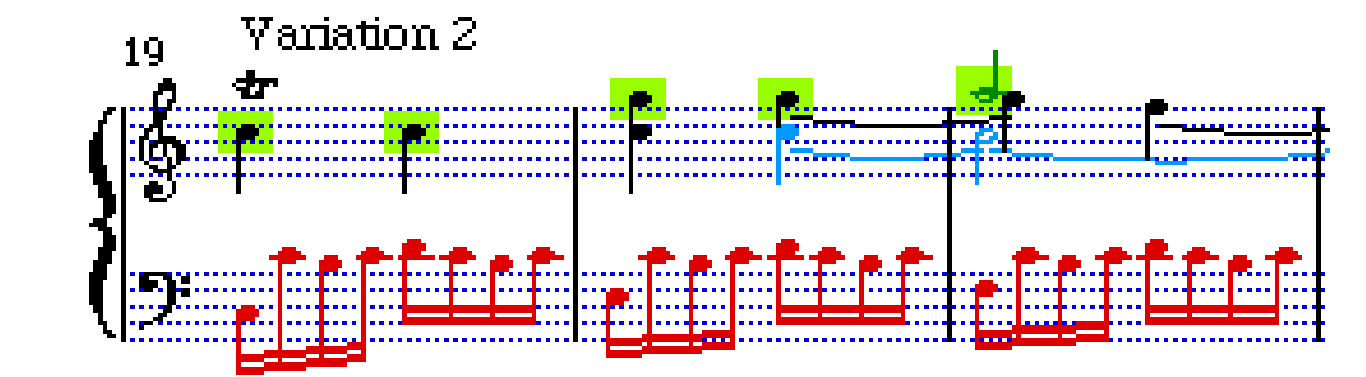


From Multi-labeling to Multi-domain-labeling: A Novel Two-dimensional Approach to Music Genre Classification

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Main Idea

Problem:

- Often multiple genre labels are assigned to music tracks due to multiple genre influences
- Problematic ground-truth for training classification models

Proposed solution:

- Apply temporal segmentation
- Use musical domains ("dimensions"):
 - Timbre
 - Rhythm
 - Melody/Harmony
- Assign genre labels to the most appropriate domain in each time segment
- Break down the multi-label into multiple single-label classification tasks

Gain about 10% of F-measure!

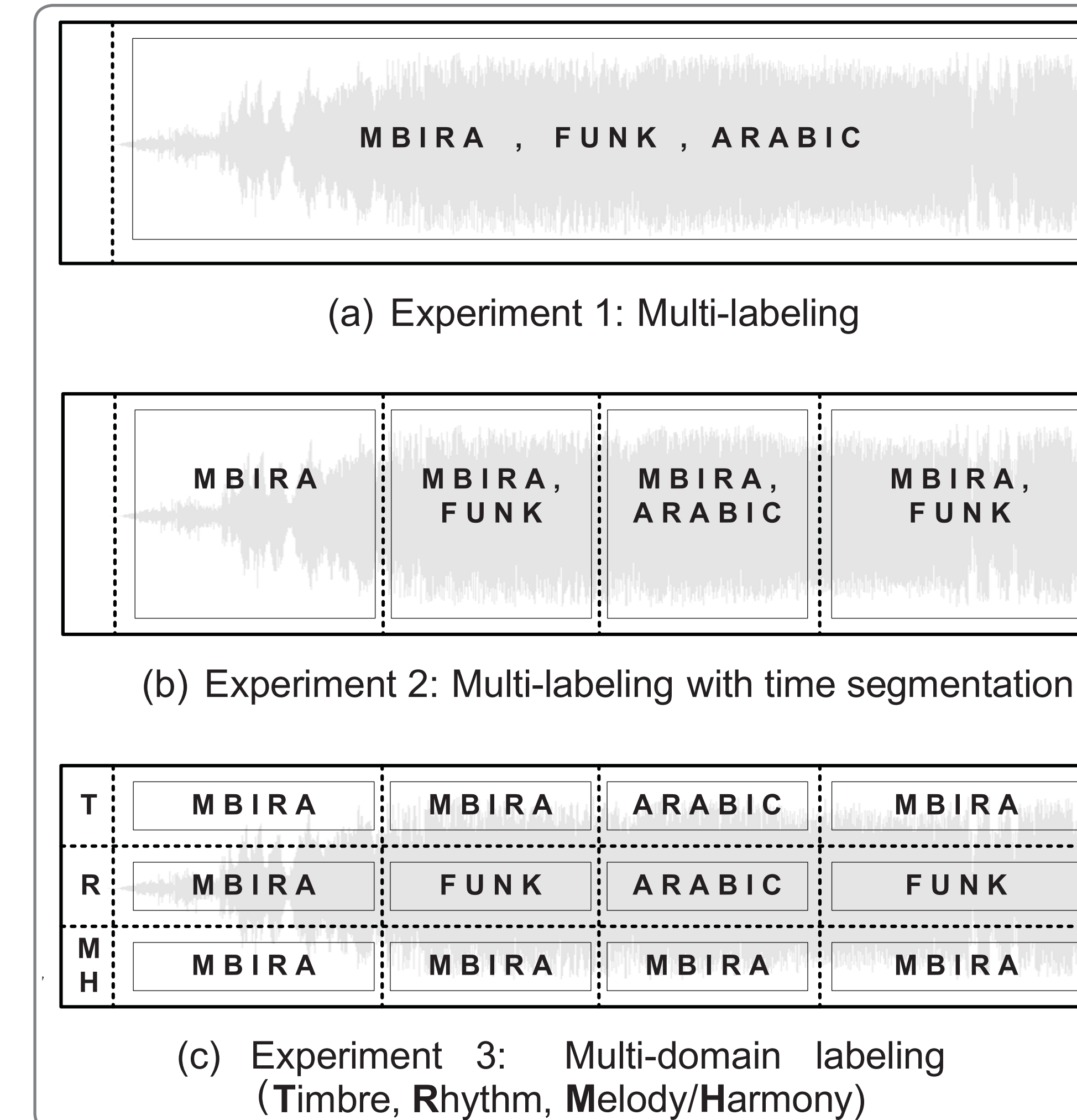
Features & Classifiers

Features:

- Besides low-level acoustic features, several mid-level representations are extracted
- Features are assigned to the musical domains
- **Timbre (233 dimensions):** Mel-Frequency Cepstral Coefficients, Audio Spectrum Centroid, Spectral Crest Factor, Spectral Flatness Measurement, Modulation Spectral Features
- **Rhythm (768 dimensions):** Audio Spectrum Envelope, Percussiveness, Envelope Cross-Correlation, derived from Auto Correlation Function, log-lag Auto Correlation Function, etc.
- **Melody/Harmony (187 dimensions):** Circular Pitch Space, features derived from Enhanced Pitch Class Profiles

Classifiers:

- Principal Component Analysis (PCA)
 - Reduction to 100 dimensions for each domain
- Gaussian Mixture Models (GMM)
 - Diagonal covariance matrix; 1, 5, 20 or 50 mixtures
- Post-processing via simple thresholding



Experimental Framework

Experiment 1: Multi-labeling (ML)

- Multiple labels (may) be assigned to each song
- No time segmentation available
- Set of binary GMMs, one model per genre

Experiment 2: ML + time segmentation (TS)

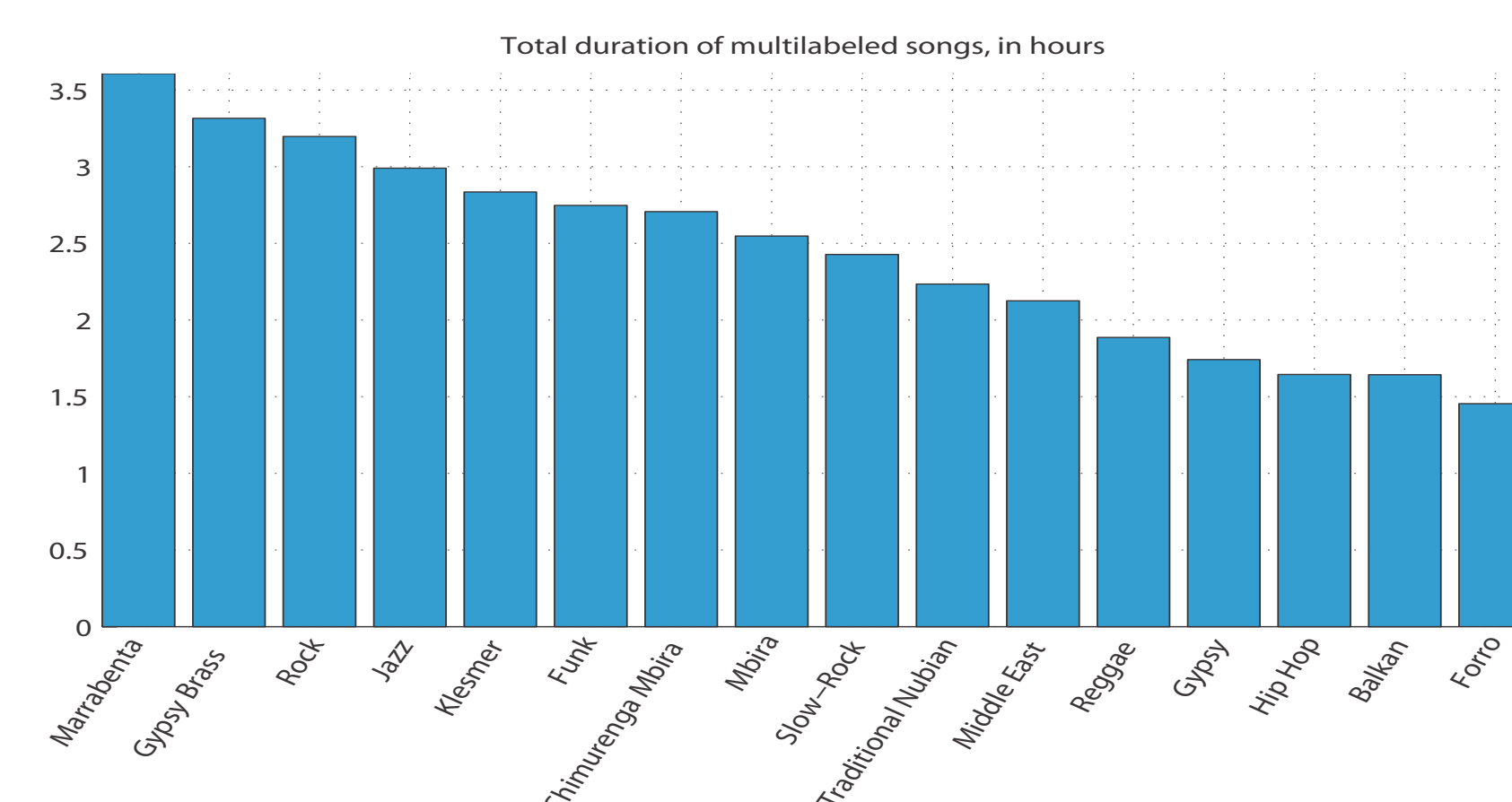
- Provide time segmentation
 - Set of labels is not changed within each time segment
- Multiple labels (may) be assigned to each segment
- Set of binary GMMs, one model per genre

Experiment 3: Multi-domain-labeling + TS

- Provide time segmentation
- Assign each label to one of musical domains
 - "This segment belongs to *Samba* because of *rhythmics*"
- Up to 4 genre labels for each time segment
 - Global, timbre-related, rhythm-related, tonality-related
- One "discriminative" GMM for each domain

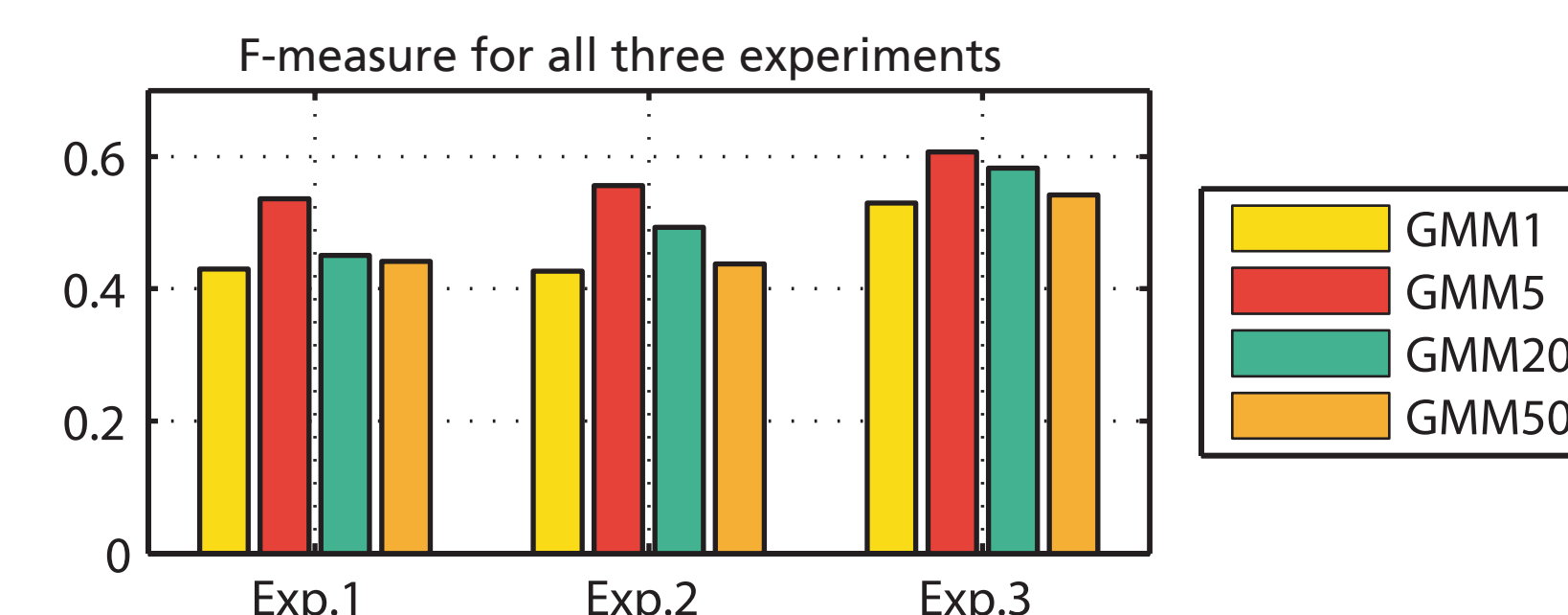
Database

- Music collection of 430 full-length tracks
- Flat taxonomy of 16 world music genres
 - Balkan, Gypsy, Forro, Klesmer, Mbira, Nubian, etc.
- Manual annotations by educated musicologists
- Follow the multi-domain-labeling principle
- Label cardinality of 1.1 to 2.0
- Split in training (70%) and test set (30%)



Results

- F-measure values for each number of mixtures in GMM are increased for Exp. 3 in comparison to Exp. 1 and Exp. 2.
- Optimal number of mixtures for all three experiments: 5 mixtures
- The best performance in Exp. 3 for GMM with 5 mixtures: F-measure of 0.61
- In Exp. 3 the classifier includes about two times fewer free parameters than in Exp. 1 and Exp. 2.



Conclusions & Future Work

Conclusions:

- Novel two-dimensional approach to music genre classification
- Decompose a multi-label classification problem by breaking it down in two dimensions
- Gain about 10% of F-measure!
- Extensible to other music genres
 - Partitioning into musical domains is universal

Future Work:

- Apply alternative classification techniques such as Support Vector Machines
 - Suppose to perform better for binary classification
- Make use of supervised feature selection and feature space transformation methods within each domain
- Add additional domains:
 - Vocals and instrumentation

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