

Learning Unsupervised Hierarchies Of Audio Concepts

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Introduction

Raw music signals are challenging to interpret
→ concept learning was proposed in *computer vision*

However, music has specificities not well handled:
→ high number of concepts
→ strong correlation of concepts (eg. Funk-Rock)

We propose a **novel task** more suited to MIR:

Learning and hierarchising concepts

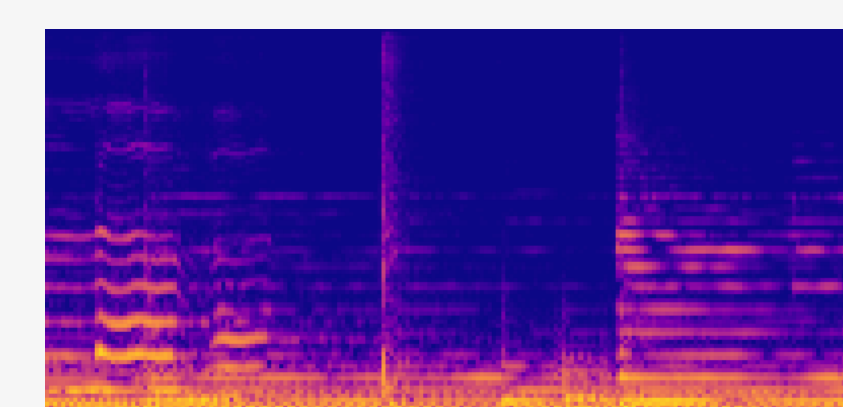
→ open the doors to new music descriptors (eg. folksonomy)
→ better understand how music is organised
→ solution to a new task in concept learning

Explainability goals

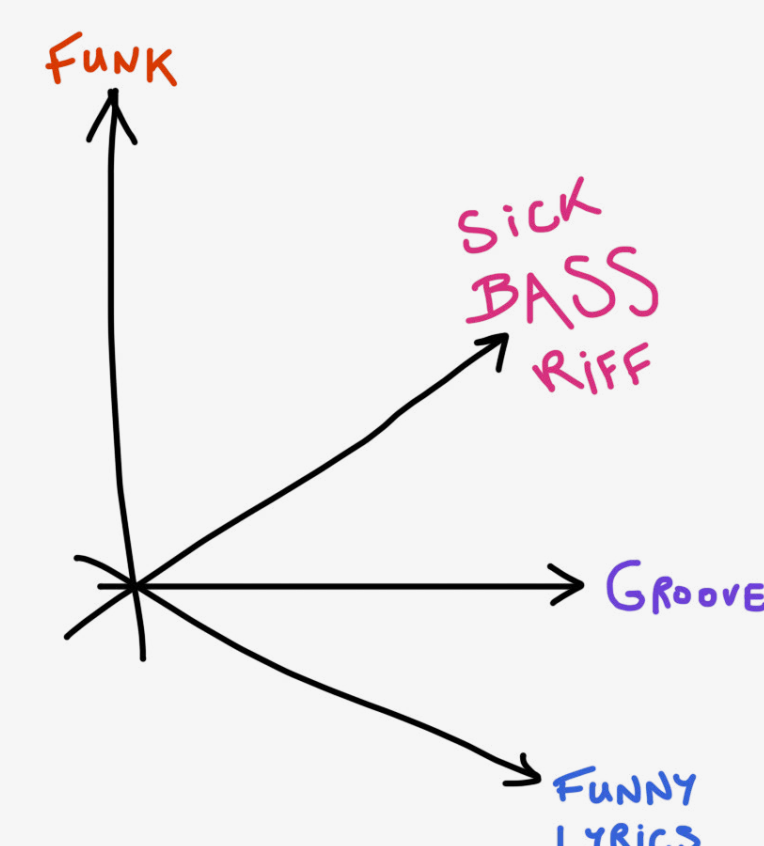
Informed by XAI papers, we define three explanation goals to ensure the relevance of our description:

- 1 Attribution**
Identify music concept within a music track
- 2 Transferability**
Invariance of concepts to setting and task
- 3 Generality**
Invariance of hierarchy, make sense in general

The general spirit of concept learning is to move from:



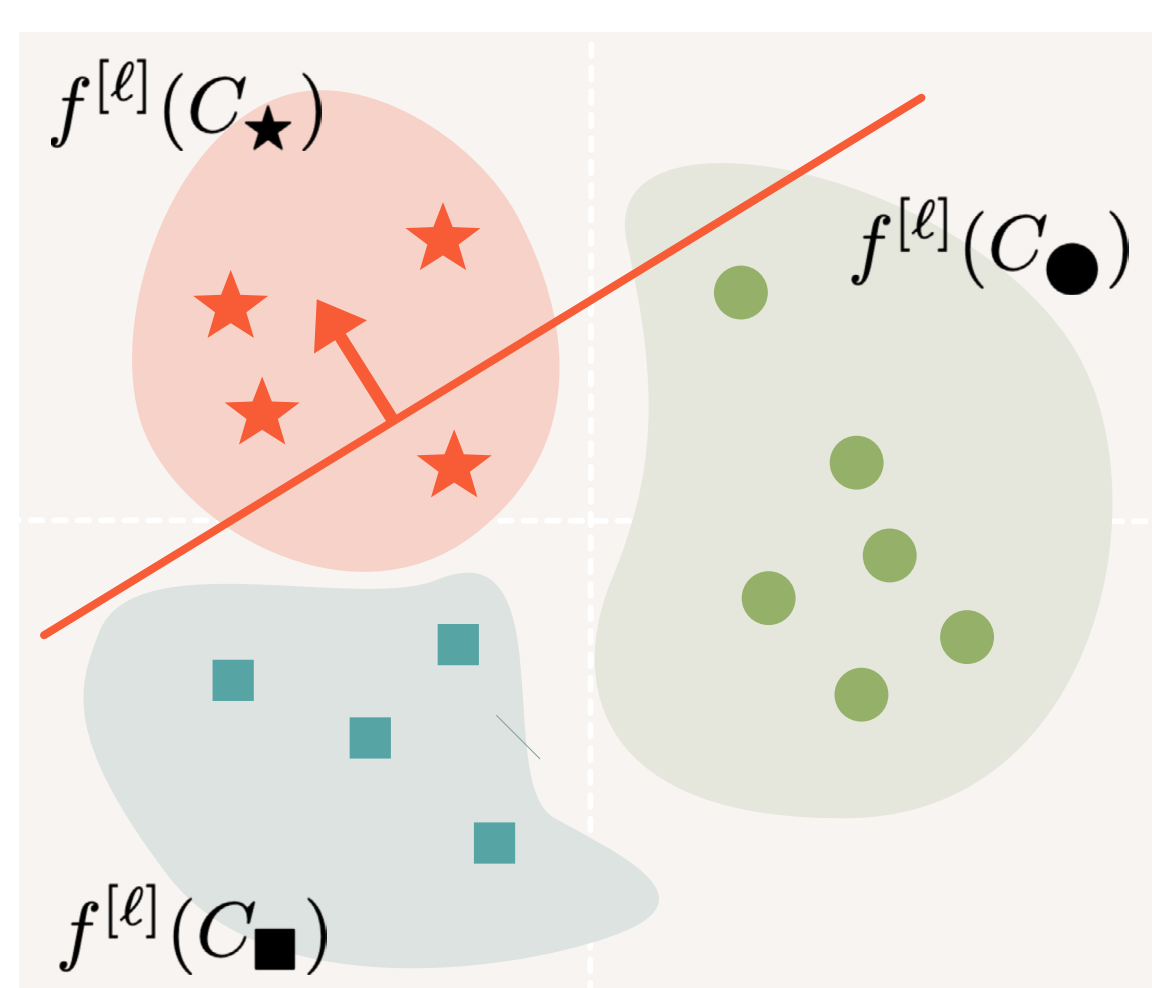
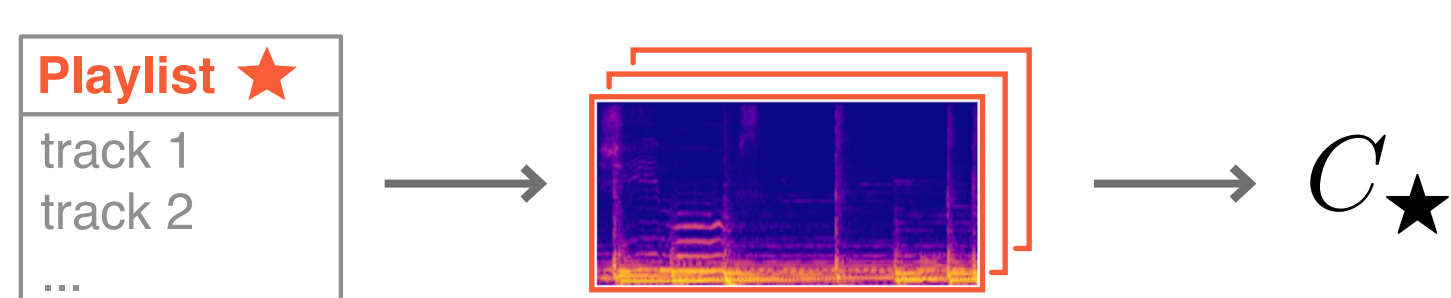
THUNDERCAT - THEM CHANGE



We explore this idea purely for audio signals
→ experiment with playlists as concepts
→ (co)-first paper on concept learning in MIR

Learning audio concepts

We **adapt TCAV** to music tracks spectrograms
→ consider playlists as source of concepts



→ we obtain a concept vector \vec{v}_\star

Background on TCAV [Kim2018]

→ was proposed in *computer vision*
→ we learn to linearly discriminate:

$$\{f^{[l]}(x) \mid x \in C\} \text{ versus } \{\dots \mid x \notin C\}$$

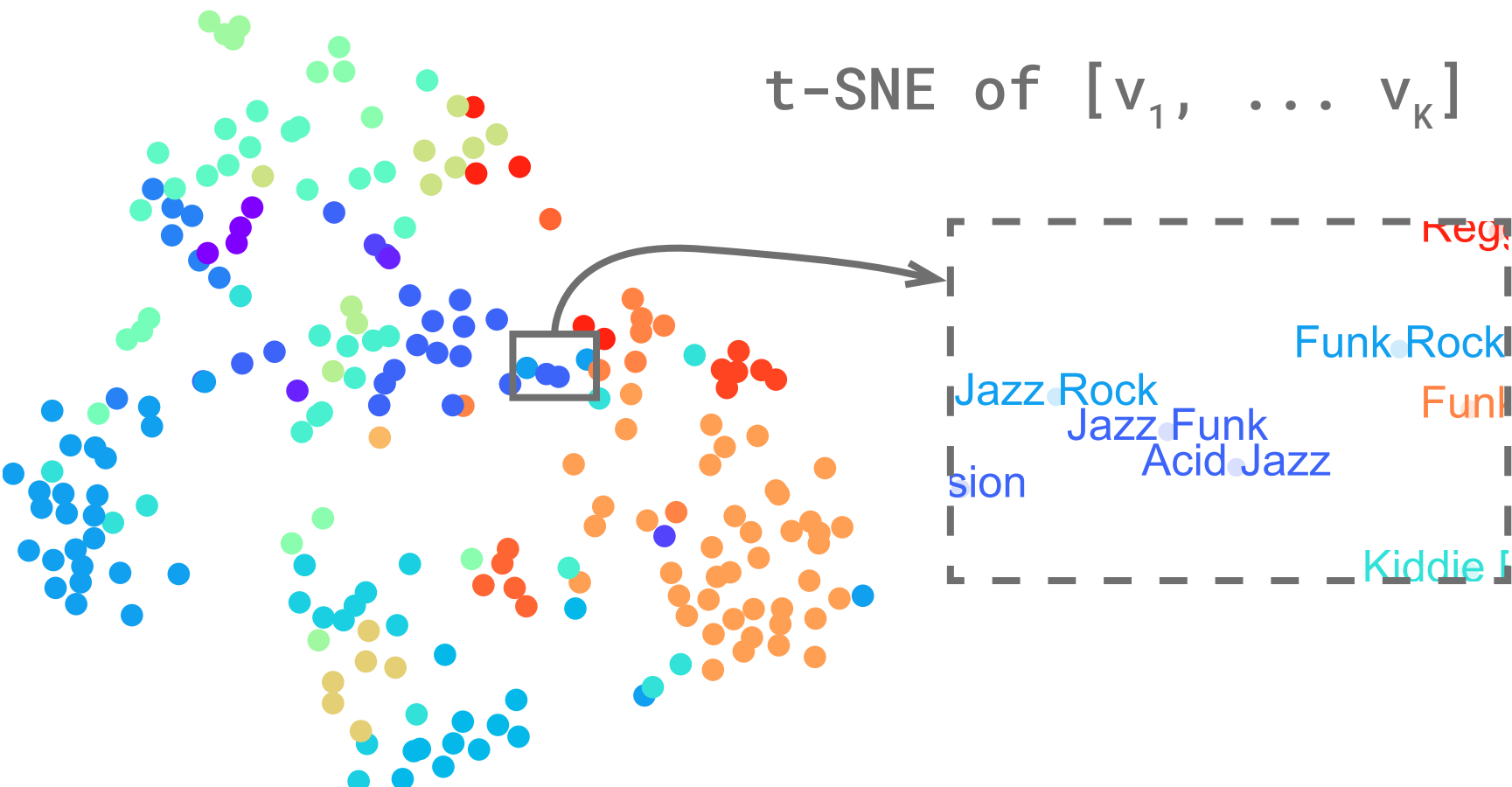
backbone few-shot examples

→ solved in practice with a logistic regression
→ learned parameter: *concept activation vector*

Hierarchising concepts

Useful observation: similarities make sense

→ vector similarity coincides with concept similarity



→ we propose a fast similarity estimation:

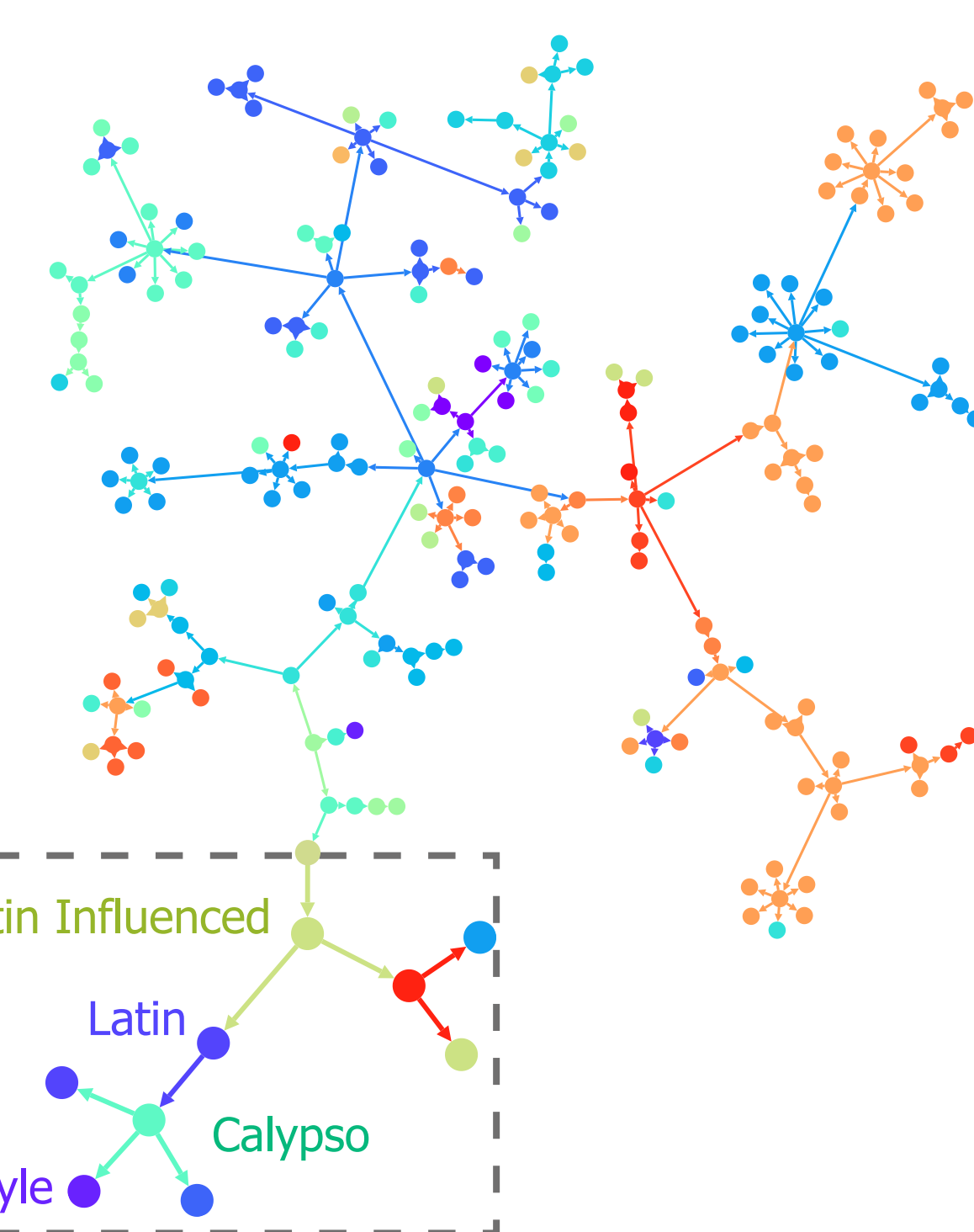
$$S_{i,j}(\mathcal{V}) = \mathbb{E}_{x \sim \mathcal{C}_i} [\sigma(\langle \vec{v}_j, \vec{f}^{[l]}(x) \rangle)]$$

Useful algorithm: centrality-based hierarchy

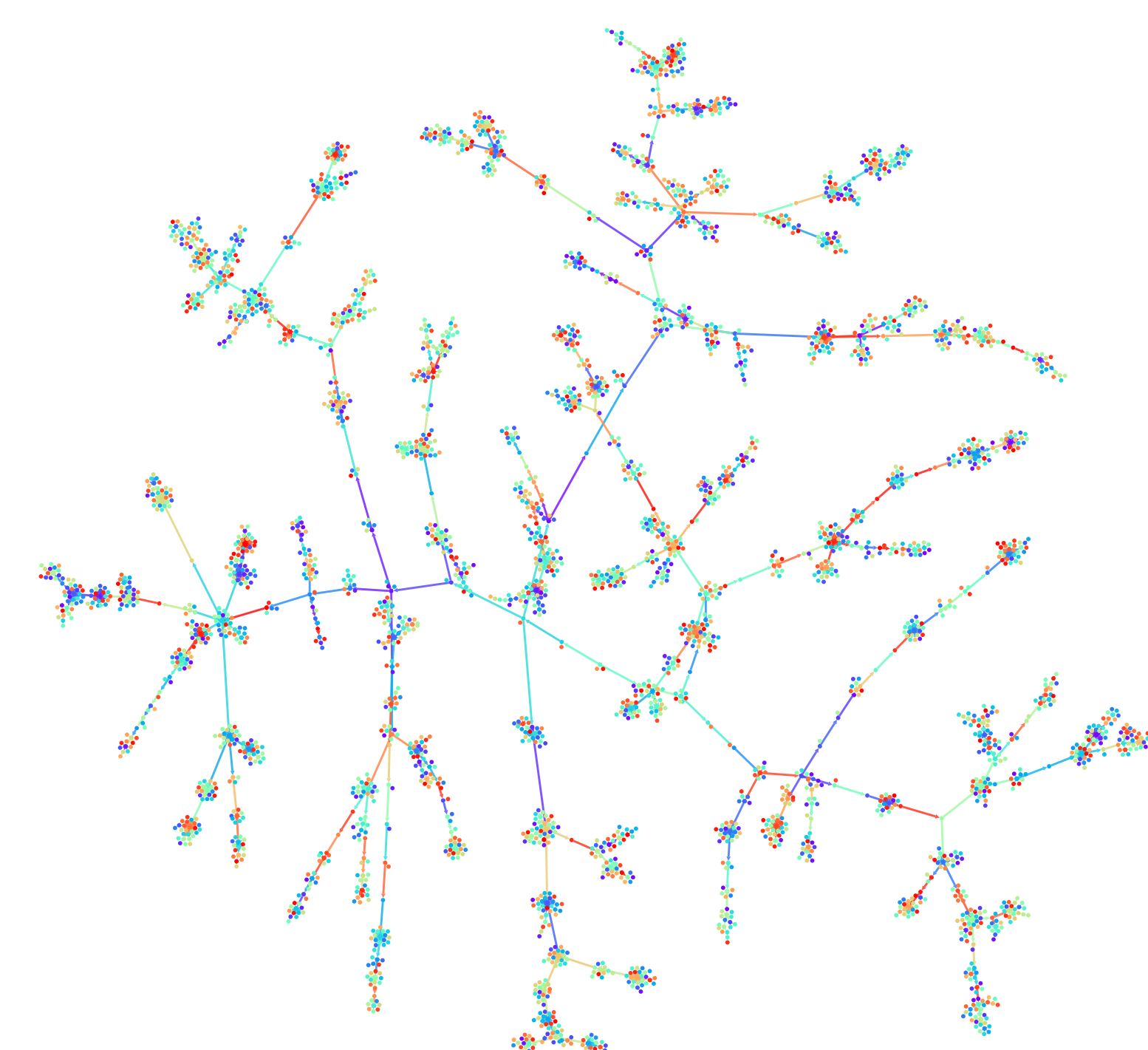
→ proposed in NLP to folksonomies from tags
→ we show that the hypotheses work for music

Betweenness hierarchy [Heymann2006]

- * INPUT: similarity graph S
- * compute betweenness-centrality of S
- * for nodes of S ordered by centrality:
- * add new node to hierarchy H
- * edge to most similar node in H
- * OUTPUT: tree H



Resulting hierarchy of APM's genres



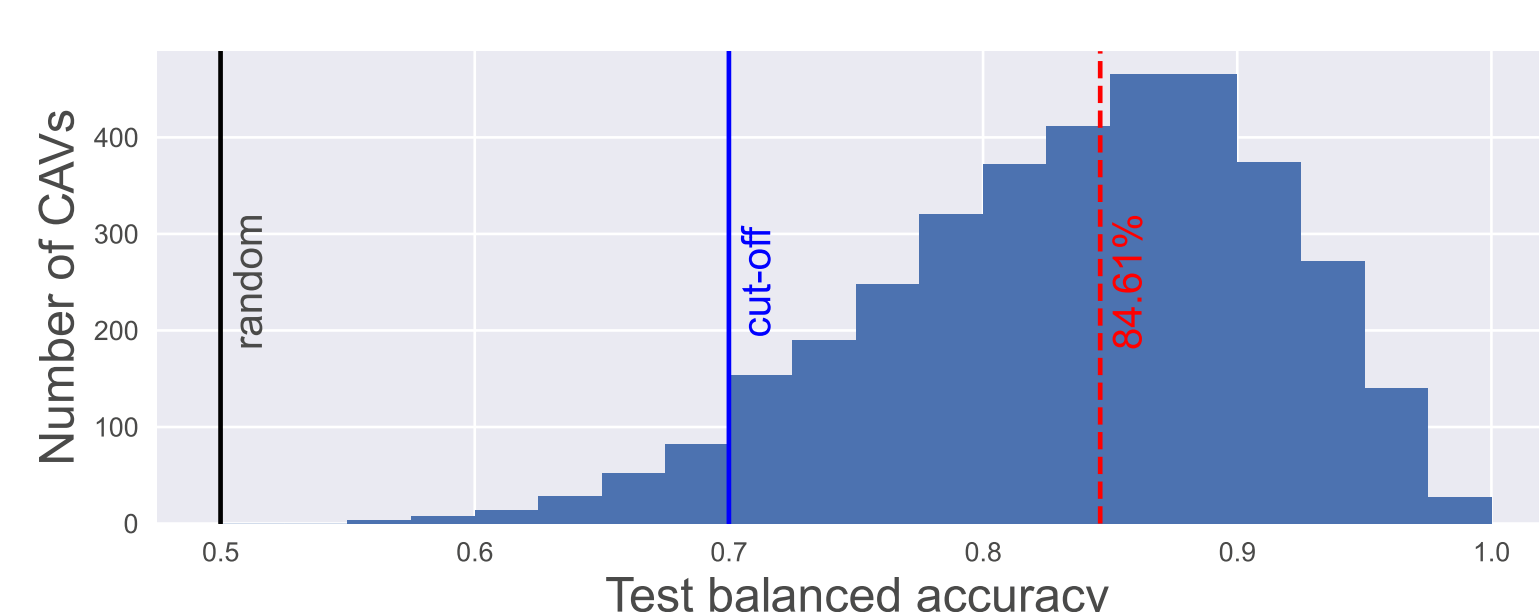
Resulting hierarchy of 3443 Deezer playlists

Experiments

Leverage three datasets:

→ APM Music's genres (219)
→ APM Music's moods (165)
→ Deezer playlists (3443) ie. 3443³⁴⁴¹ trees

1 Attribution



2

Transferability

→ related to the design, choice of backbone, concepts

3

Generality

→ discuss and analysis of priors and expected hierarchy
→ ground-truth evaluation for genre and mood
→ proxy evaluation for concept of mixed types
→ quantitative discussion

Source	Audio (↓)	CF (↓)	BERT (↑)	W2V ₁ (↑)	W2V ₂ (↑)
$H(S(V_D))$	2.449 ± 0.022	0.845 ± 0.013	0.345 ± 0.007	0.286 ± 0.009	0.542 ± 0.007
$H(S_{CF})$	2.413 ± 0.021	0.345 ± 0.007	0.416 ± 0.008	0.336 ± 0.010	0.601 ± 0.008
$H(S_{BERT})$	2.858 ± 0.028	0.868 ± 0.013	0.726 ± 0.005	0.505 ± 0.011	0.652 ± 0.008
$H(S_{W2V_1})$	2.952 ± 0.028	0.932 ± 0.012	0.523 ± 0.008	0.804 ± 0.005	0.721 ± 0.007
$H(S_{W2V_2})$	2.843 ± 0.026	0.847 ± 0.012	0.531 ± 0.008	0.596 ± 0.009	0.836 ± 0.004
Random	3.388 ± 0.027	1.104 ± 0.006	0.239 ± 0.004	0.142 ± 0.006	0.452 ± 0.006

Gist of the table: the hierarchy found from audio compared to one we can find through collaborative filtering or specialised w2v.

Take-away

→ First paper of concept learning on spectrograms
→ Novel idea to hierarchise learned concepts
→ Framed through XAI to be as relevant as possible
→ Complementary to expert ontologies, our method enables seeing how music is organically described by users and editors

Future work

→ Better performances, curb biases, save humanity
→ Dynamic recommendations
→ Play with concepts evolution through time

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