



Song Recommendation System

Machine Learning for Global EDM Playlists

Clustering and Content-Based Song Recommendation

Lora Chuaner Ding

Date: 04.07.2024

Project Brief



1

Data Preparation – Spotify API

2

Feature Engineering

3

Machine Learning Model – Clustering

4

Model Evaluation and Visualization

5

Model Optimization – PCA&ISOMAP

6

Limitations and Learnings

Spotify API



Step1

Authenticate with Spotify API → Spotify Developer → **Client Key & Secret**



Step2

Fetch Playlists → Search and fetch to Country **public playlists**

```
# 3.1 functions search playlist
def search_playlists(query, country, limit=50):

# 3.2 Country playlists
def get_playlists_for_countries(query, countries,

# 4.2 Get track features
def get_audio_features_with_info(track_ids):

# 4.3 Get all features in playlists
def get_playlist_audio_features(playlist_id):
```

Step3

Extract Tracks → Extract **tracks and audio features** from playlists

Step4

Store Data → **Loop step 3** for all playlists and save the data

** Playlists are up-to-date*

```
Getting audio features for playlist: EDM Music 2024 (Top 100)
Getting audio features for playlist: mint
Getting audio features for playlist: EDM House Mix
Getting audio features for playlist: EDM BANGERS 🎵🎵
Getting audio features for playlist: Chill Tracks
Getting audio features for playlist: EDM Hard Bass Mix
Getting audio features for playlist: EDM Classics (Top 100)
Getting audio features for playlist: Workout EDM Mix
Getting audio features for playlist: Summer EDM Mix
Getting audio features for playlist: Happy Beats
Getting audio features for playlist: Hot girl EDM workout mix 🎵
Getting audio features for playlist: Pop Hits 2000s - 2024
Getting audio features for playlist: Club EDM Mix
Getting audio features for playlist: EDM BANGERS 🎵🎵
Getting audio features for playlist: EDM 2024
Getting audio features for playlist: EDM MIX 2024 🎵 EDM HITS 🎵 TOMORROWLAND 2024
Getting audio features for playlist: Ultra Gaming
Getting audio features for playlist: Best EDM Songs of All Time - Most Popular EDM Music Playlist
Getting audio features for playlist: BRAZILIAN PHONK 🎵
Getting audio features for playlist: Best EDM Playlist 2024
Getting audio features for playlist: Deep House Relax
Getting audio features for playlist: 2013-2015 EDM Hits
Getting audio features for playlist: Dance Hits
Getting audio features for playlist: Hype EDM Mix
Getting audio features for playlist: Dance Party
Getting audio features for playlist: EDM Remixes of Popular Songs
Getting audio features for playlist: Running EDM Mix
Getting audio features for playlist: EDM Workout 2024 | Training Hits
Getting audio features for playlist: frat songs that could resurrect me
Getting audio features for playlist: EDM Hits
Getting audio features for playlist: Hype
Getting audio features for playlist: EDM Pop Mix
Getting audio features for playlist: Can Music
Getting audio features for playlist: as i run (to EDM hits)
```

Spotify Recommendation Methods



Collaborative Filtering

Principle

- Find users with similar music tastes to you
- Find songs similar to the ones you like.

Data

Private user-Item Interaction Data

Data Security



Content-Based Recommendation

Recommend songs that are similar in features

Song Features Data

Public Playlists



Data Preparation



Dataset

Top 50 EDM Playlists from Germany, Netherlands, Portugal, Spain, US and UK

Rows

Total tracks: 12183 → **9819** after dropped duplicates and missing values

Columns

- **Track info:** track_id, track_name, artist_name, album_name, release_date, duration_ms, popularity
- **Audio features:** danceability, energy, key, loudness, mode, speechiness, acousticness, instrumentalness, liveness, valence, tempo → **all numerical**

Feature Engineering



In the context of Spotify's recommendation system, **popularity** plays a crucial role. Spotify's algorithms use a combination of user listening data, playlists, and other behavioral signals to generate personalized recommendations.

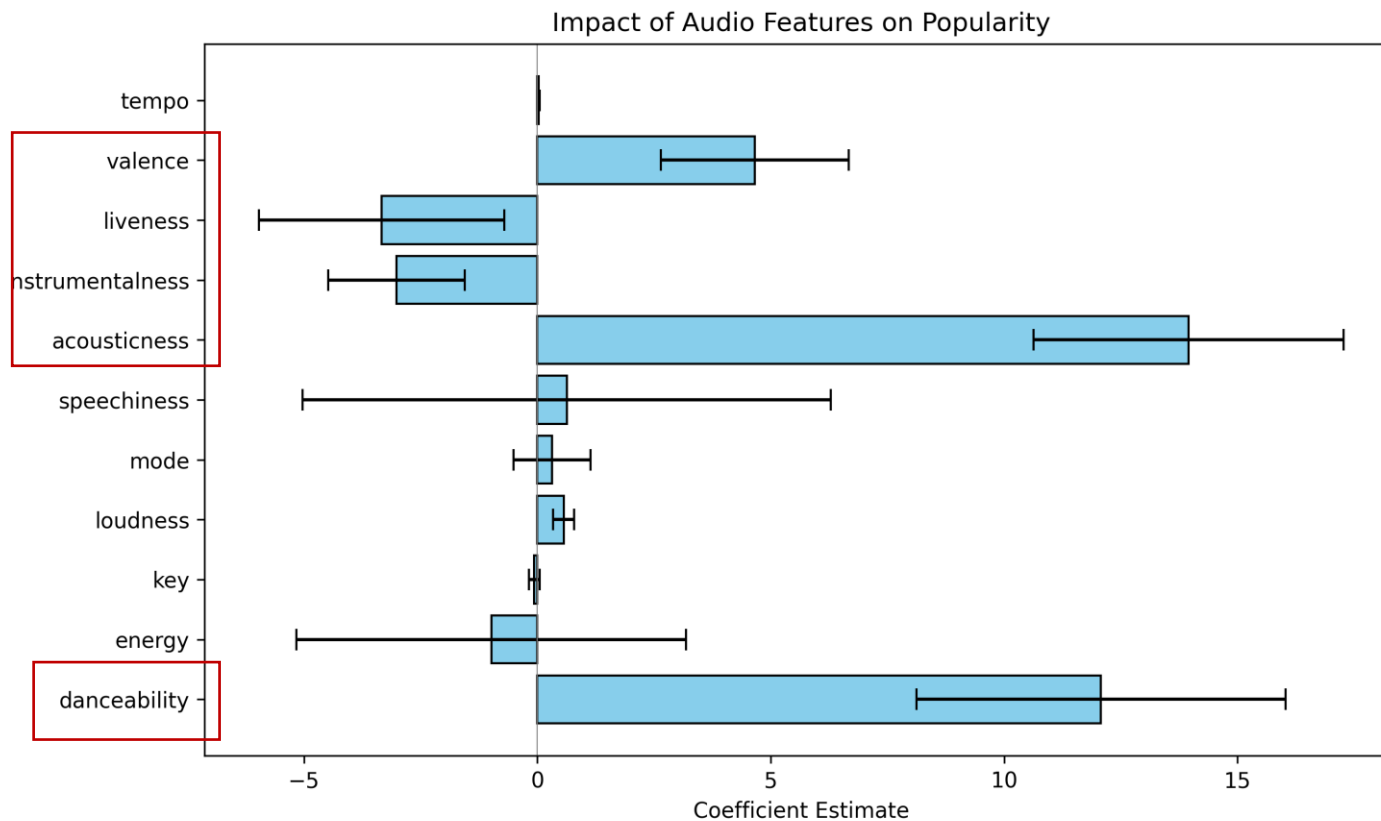
H0

There is **no significant** difference in the effect of all audio features on popularity.

H1

There is **a significant** difference in the effect of all audio features on popularity.

REJECTED



ANOVA results:

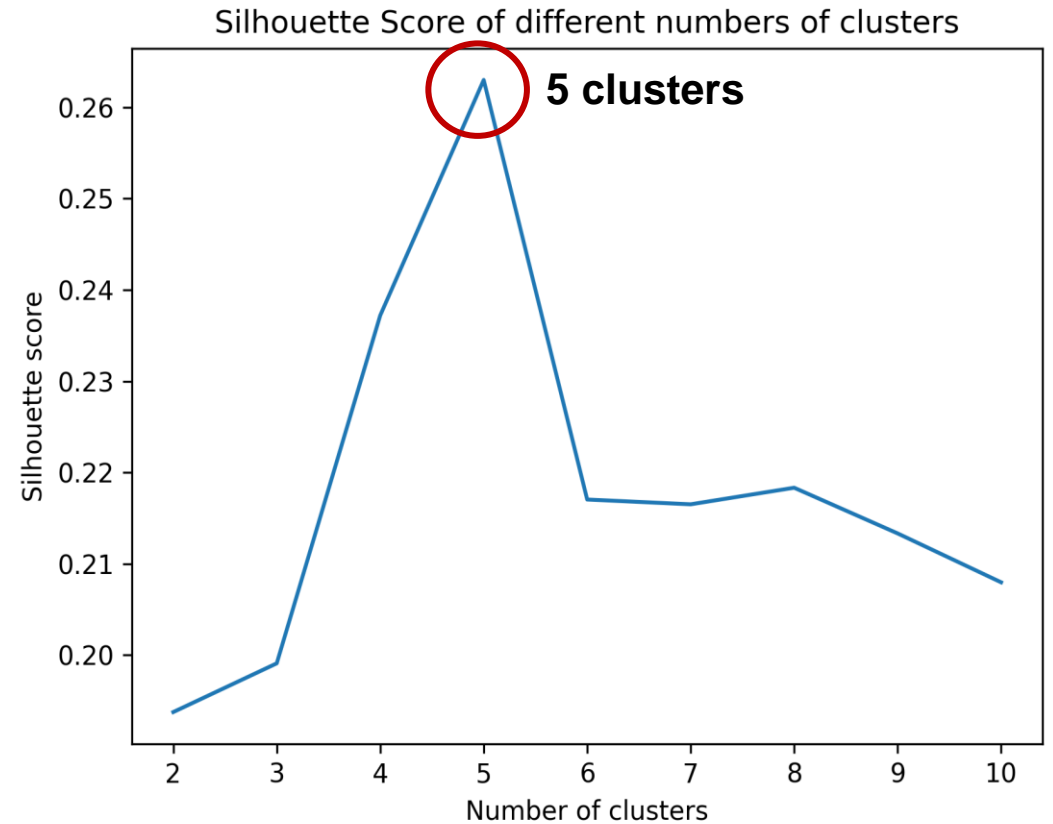
P-value = 0.0

ML Model – Clustering - Kmeans



Kmeans (K = n_clusters)

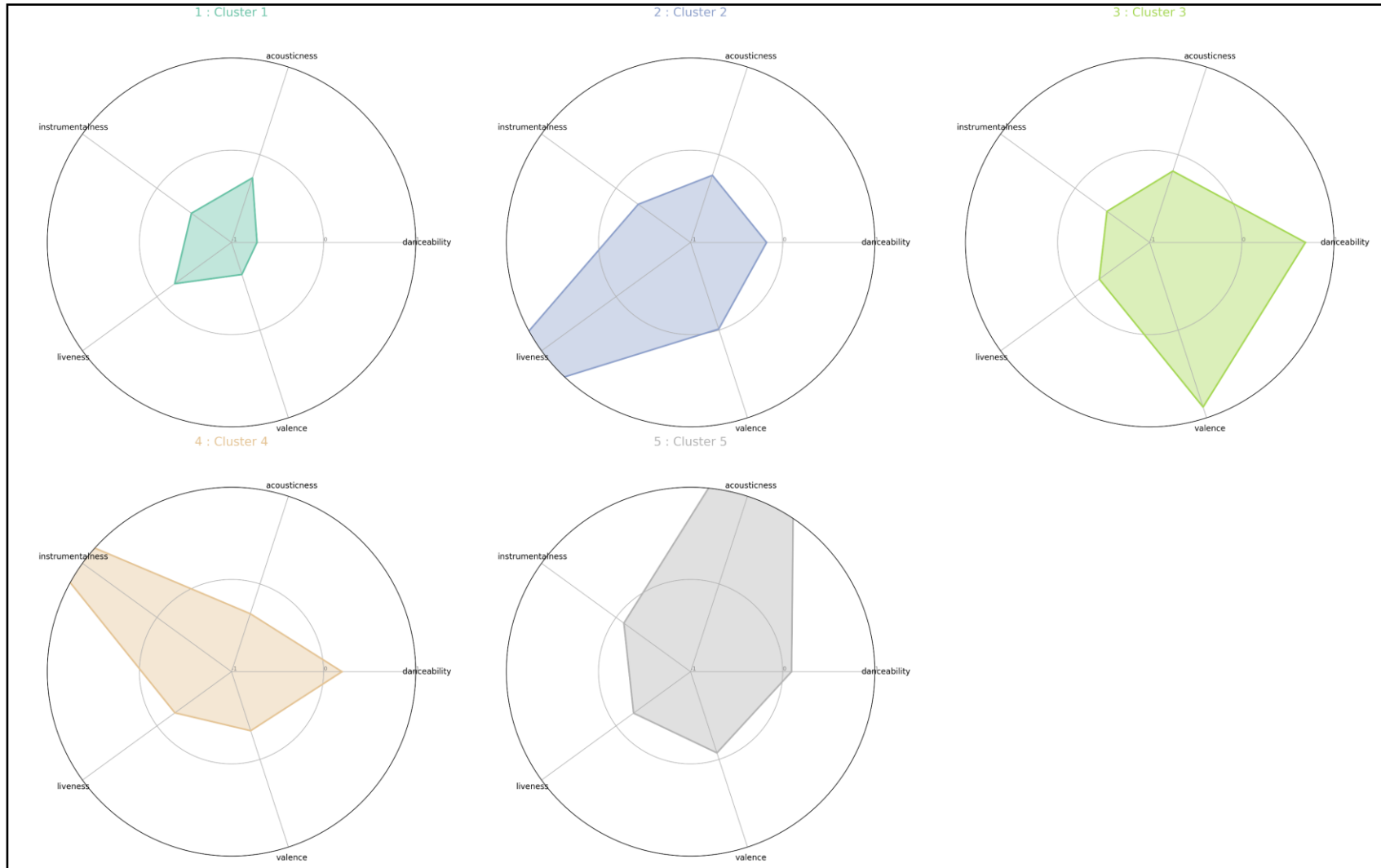
| Number of clusters | Silhouette Score |
|--------------------|---------------------------|
| 2 | 0.19375604269713528 |
| 3 | 0.19909345074975598 |
| 4 | 0.23720246143321097 |
| 5 | 0.2629938078904257 |
| 6 | 0.21703586804830247 |
| 7 | 0.21651330919147585 |
| 8 | 0.21332454520535876 |
| 9 | 0.20796953693676004 |



Clusters Visualization



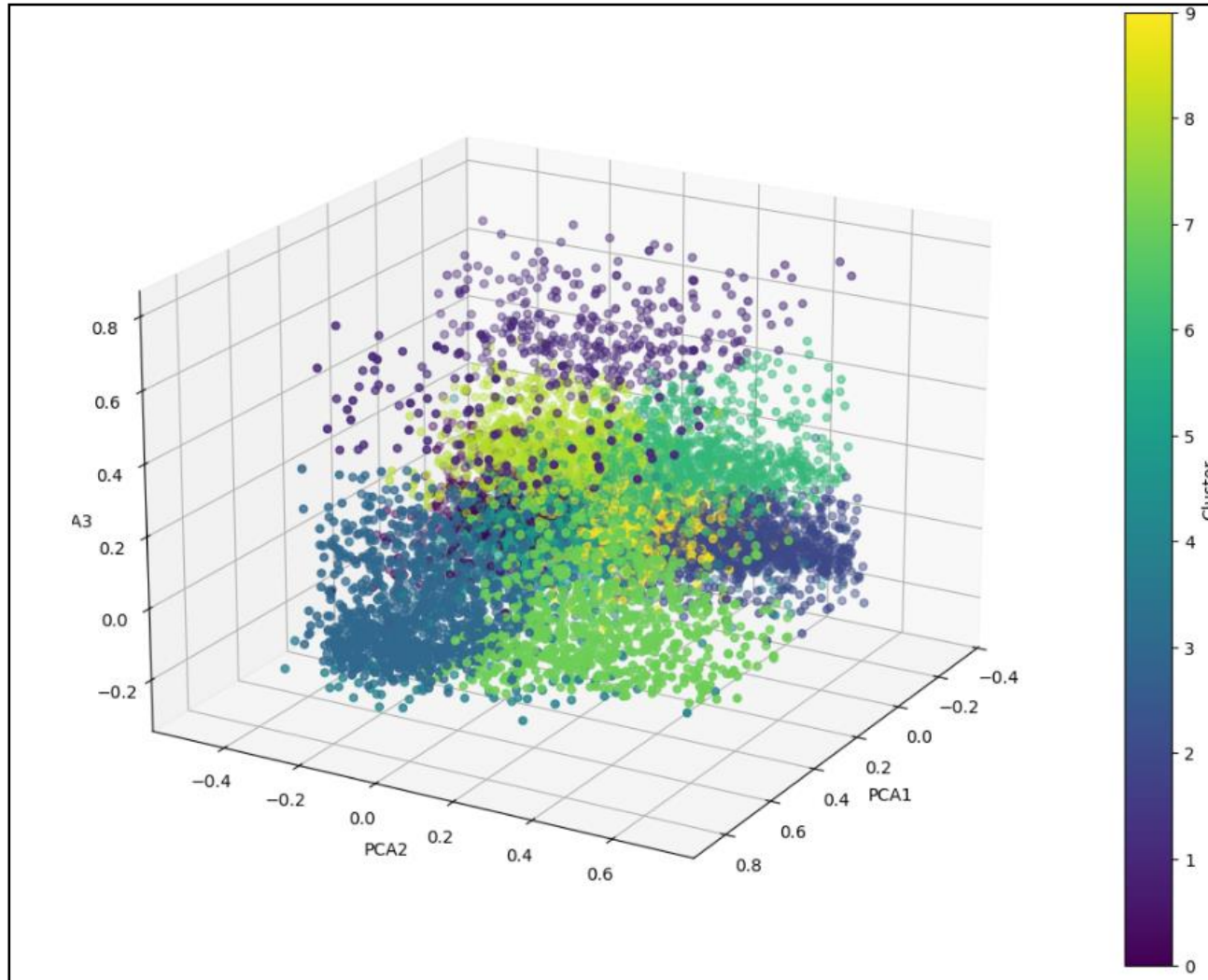
Radar Charts of 5 Clusters



Model Optimization – PCA



5 Clusters 3d View after PCA



PCA (Principal Component Analysis)

- Transform data from a high-dimensional space to a lower-dimensional space
- Noise reduction, reducing computational complexity

| | PCA1 | PCA2 | PCA3 | cluster |
|-------|-----------|-----------|-----------|---------|
| 0 | -0.133918 | 0.110011 | 0.200144 | 1 |
| 1 | -0.150705 | -0.014893 | -0.110876 | 2 |
| 2 | -0.206502 | 0.096055 | 0.030194 | 2 |
| 3 | -0.220934 | 0.094950 | 0.137786 | 1 |
| 4 | -0.237217 | 0.238084 | -0.016491 | 2 |
| ... | ... | ... | ... | ... |
| 12177 | 0.690894 | 0.155542 | 0.011461 | 3 |
| 12179 | -0.043342 | -0.200245 | 0.216556 | 4 |
| 12180 | 0.768750 | -0.135039 | 0.087999 | 3 |
| 12181 | -0.174781 | 0.007284 | 0.132586 | 2 |
| 12182 | -0.163114 | -0.178501 | 0.032709 | 4 |

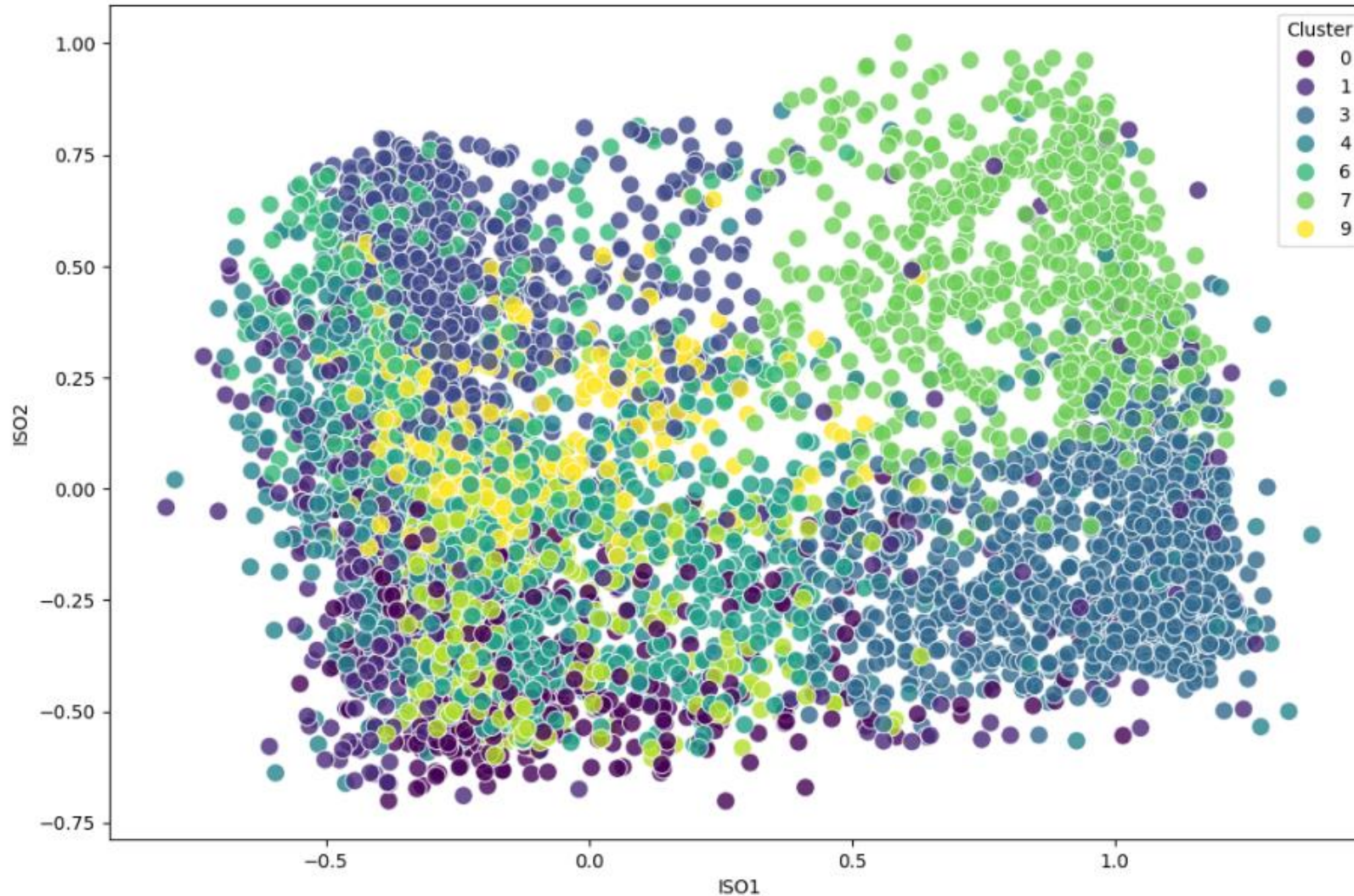
9818 rows x 4 columns

* **PCA (Principal Component Analysis)** and **ISOMAP (Isometric Feature Mapping)** are two commonly used dimensionality reduction algorithms for different types of data processing needs.

Model Optimization – ISOMAP



5 Clusters of ISOMAP



ISOMAP (Isometric Mapping)

- A non-linear dimensionality reduction method
- Preserve the manifold structure of the data in the high-dimensional space while mapping it to a lower-dimensional space.

Recommendation Demo



Demo in Jupyter Notebook

```
[39]: 1 # test by name and artist
2 track_name = "Blinding Lights"
3 artist_name = "The Weeknd"
4 recommended_songs, error = recommend_songs(track_name, artist_name)
5 if error:
6     print(error)
7 else:
8     display(recommended_songs[['track_name', 'artist_name', 'album_name', 'release_date']])
```

Song not found in the dataset.

```
[42]: 1 # test by name and artist
2 track_name = "Wake Me Up"
3 artist_name = "Avicii"
4 recommended_songs, error = recommend_songs(track_name, artist_name)
5 if error:
6     print(error)
7 else:
8     display(recommended_songs[['track_name', 'artist_name', 'album_name', 'release_date']])
```

| | track_name | artist_name | album_name | release_date |
|-------|---|--|--|--------------|
| 7058 | Follow Me (feat. Jason Derulo) - Bingo Players... | Hardwell, Jason Derulo, Bingo Players | United We Are (Remixed) | 2015-12-04 |
| 4934 | Devil (feat. Busta Rhymes, B.o.B & Neon Hitch) | Cash Cash, Busta Rhymes, B.o.B, Neon Hitch | Devil (feat. Busta Rhymes, B.o.B & Neon Hitch) | 2015-08-07 |
| 3503 | Under The Water | Camden Cox | Under The Water | 2021-02-19 |
| 10434 | Lost In Sound | ROY KNOX | Lost In Sound | 2019-03-23 |
| 430 | Better | Sonny Fodera | Better | 2022-06-10 |

Limitations and Learnings



Limitations

- Silhouette Score < 0.5
→ Low score
- Clusters have too many overlapping areas
- Clustering not very distinct

Learnings

- All tracks from "EDM"
- The differences in audio features of subgenres are not particularly significant

Next Step

- Using API to fetch more diverse playlists
- Streamlit – App interface
- Deep Learning Models

THANK YOU!

Congrats on graduation, geeks! Enjoy your summer!

