

Characterizing music for pain relief: a computational approach using music features and deep audio embeddings

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ERASMUS MC
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BACKGROUND AND AIM

- Music shows promise as an adjunctive treatment for acute and chronic pain, reducing the need for pharmacological analgesics and their associated side effects [1].
- While self-selected and favourite music enhances music-induced analgesia [2], the specific characteristics of music used for pain relief remain poorly understood.
- Aim: identify and analyze the characteristics of music for pain relief, selected by participants in two clinical studies, with "everyday music" (Dutch Top 40) using Spotify Audio Features, deep audio embeddings [3], and genre analysis.**

MATERIALS AND METHODS

Pain Music Dataset (PMD)

- 164 playlists / participants.
- ~1.900 tracks.



Everyday Music Dataset (EMD)

- Music from 1980 – 2023.
- ~9.000 tracks.

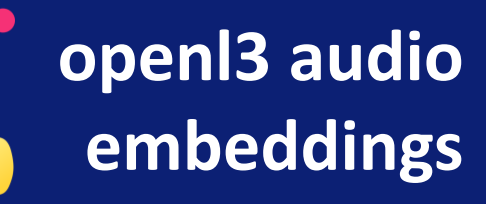


Feature extraction

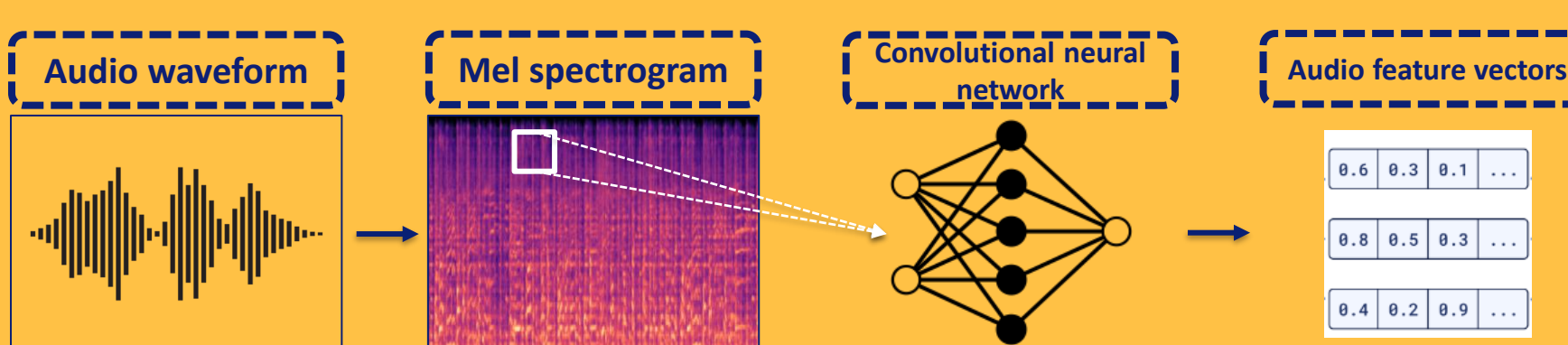


Spotify Audio
Features

- Interpretable music features, such as **danceability**, **energy**, and **valence**, were extracted from the Spotify API using the Python package Spotipy.



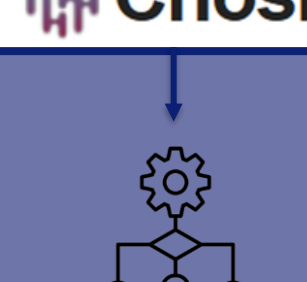
open3 audio
embeddings



PMD and EMD
comparison



Genre analysis



Feature quality
assessment in genre
clustering



RESULTS

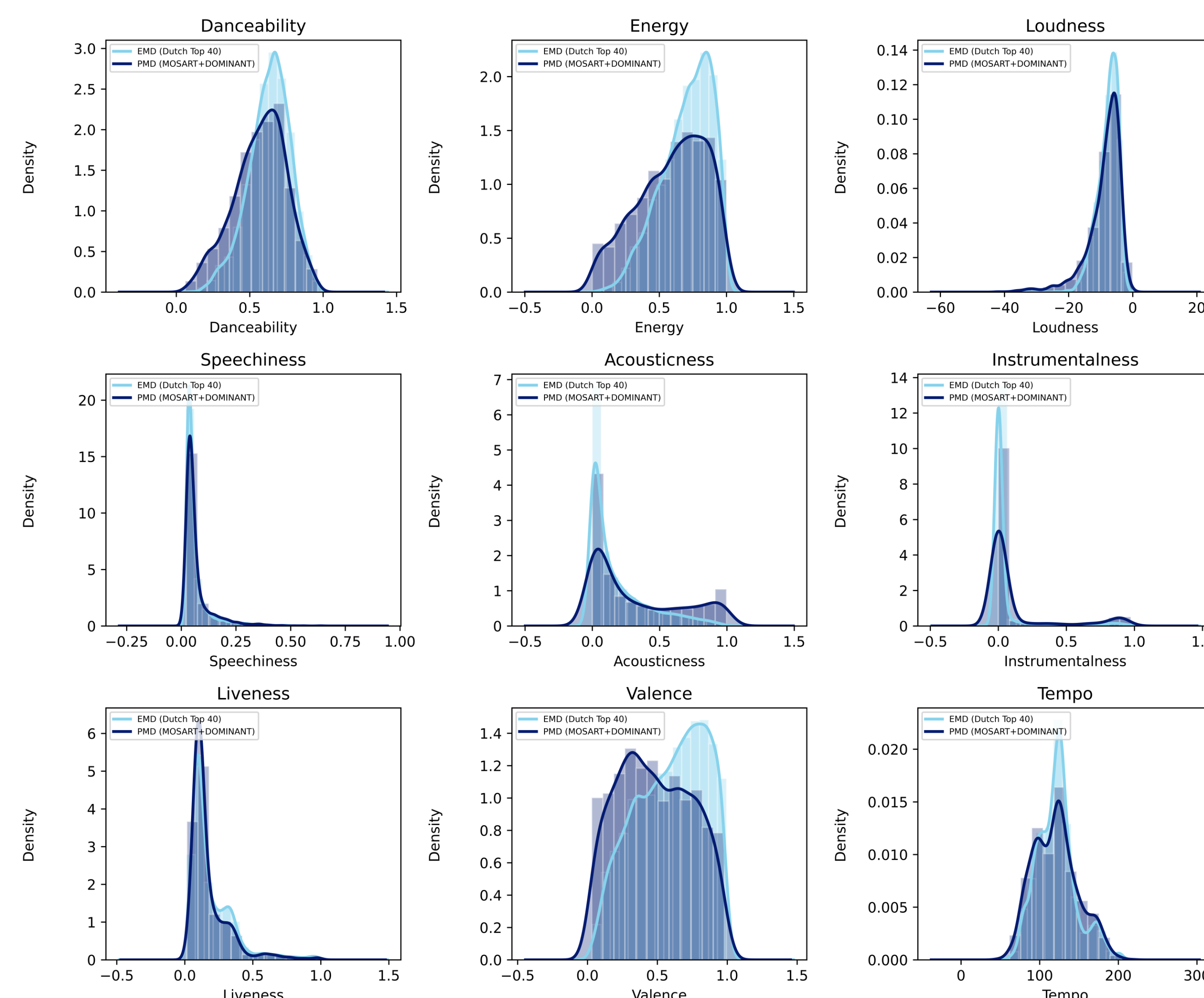


Figure 1. Comparison of Spotify Audio Features distributions between PMD and EMD. All audio features exhibit significant differences as determined by the Mann-Whitney U and the Kolmogorov-Smirnov test.

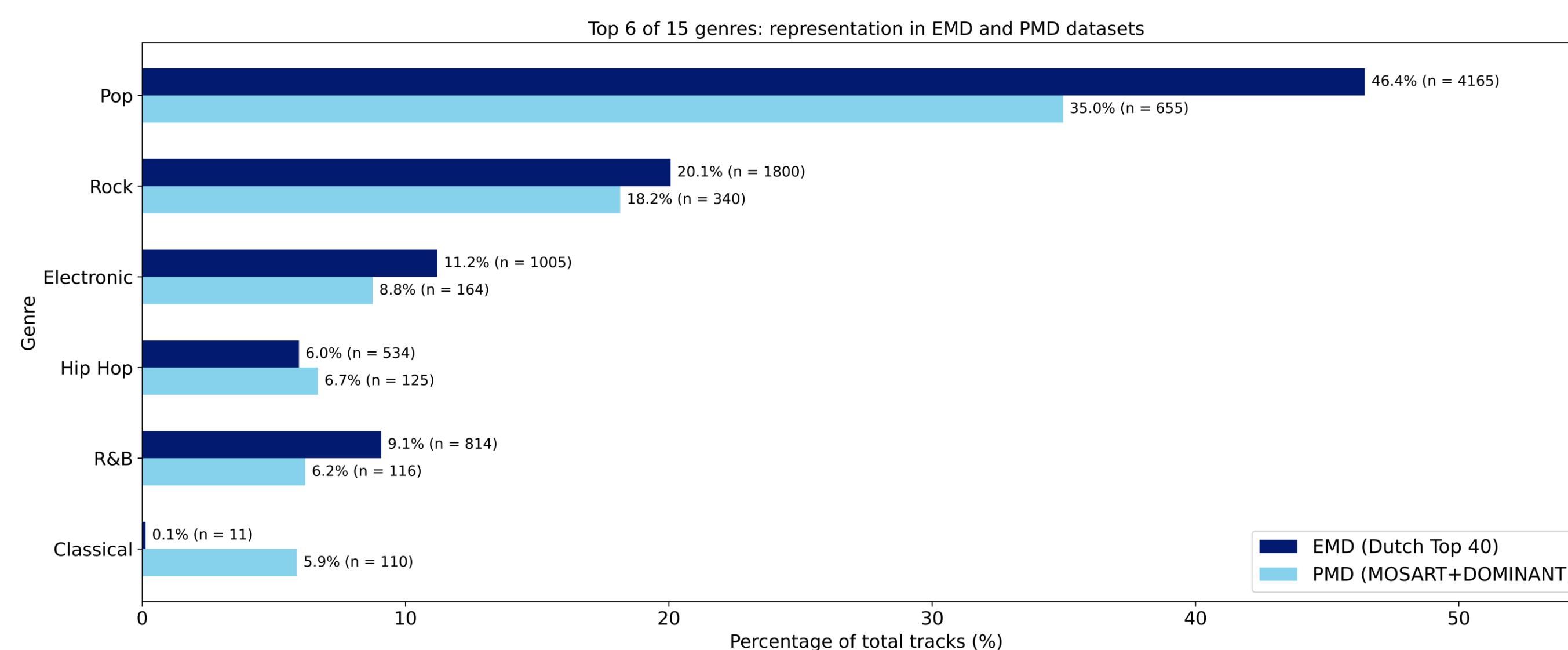


Figure 2. Genre analysis reveals overlap in the number of tracks per genre between EMD and PMD.

- The genre diversity score (entropy) for PMD is **3.06**, compared to **2.38** for EMD, indicating greater genre diversity in PMD.

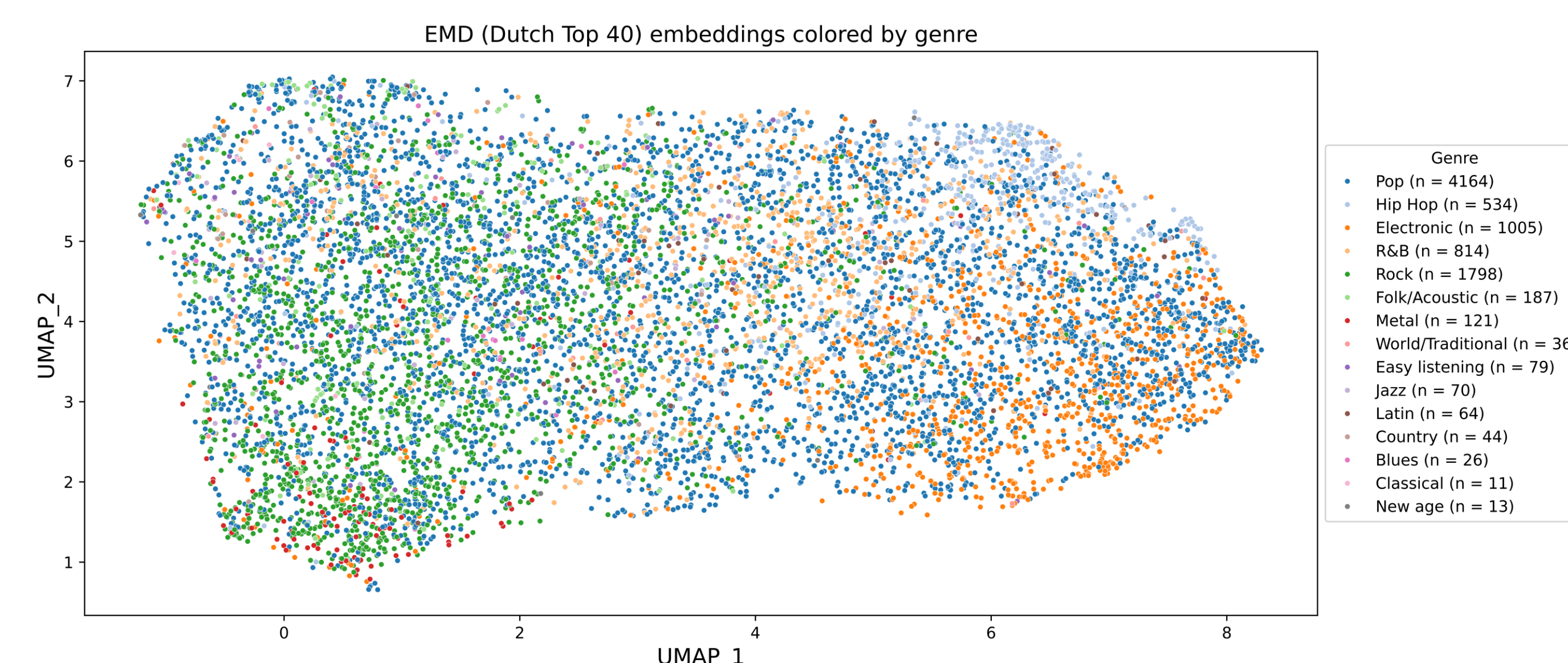


Figure 3. Deep audio embeddings clearly capture and represent genre-specific characteristics.

Pairwise k-means clustering (Everyday Music Dataset)

Feature set	mean Silhouette score	mean Adjusted Rand Index	mean Normalized Mutual Information
Spotify Audio Features	0.4270	0.0243	0.0309
open3 audio embeddings	0.4848	0.1050	0.0967

CONCLUSION AND RELEVANCE FOR PATIENT CARE

- Our analysis revealed that music chosen by individuals against pain significantly diverges from the mainstream music typically played on the radio (EMD).
- This variation underscores the need to account for these differences when implementing music in clinical settings.
- Given the discontinuation of Spotify Audio Features through the Spotify API, this project emphasizes the potential of deep audio embeddings as a viable and effective approach for music characterization and future research into personalized music.

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