iterative refinement techniques, such as pseudo-labeling based on human feedback, enhance the validity and interpretability of weakly-labeled datasets?

Methodology

Dataset Preparation

- Source: uCue's music library containing 10k+ possible song combinations.
- Selection criteria: Chose 432 tracks for diversity in emotion features.
- Structure: Modular arrangements with separate layers that can be added/removed.
- Metadata: Includes song IDs, modular layer info, and predefined emotion feature definit
- **Purpose**: Provide a representative subset of the full library for CLAP annotation and human validation.

Methodology cont.



GPT Description Generation

Prompt

- 3 most relevant features
- Feature definitions

Output

GPT generates a concise natural language description of the track's emotional character

Crowdsourced Evaluation (MTurk)

- Participants: Pre-qualified workers who passed a gold-standard clearance test.
- Task Flow:
 - Listen to the full audio clip.
- 2. Rank the top 3 emotion features.
 - a. 8 features given with definition drop-down available
- 3. Give 2-3 sentence description.
- 4. Rate agreement with generated description.
 - a. Only shown after "Next" button is clicked to prevent bias

MTurk Interface Workflow

Listen to Audio -

• Full playback required

(skip + mute prevention) **Paple Fe**

(skip + mute prevention) Rank Features

• Select and order top-3

Write Description emotion features

• Write a 2-3 sentence natural

language description

Rate AI Description

• Score GPT description on a Likert scale (-2 to 2)

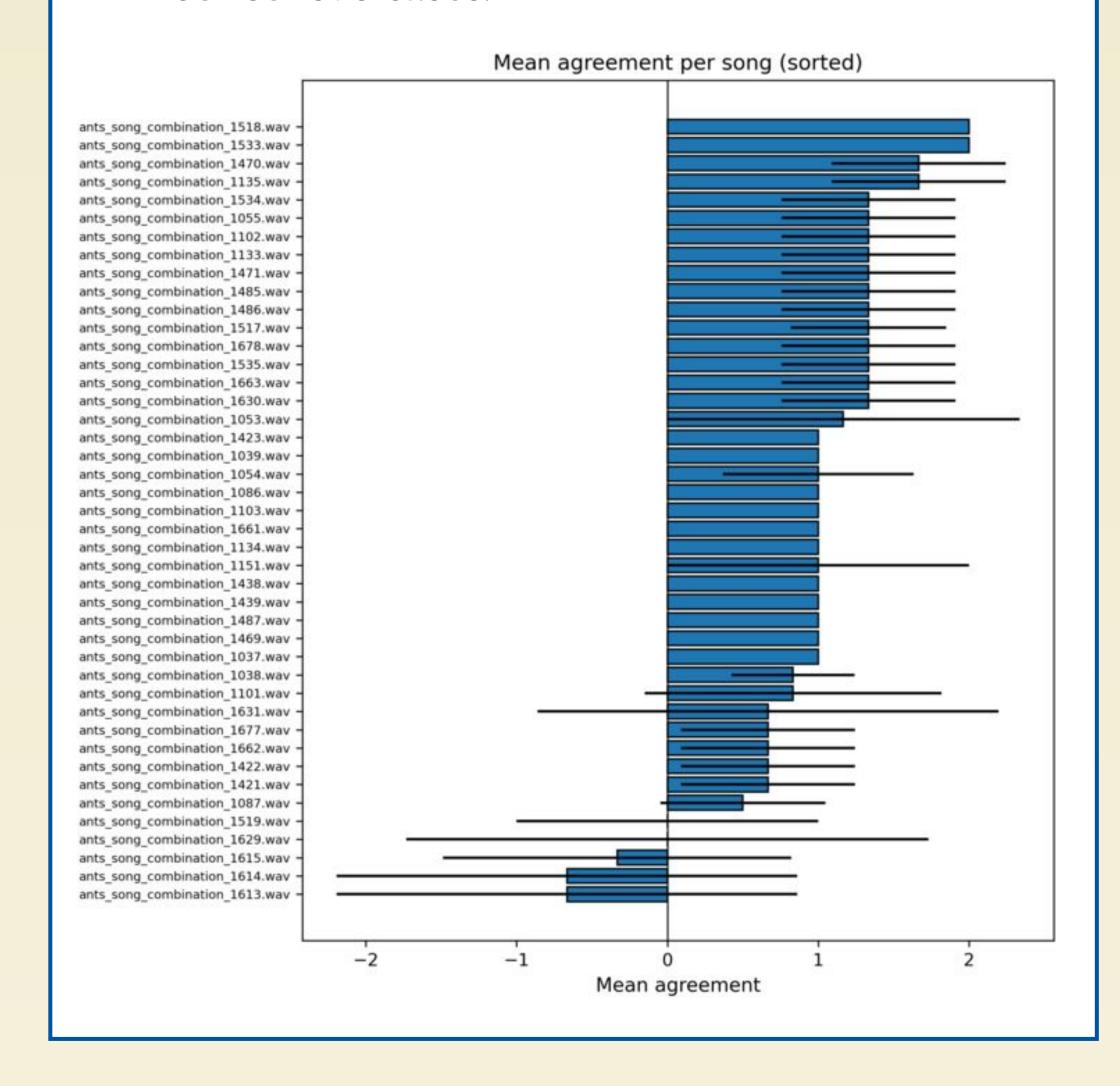
Methodology cont.

Full Workflow Dataset 7 **CLAP** Annotation Audio Encoder → Audio Embedding. Text Encoder → Feature Embeddings. Match via cosine similarity → Top-3 Description in ___ Chat GPT generates natural language description from top-3 features + definitions. MTurk Evaluation • Listen to full clip • Rank top-3 features • Provide human written description • Rate GPT descriptions SQL Database Stores timestamps, rankings, ratings, text feedback, and Analysis 📊 completion code Agreement metrics, Tag overlap, Semantic similarity, Pseudo-label

Preliminary Results

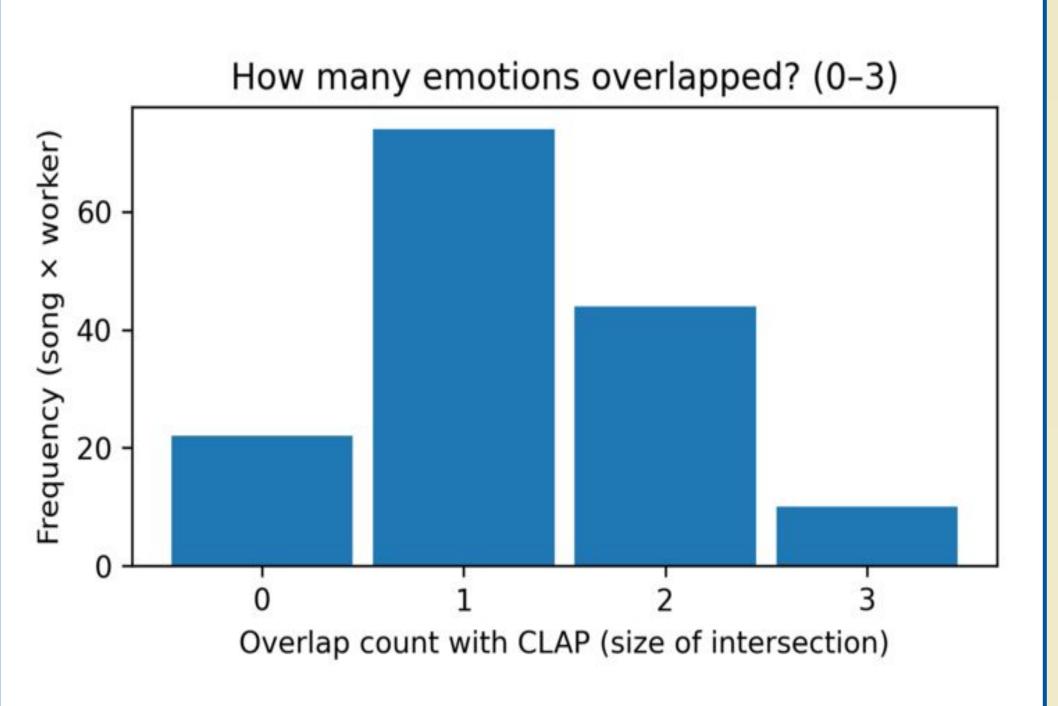
refinement

• Agreement Score Analysis: Mean & variance per song; identify high/low consensus cases.



Preliminary Results cont.

• Top-3 Overlap: Jaccard index & exact match counts between CLAP and human rankings.



Takeaway: CLAP typically shares exactly one label with a given worker.

Future Work

- Complete full dataset evaluation with MTurk.
- Apply pseudo-label refinement to improve annotation accuracy.
- Conduct demographic and contextual analyses of perception patterns.
 - Description Similarity:
 Embedding-based semantic similarity
 between GPT and human descriptions.
 - Cluster Analysis: Identify patterns of disagreement by emotion type or complexity.
- Pseudo-Label Refinement: Use high-agreement samples to improve dataset accuracy.
- Publish validated therapeutic music dataset and open-source evaluation platform.
- Extend methodology to other domains, including environmental audio and affective speech.