

Research Question: "Can visual analytics reveal the hidden evolution of musical genres beyond marketing labels?"

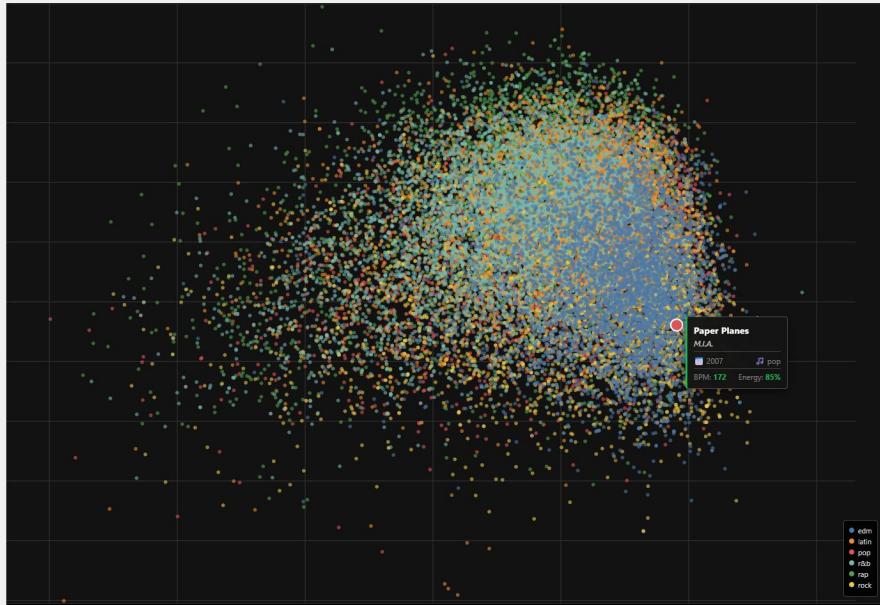
# Emotional and Structural Evolution of Modern Music (1970-2020)

# Data & Preprocessing Pipeline

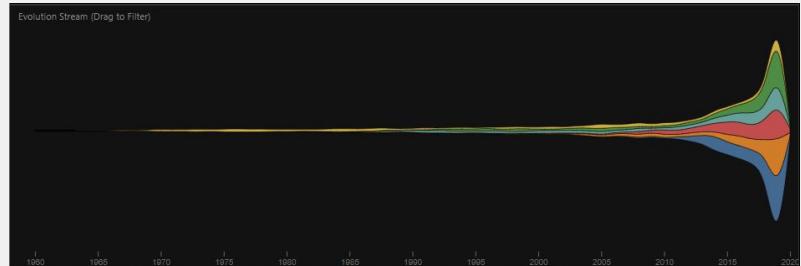
Raw Data	Cleaning	Normalization	DR (PCA) & Clustering (K-Means)
01 <b>Spotify Tracks Dataset :</b> <ul style="list-style-type: none"><li>• 18 dimensions</li><li>• 100k Tracks</li></ul>	02 9 numerical audio features are selected.	03 Applying a <b>Z-Score normalization</b> to ensure they contribute equally. $z = \frac{x - \mu}{\sigma}$	04 <b>PCA :</b> <ul style="list-style-type: none"><li>• 9 dimensions into 2D model</li></ul> <b>K-Means clustering :</b> <ul style="list-style-type: none"><li>• Find natural groups of songs based on their actual sound, not just their genre.</li></ul>

# Visual Design & Interaction

*"Overview first, then details-on-demand"* - Shneiderman's Mantra



1. The Main View : Scatterplot



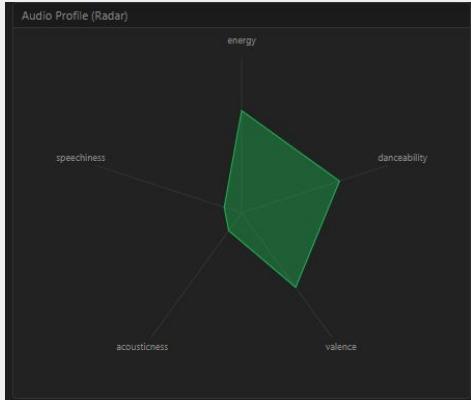
3. Temporal View



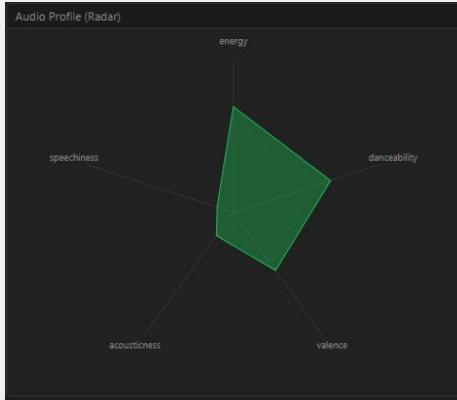
2. Analytics Panel

# Insights (Results)

## Insight 1: "The Sadness of Modern Pop"



1990 - 1999

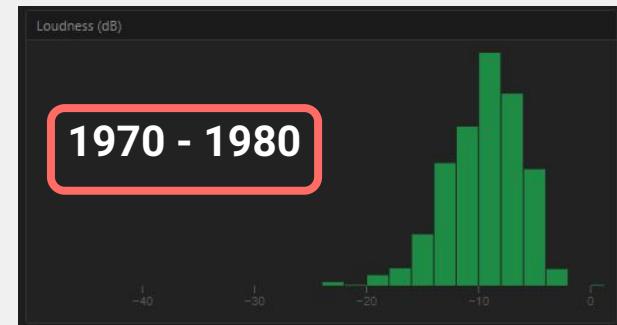


2015 - 2020

"While Pop music became louder, it also became significantly sadder."

## Insight 1: "Loudness War"

1970s VS 2010s, shift in the loudness distribution towards higher decibels : Loss of Dynamic Range



1970 - 1980



2010 - 2020

# Conclusion & Limitations

- ✓ Successful dimensionality reduction (PCA).
- ✓ Real-time interaction (no latency).
- ✗ Limitation: Subjectivity of genre labels.

Thank you for your attention !

Next : DEMO TIME