

# *Title: Amazon Music Clustering — Genre & Mood Segmentation*

**by Kanish Midhun K**

# **Objective:**

To group songs from Amazon Music into meaningful categories (genres or moods) using **unsupervised machine learning**.

## **Tools & Libraries:**

Python, Pandas, Scikit-learn, Seaborn, Matplotlib, Streamlit

## **Key Techniques:**

EDA · Feature Scaling · K-Means Clustering · PCA · Cluster Profiling · Visualization

# Problem Statement

With millions of songs on streaming platforms, manually tagging genres or moods is not scalable.

## Goal:

Use audio features (danceability, energy, loudness, tempo, etc.) to automatically cluster songs that *sound alike*.

## Applications:

- Personalized playlist generation 
- Mood-based recommendation systems 
- Artist trend analysis 

# Dataset Overview

**Dataset:** single\_genre\_artists.csv

**Features Used:**

- danceability, energy, loudness, speechiness, acousticness
- instrumentalness, liveness, valence, tempo.

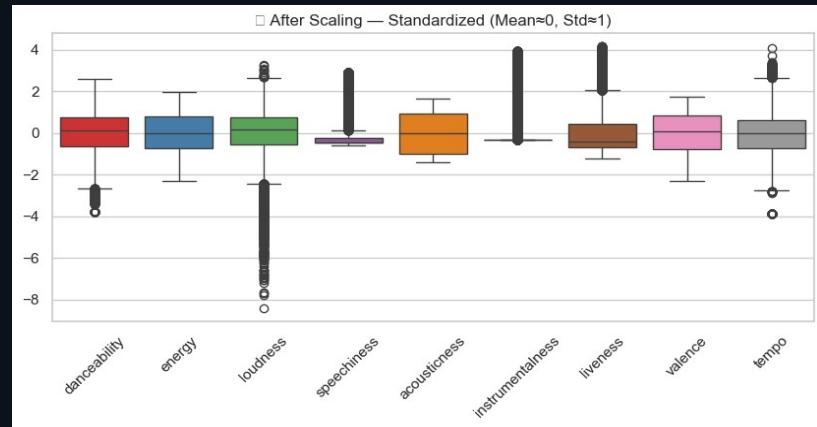
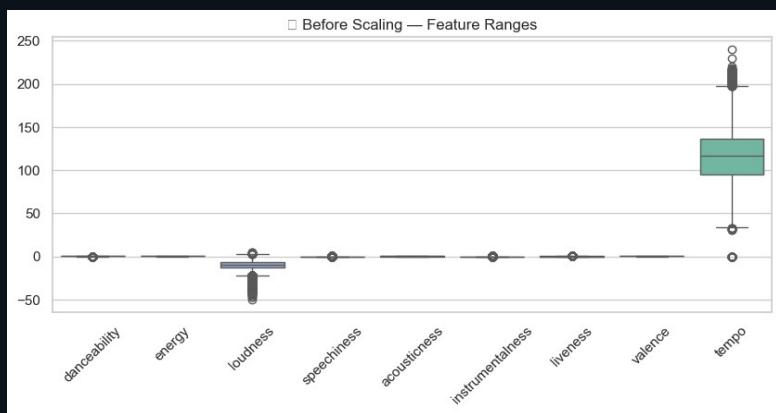
**Removed Columns:**

- track\_id, track\_name, artist\_name (used later for mapping results)

# Data Preprocessing

## Steps Performed:

1. Handled missing values (if any)
2. Selected relevant features for clustering
3. Applied StandardScaler — standardized all features to mean=0, std=1
4. Visualized scaling using before/after boxplots



# K-Means Clustering

## Why K-Means?

- Simple and effective for numerical features
- Finds centroids based on Euclidean distance

## Optimization:

Tested  $k = 2 \rightarrow 10$

## Evaluated using:

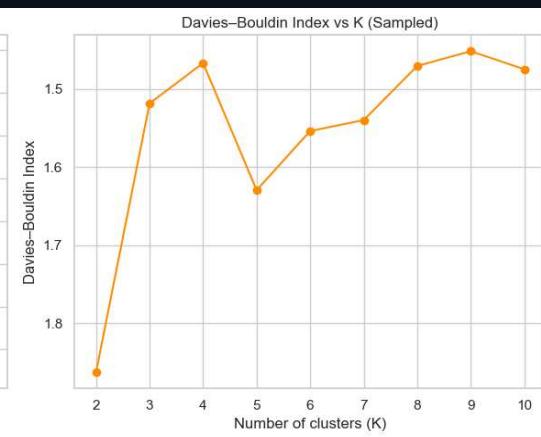
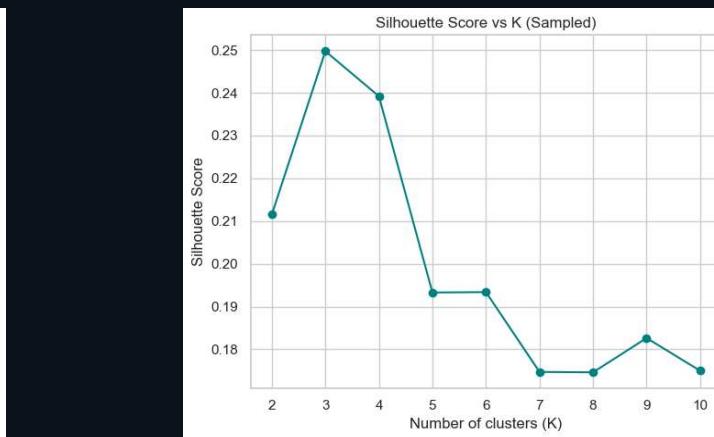
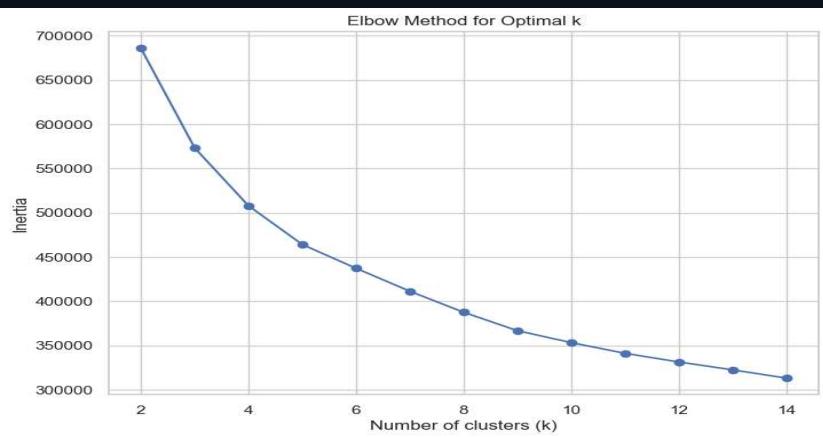
- Silhouette Score ( $\uparrow$  better)
- Davies–Bouldin Index ( $\downarrow$  better)
- Inertia (Elbow Method)

# Final Model ( $k = 3$ )

Model: KMeans(`n_clusters=3, random_state=42`)

## Interpretation:

Moderate separation — clear mood-based clusters form



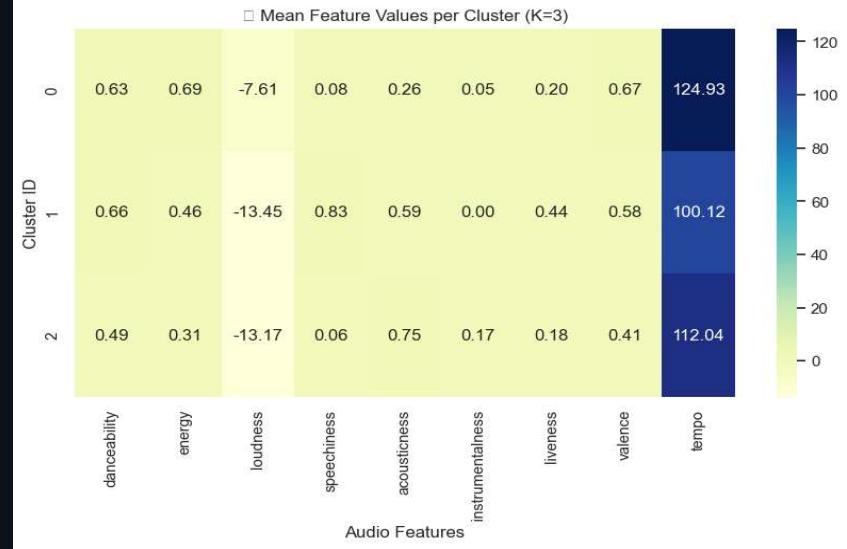
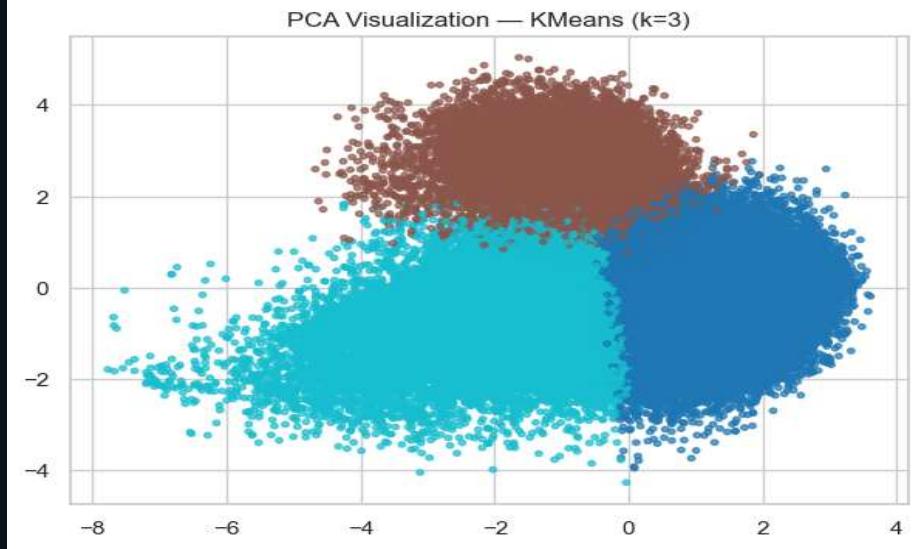
# Cluster Profiles

Cluster	Description	Key Traits
0	<b>Party / Upbeat</b> 🎉	High energy, high danceability, loud
1	<b>Chill Acoustic</b> 🌙	High acousticness, low energy
2	<b>Instrumental / Ambient</b> 🎵	High instrumentalness, low speechiness

# Visualization

## Techniques Used:

- PCA 2D Scatter Plot — visualize cluster separation
- Feature Heatmap — compare mean feature values per cluster



## Key Takeaways

-  Successful mood-based clustering of songs
-  3 natural groupings: *Upbeat, Chill, Instrumental*
-  Combined ML + Music Analytics + Visualization
-  Ready for playlist recommendation integration