

Music Recommendations using Nearest Neighbors

Simpler Techniques

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Outline

Motivation

Data

Methods

Result

Discussion

Motivation

- **Domain:** When the end of a playlist approaches, users alter their attention to choosing the next song.
 - Distracting or time-consuming when the user is driving or at work.
 - To increase the user experience by personalization
 - Seamless continuation of a users playlist — curated playlists
- **Statistical:** Light-weight solution
 - Published literature
 - Acquire a foundational understanding of modern issues and methods

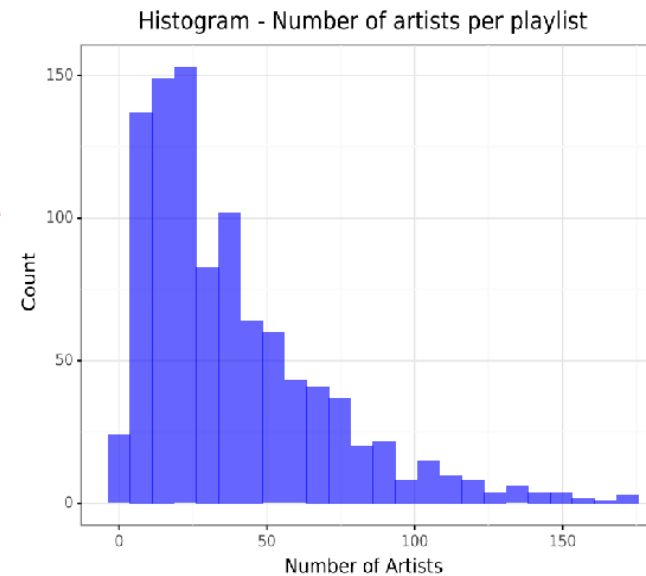
Data

Table 1: Playlists – Collaborative-based filtering. $n < p$, $k = \{\text{no: 0, yes:1}\}$

| | Song ₁ | Song ₂ | ... | Song _p |
|-----------------------|-------------------|-------------------|-----|-------------------|
| Playlist ₁ | k_1 | k_{12} | ... | k_{1p} |
| Playlist ₂ | k_{2p} | k_{2p} | ... | k_{2p} |
| ⋮ | ⋮ | | | |
| Playlist _i | k_{ip} | k_{ip} | ... | k_{ip} |
| ⋮ | ⋮ | | ... | |
| Playlist _n | k_{np} | k_{np} | ... | k_{np} |

Table 2: Audio Features – Content-based filtering. $n > p = 13$, $K \in \mathbb{R}$

| | playlist ID | loudness | ... | danceability | energy | track ID |
|---|-------------|----------|-----|--------------|----------|--------------------|
| 1 | 1 | K_{12} | ... | ... | K_{1p} | track ₁ |
| 1 | 1 | K_{12} | ... | ... | K_{1p} | track ₂ |
| ⋮ | ⋮ | | | | | |
| 2 | 2 | K_{22} | ... | ... | K_{2p} | track ₁ |
| ⋮ | ⋮ | | ... | | | |
| n | n | K_{n2} | ... | ... | K_{np} | track ₁ |



Source: 1000 random playlists from [Spotify Million Playlist Dataset Challenge](#).

- Spotify's API

Methods

- KNN for distance/similarity metrics
 - Collaborative-based Filtering — considers **multiple viewpoints** (playlists). Assumes that those "determined" to have closely related behaviours, also would like the same songs.
 - Content-based filtering — Uses an items (tracks) feature to **extract information** about the users and therefore, suggest a song.
- Similarity metrics: jaccard and euclidean
- Standardization for audio features
- To reduce dimensionality, Principal Component analysis (PCA) suggested 8 features

Results

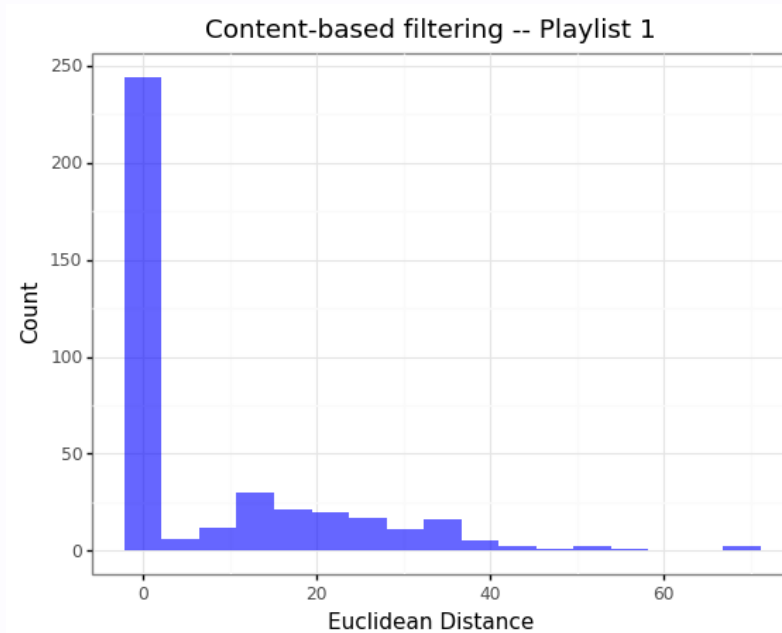
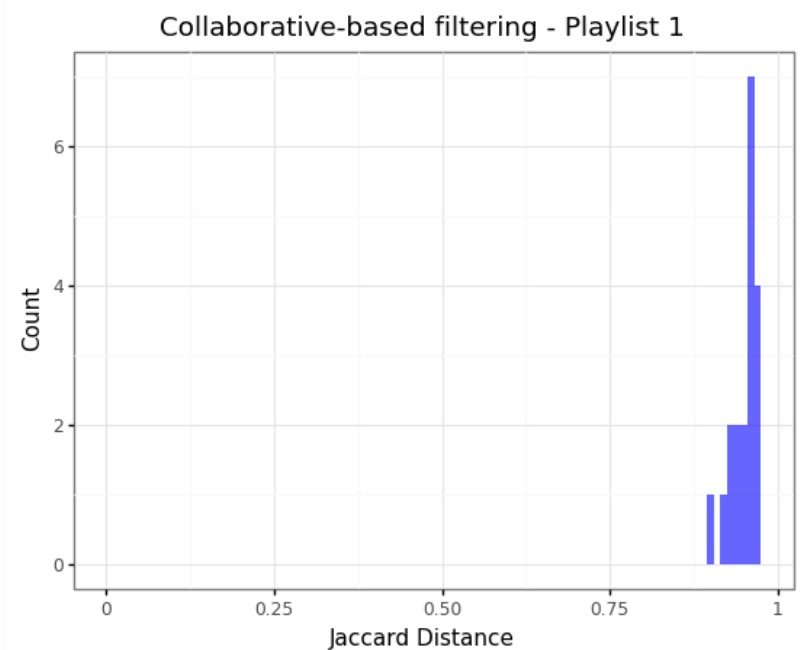


Table 3: Evaluation

| | MAP | NDCG |
|---------------|-----|------|
| Collaborative | 0.0 | 0.0 |
| Content | 0.0 | 0.0 |

- Both approaches give Mean Average Precision (MAP): 0.0
 - **R-precision** is defined as the proportion of the top-R retrieved documents that are relevant, where R is the number of relevant documents for the current query

Discussion

- The methods considered give a baseline for cases (clusters of user behaviors)
 - When a user does not have many songs, a collaborative approach is useful
 - When there are many songs, then use content-based filtering
 - **Hybrid**
- Slow learning rate. Large amount of data is needed.
- Interpretation of "similarity" can be misleading. Songs can be deemed to be "close" but upon listening to the recommended songs, it is much different.
- Simpler techniques relies **algorithmic approaches**
 - Further filtering needed to get the recommended songs
- More features — artist popularity, genre
- Feature Selection on K-Nearest Neighbor Algorithm Using Similarity Measure

Thank you!

References

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