Towards Explainable Emotion Recognition in Music: The Route via Mid-level Features



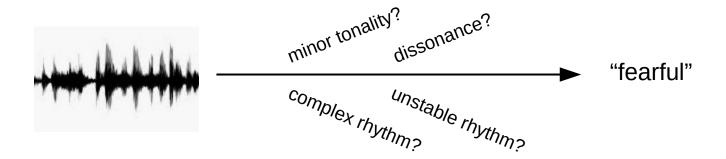
Shreyan Chowdhury, Andreu Vall, Verena Haunschmid, Gerhard Widmer





Motivation

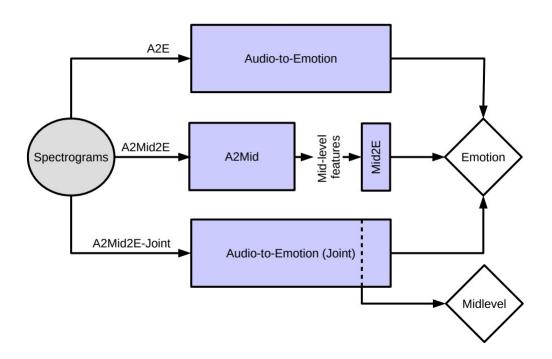
"Emotion" is a subjective descriptor of music.



An approach to get a more precise handle on vague emotions is required, for possible applications in analysis of musical expression, recommendations, production, style transfer, etc.

Approach

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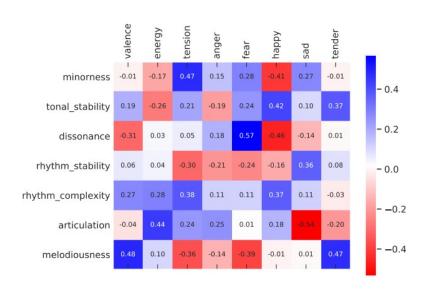
Questions We Addressed

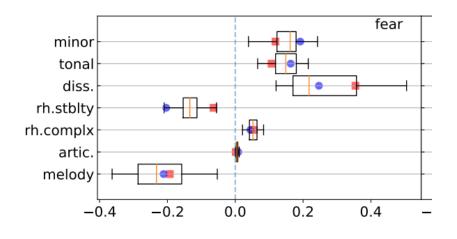
Intermediate linear layer = lower final predictive performance?

Joint training > separate training?

Single instance explanations + interpretable model?

Model-level and Song-level Explantions





Weights

Effects

Poster **B-09**

Shreyan Chowdhury shreyan.chowdhury@jku.at





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Shreyan Chowdhury, Andreu Vall, Verena Haunschmid, Gerhard Widmer Institute of Computational Perception Johannes Kepler University Linz

INTRODUCTION

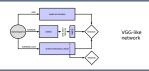
Problem:

It is difficult to interpret emotional predictions in terms of musical content.

To give musically or perceptually meaningful justifications or explanations for predictions.

Train a model with an intermediate linear layer on data annotated with perceptual feature ratings and emotion ratings. The weights of the linear layer can then be used to study the effects of each perceptual feature on each final emotion prediction.

ARCHITECTURE



DATASETS

Mid-level Perceptual Features Dataset: Aljanaki et al., 2018 Mid-level perceptual features are musical qualities that are supposed to be meaningful and intuitively recognizable by most listeners, without requiring music theoretic

Perceptual Feature	Question asked to human raters		
Melodiousness	To which excerpt do you feel like singing along?		
Articulation	Which has more sounds with staccato articulation?		
Rhythmic Stability	Which is easier to march along with?		
Rhythmic Complexity	Difficult to repeat by tapping? Difficult to find the meter? Rhythm has many layers?		
Dissonance	Noisier timbre? Has more dissonant intervals?		
Tonal Stability	Easier to determine the tonic and key?		
Modality ('Minomess')	Which song would have more minor chords?		

Soundtracks Dataset (Emotion Ratings): Eerola et al., 2011

Valence	Energy	Tension	Anger
Fear	Нарру	Sad	Tender

EXPERIMENTS

Training Schemes:

•A2E (Audio-to-Emotion) Predict emotion values directly from spectrogram (baseline).

- A2Mid2E (Audio-to-Mid-level-to-Emotion) Learn a spectrogram to mid-level feature extractor, and a midlevel to emotion predictor separately.
- · A2Mid2E-Joint (Audio-to-Mid-level-to-Emotion-Joint) Learn mid-level feature extractor and emotion predictor jointly.

Song-level Explanations:

· Effects: weights times feature values for the linear layer. Distribution over a set of examples is plotted as boxplots (called effects plots).

$$effect_i^{(i)} = w_i x_i^{(i)}$$

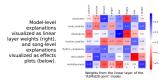
·For a particular song, the effects of each feature contributing to a prediction can be plotted and visualized. Example case: two songs with similar emotion profile but different mid-level feature profile

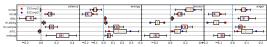
RESULTS

Evaluation metric: Pearson's Correlation

Model	Avg. Emotion Correlation	Cost Type	Cost Value
		CoE _{AZMidZE}	0.05
A2E	0.76		
A2Mid2E	0.71	CoE _{A2Mid2E-Joint}	0.01
A2Mid2E-Joint	0.75	CoE = "Cost of Explainability"	

Explainable predictions of emotion from music can be obtained by introducing an intermediate representation of mid-level perceptual features in the predictor model.





Effects Plot for the dataset (boxes), and two songs (dots) with similar emotion profile but different mid-level features profile





