

# Rhythms Through Time: A Data-Driven Analysis of Lyrics and Audio Features

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## 1. Background

This study analyzes the relationship between audio features and lyrical content using extensive Spotify datasets. By integrating machine learning we aim to make a user tailored recommender system. Additionally, we explore semantic meanings in song lyrics and develop an improved recommender system. This research aims to prototype a recommender system that is more personalized than those in music apps.

## 2. Motivations & Objectives

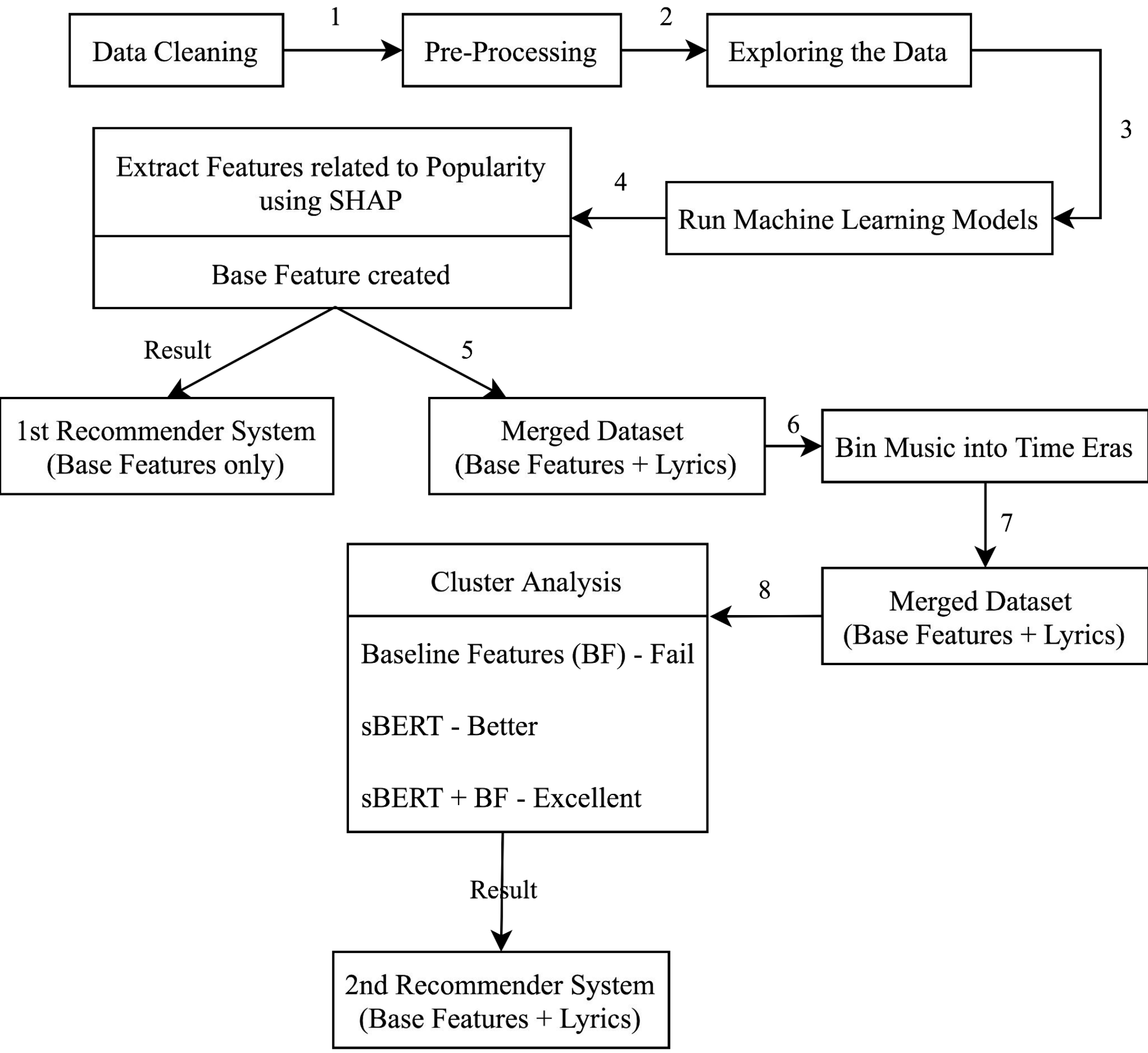
This research has two core objectives: (1) Creating a recommender system tailored to user preferences and (2) analyzing lyrical meanings within lyrics using the sBERT deep learning model. The study motivations include: (a) enhancing user engagement and revenue by increasing user satisfaction (b) create monetization opportunities through targeted advertising from user preferences

## 3. Methods

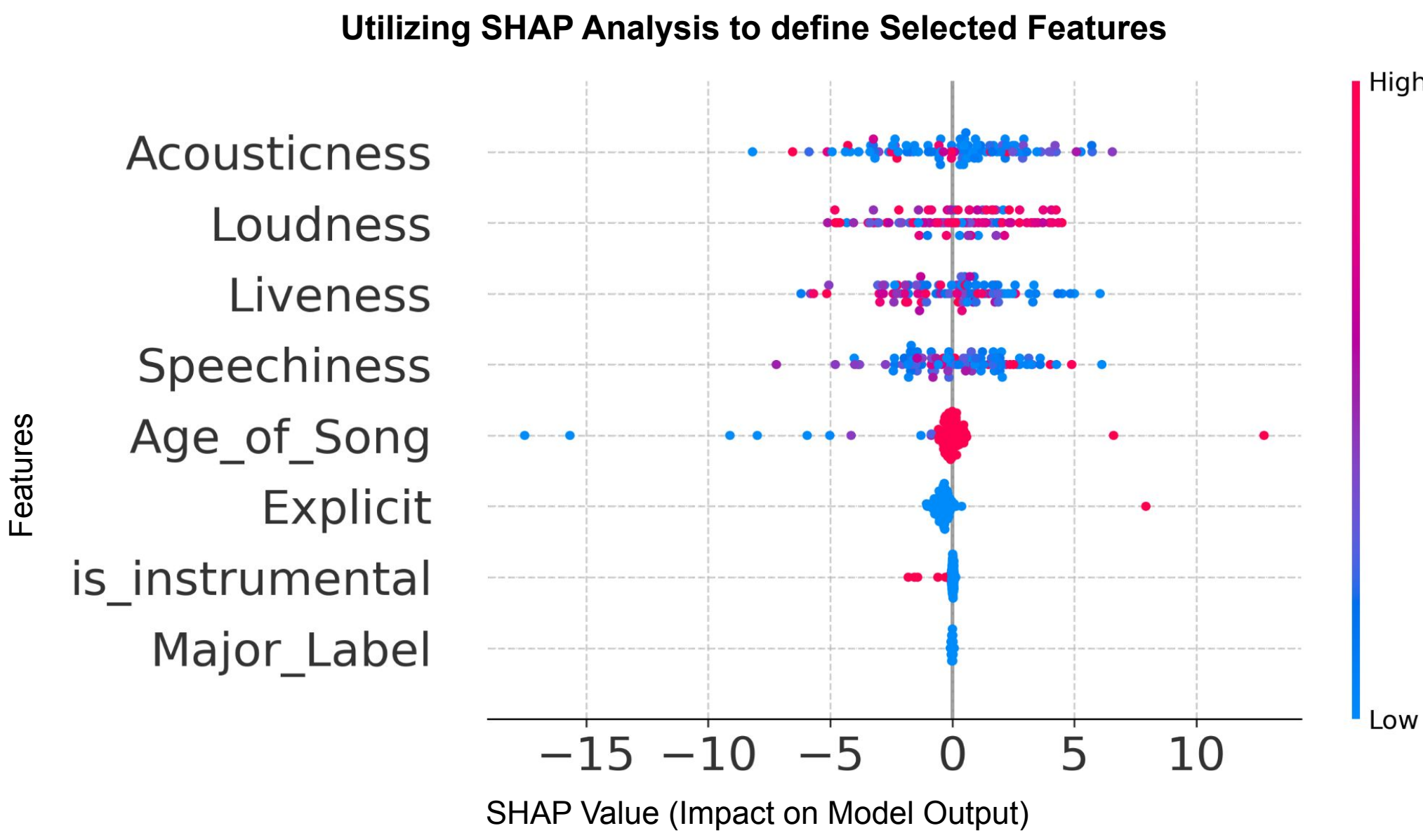
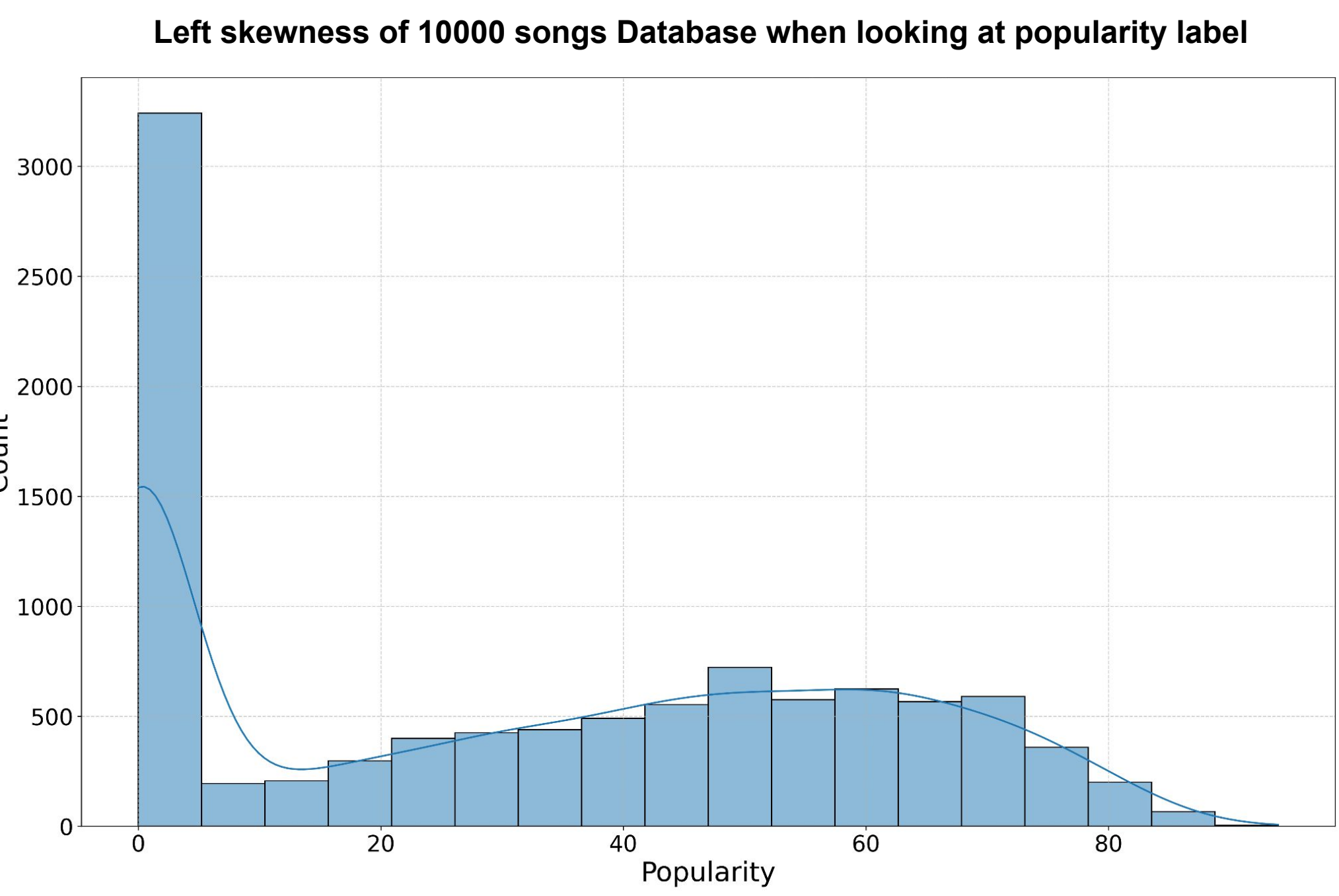
The study utilized two comprehensive Spotify datasets: one comprising 10,000 songs with audio features and another featuring 57,651 songs with lyrics. Models are ran on data here to predict popularity and the related features are captured using SHAP analysis. These based features are used for the 1st recommender.

The merged dataset required extensive preprocessing and after cleaning, the final dataset contained 1,634 entries, blending audio features and lyrics. sBERT is ran on this dataset and clustering analysis is used to determined the best combinations of features. These features are used for the 2nd recommender.

Below is a flow chart of our processes and data flow.



## 4. Data



## 4. Experimental Result

Regression models on average achieved an  $R^2$  of  $\sim 0.5$ , explaining about 50% of popularity variance. Clustering improved in selected features enriched with sBERT embeddings.

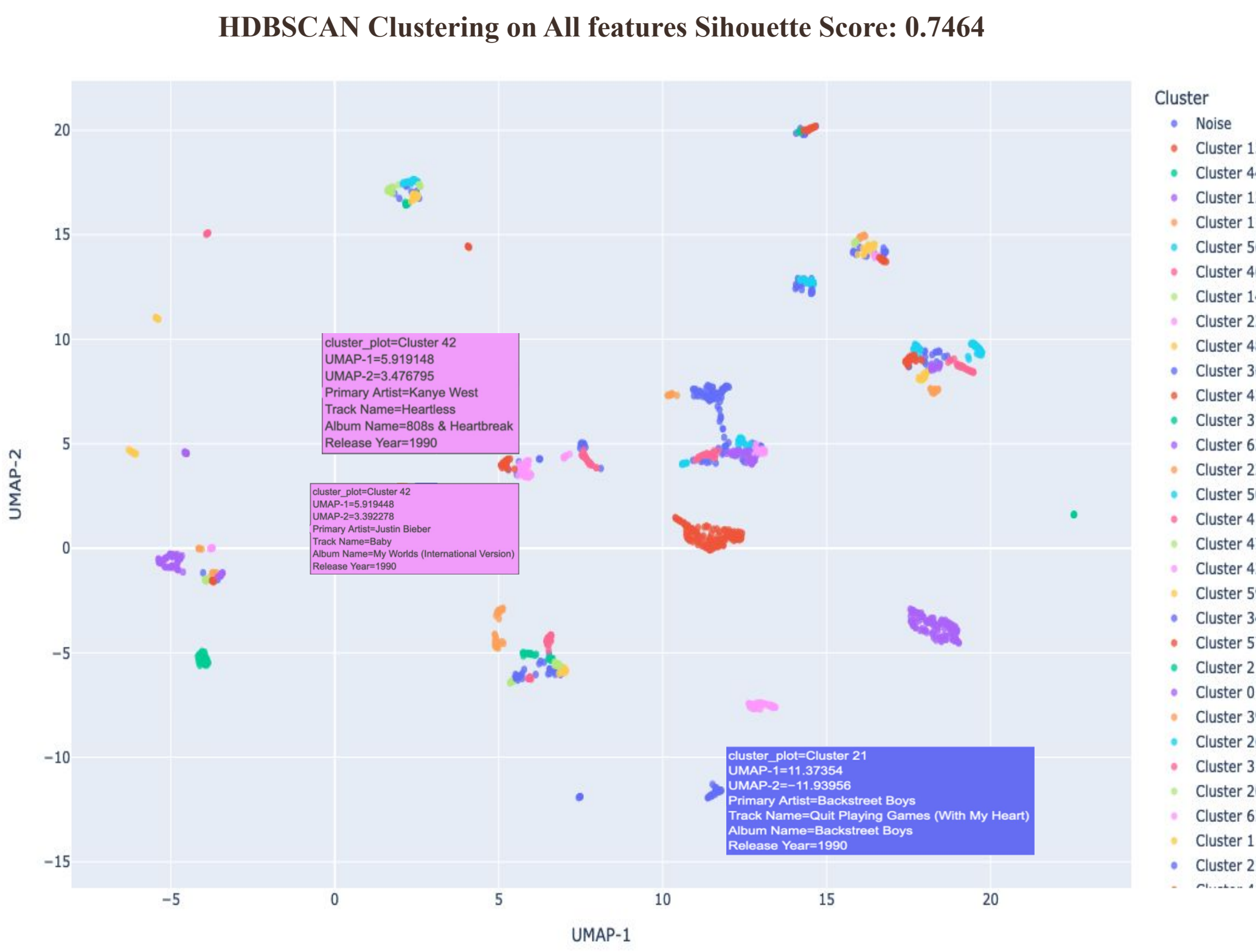
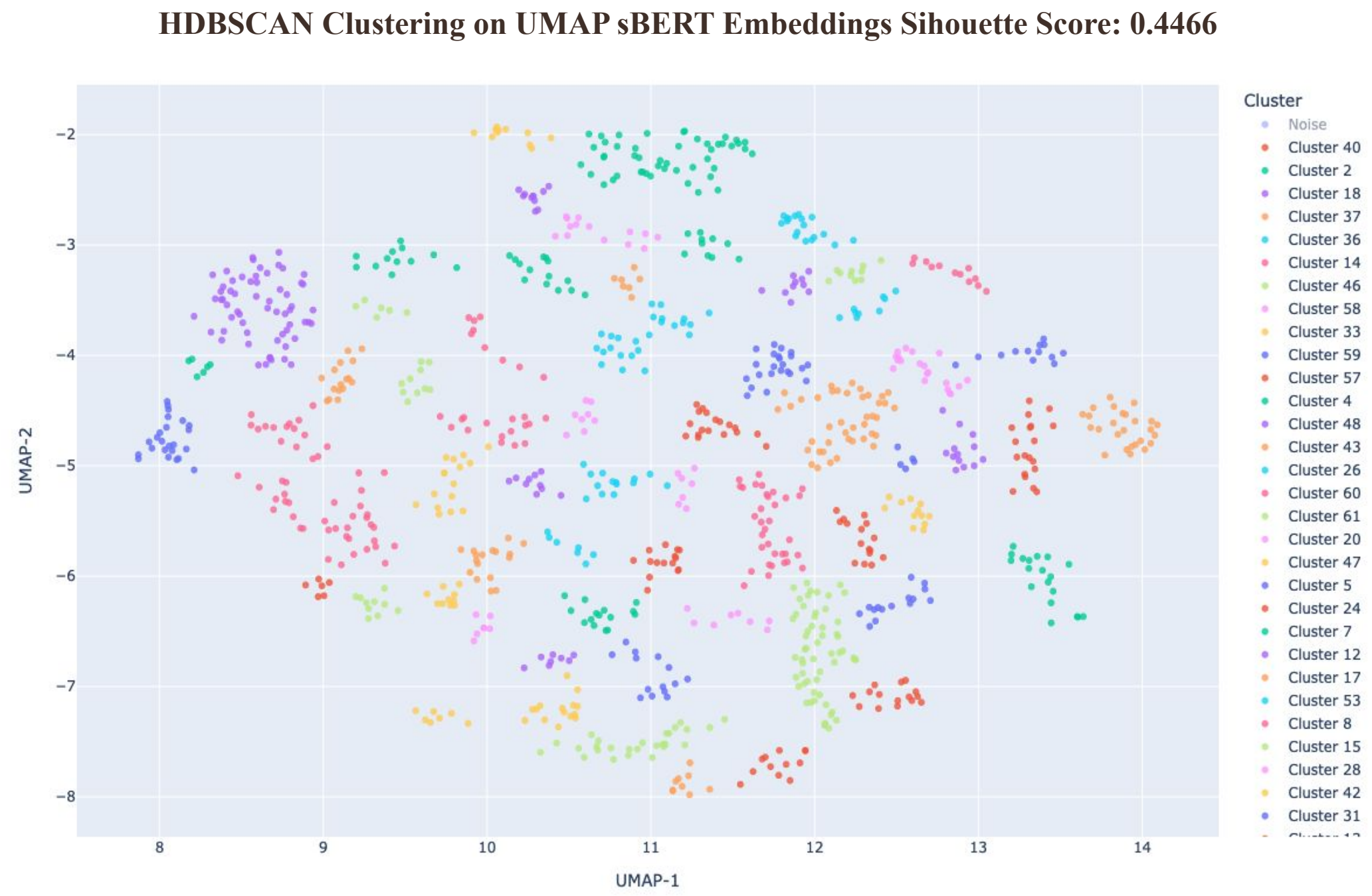
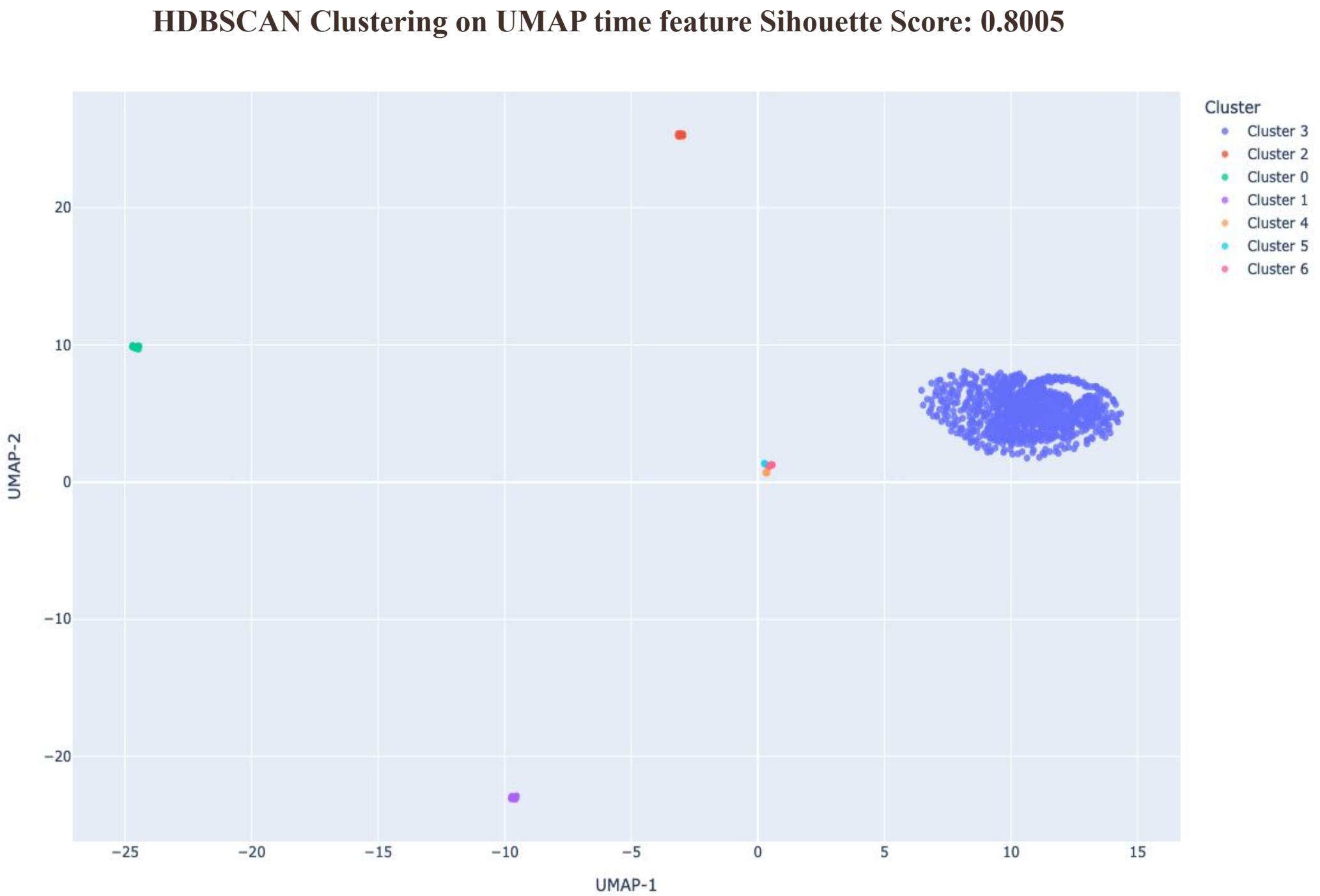
Table 1: Regression Model Performance Metrics

Model	MSE	R <sup>2</sup>
Linear Regression (Including Artist_Popularity)	368.78	0.512
Ridge Regression (Including Artist_Popularity)	368.77	0.512
Lasso Regression (Including Artist_Popularity)	368.76	0.512
Random Forest Regressor (Including Artist_Popularity)	473.25	0.373
Gradient Boosting Regressor (Including Artist_Popularity)	380.86	0.496
Keras Sequential Model (Including Artist_Popularity)	444.00	0.412
Tuned XGBoost Regressor (Including Artist_Popularity)	377.18	0.501
Tuned LightGBM Regressor (Including Artist_Popularity)	389.92	0.484
Baseline Linear Regression (Joe Beach Capital, 2023)	883.80	-0.0053
Baseline Gradient Boosting (Joe Beach Capital, 2023)	881.82	-0.0021
Baseline Random Forest (Joe Beach Capital, 2023)	944.95	-0.8748

Table 2: Clustering Model Performance Metrics

Model	Silhouette Score
Benchmark Model (Time Period Bin Only)	0.8084 (Bad Clusters)
sBERT Model (Time Period Bin Only)	0.4286 (Decent Clusters)
sBERT Enhanced Model (Time Period Bin + Selected Features)	0.7464 (Excellent Clusters)

Clustering progression visualizations of time binning baseline, baseline + sbert, and baseline + sbert + selected features.



The first recommender system utilized only the selected features. The second recommender system utilizes sBERT, time binning, and the selected features. These features are shown in the final HDBSCAN clustering results. Very good cluster with Silhouette Score of 0.7464.

## 5. Key Learnings and Discussions

Word Cloud analysis revealed trends in lyrical themes of expressions across the decades. Shap analysis selected good features for a baseline recommender system. sBERT embedding maps the cultural relationship between music/musicians. These features improved our recommender system from one that recommended songs of similar “mood” to one that capture both “mood” and “semantic meaning”. The integration of audio features and lyrical semantics lead to a personalized recommender system.



Base Features Recommender System

Enter the name of the song (required): juda  
Enter the artist name (optional): layd gag  
Enter the number of recommendations you want (default 10): 15

Matched Song: 'Judas' with score 89  
Matched Artist: 'Lady Gaga' with score 82

Using songs by 'Lady Gaga' titled 'Judas' for recommendations.

Recommended Songs (Top 15):  
- 'Somebody to Love Me' by Mark Ronson, The Business Intl from the album 'Record Collection'  
- 'Doing It (feat. Rita Ora)' by Charli xcx, Rita Ora from the album 'SUCKER'  
- 'Rain On Me (with Ariana Grande)' by Lady Gaga, Ariana Grande from the album 'Chromatica'  
- 'Rain On Me (with Ariana Grande)' by Lady Gaga, Ariana Grande from the album 'Rain On Me (i  
- 'Harlem' by New Politics from the album 'A Bad Girl In Harlem'

Refined Features Recommender System

Enter the name of the song (required): juda  
Enter the artist name (optional): layd gag  
Enter the number of recommendations you want (default 10): 15

Matched Song: 'Judas' with score 89  
Matched Artist: 'Lady Gaga' with score 82

Using songs by 'Lady Gaga' titled 'Judas' for recommendations.

Recommended Songs (Top 15):  
- 'The Edge of Heaven' by Wham! from the album 'Music From The Edge Of Heaven'  
- 'When You Were Young' by The Killers from the album 'Sam's Town'  
- 'Original Sin' by INXS from the album 'INXS Remastered'  
- 'Stronger' by Kanye West from the album 'Graduation'  
- 'Sorry' by Justin Bieber from the album 'Sorry'  
- 'Devil Inside' by INXS from the album 'Kick 25 (Deluxe Edition)'  
- 'Karma Chameleon - Remastered 2002' by Culture Club from the album 'Colour By Numbers (R

## 6. Future Work

We will focus on making different types of recommender systems that have bias towards genres, artists, or specific song feature. Our study limitations lies our merged dataset having around 2000 songs. We can revisit the merging process to incorporate more lyrics into our original 10,000-song dataset.

