DSCI 510: Principles of Programming Informatics

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Analyzing Genre Dominance and Release Timing in East Asian Music Project

1. Introduction

Every year, East Asia publish tens of thousands of songs. Therefore, I am curious about the current dominant song category. Additionally, I aim to explore the relationship between artist followers and top track popularity. To conduct this investigation, I've obtained two datasets from Kaggle, a popular platform for data science. Utilizing Python, I plan to employ code for reading and processing these datasets comprehensively. This analytical approach will unveil insights into the evolving musical landscape of East Asia and shed light on the dynamics between release dates and the ascent of top tracks in popularity.

2. Data and Data Processing

Data Collection

For the project, I use two main sets of data: one is about artists, called east_asia_top_artists, and the other is about music tracks, named east_asia_top_tracks. These files contain a lot of useful information. The artist data shows the names of singers or bands from East Asia, how many people follow them on Spotify, their music styles, and how popular their most famous songs are, which shown in the Figure 1. Also, the track data in the Figure 2 gives details about their hit songs, like when each song came out and how much people like it. Both of two datasets have 700 rows in total.

To get these files, I first had to sign up on a website called Kaggle, where people share a lot of different data. After signing up, Kaggle gave me a special username and a key to get into their system. With this key, I could go to a specific page on Kaggle (here: https://www.kaggle.com/datasets/crxxom/spotify-popular-east-asian-artists-and-tracks) and download the data I needed for my project. This made it easy to get all the information quickly and start working on understanding what makes these artists and songs so popular.

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|---|-------------|---------------------------------------|-----------|--|--------------------------------------|---|--|----------------------|
| | artist_name | popularity | followers | artist_link | genres | top_track | top_track_album | top_track_popularity |
| 0 | BTS | 88 | 67507448 | https://open.spotify.com/artist/3Nrfpe0tUJi4K4DXYWgMUX | ['k-pop', 'k-pop boy group', 'pop'] | Left and Right (Feat. Jung Kook of BTS) | CHARLIE | 88. |
| 1 | BLACKPINK | 82 | 43385244 | https://open.spotify.com/artist/41MozSoPIsD1dJM0CLPjZF | ['k-pop', 'k-pop girl group', 'pop'] | Shut Down | BORN PINK | 79. |
| 2 | TWICE | 79 | 18278225 | https://open.spotify.com/artist/7n2Ycct7Beij7Dj7mel4X0 | ['k-pop', 'k-pop girl group', 'pop'] | MOONLIGHT SUNRISE | READY TO BE | 75. |
| 3 | j-hope | 70 | 15407840 | https://open.spotify.com/artist/0b1slQumlAsNbqAolClSpy | ['k-pop', 'k-rap'] | on the street (with J. Cole) | on the street (with J. Cole) | 83. |
| 4 | V | 66 | 14088149 | https://open.spotify.com/artist/3JsHnjpbhX4SnySpvpa9DK | ['j-division', 'korean ost'] | Christmas Tree | Our Beloved Summer (Original Television Soundtrack), Pt. 5 | 84. |
| 5 | Stray Kids | 82 | 11990935 | https://open.spotify.com/artist/2dlgFjalVxs4ThymZ67YCE | ['k-pop', 'k-pop boy group', 'pop'] | S-Class | 5-STAR | 89. |
| 6 | SEVENTEEN | 78 | 9731900 | https://open.spotify.com/artist/7nqOGRxiXj7N2JYbgNEjYH | ['k-pop', 'k-pop boy group'] | Super | SEVENTEEN 10th Mini Album 'FML' | 87. |
| 7 | EXO | 75 | 9701818 | https://open.spotify.com/artist/3cjEqqelV9zb4BYE3qDQ4O | ['k-pop', 'k-pop boy group'] | Cream Soda | EXIST - The 7th Album | 83. |
| 8 | Jung Kook | 84 | 8656152 | https://open.spotify.com/artist/6HaGTQPmzraVmaVxvz6EUc | ['k-pop'] | Seven (feat. Latto) (Explicit Ver.) | Seven (feat. Latto) | 97. |

Figure 1. east asia top artists dataset

| | song_name | album_name | album_link | artist_name | popularity | release_date |
|---|-------------------------------------|----------------------|---|-------------|------------|--------------|
| 0 | Cupid - Twin Ver. | The Beginning: Cupid | https://open.spotify.com/album/5letLUZIFsQikJYShfGNs4 | FIFTY FIFTY | 98 | 2023-02-24 |
| 1 | Seven (feat. Latto) (Explicit Ver.) | Seven (feat. Latto) | https://open.spotify.com/album/53985D8g3JcGBoULSOYYKX | Jung Kook | 97 | 2023-07-14 |
| 2 | Like Crazy | FACE | https://open.spotify.com/album/4xc3Lc9yASZgEJGH7acWMB | Jimin | 96 | 2023-03-24 |
| 3 | MONEY | LALISA | https://open.spotify.com/album/4ASxFYWyk2216OloHoaSh8 | LISA | 96 | 2021-09-10 |
| 4 | Take Two | Take Two | https://open.spotify.com/album/3jeQDa9OFZ6GndLindHx3k | BTS | 95 | 2023-06-09 |
| 5 | Like Crazy | Like Crazy (Remixes) | https://open.spotify.com/album/639nejcoHHwxJCKqr35ww2 | Jimin | 93 | 2023-03-26 |
| 6 | Super Shy | NewJeans 'Super Shy' | https://open.spotify.com/album/5V729UqvhwNOcMejx0m55I | NewJeans | 93 | 2023-07-07 |
| 7 | Seven (feat. Latto) (Clean Ver.) | Seven (feat. Latto) | https://open.spotify.com/album/53985D8g3JcGBoULSOYYKX | Jung Kook | 93 | 2023-07-14 |

Figure 2. east asia top tracks dataset

ii. **Data Cleaning**

Since there are some information which are irrelevant to my project, I refined the raw data by selecting key columns such as "artist_name", "followers", "top_track", "top_track_popularity", "top_track_release_date", "genres", and "query_genre". When I processed the code without definition formula, I found that there are four artists were missing value: Yui Aragaki, French Kiss, Rino Sashiara, Eriko Tamura. To address this, I tried to the artist data within the track data. However, Yui Aragaki had the track information available, and other three artists did not have which meant that their records could not contribute to a precise analysis. Therefore, I decided to exclude these incomplete entries from the dataset. I utilized libraries such as pandas for data manipulation and JSON for data storage. It allowed for an efficient workflow, where the necessary data was loaded, filtered, and then saved in a JSON format for further use. (Figure 3)

```
"BTS"
    artist name:
                            67507448
    followers:
   top_track:
                            "Left and Right (Feat. Jung Kook of BTS)"
                          88
    top_track_popularity:
   top_track_release_date: "2022-10-06"
                            "['k-pop', 'k-pop boy group', 'pop']"
    genres:
                            "j-pop"
    query_genre:
▼ 1:
    artist_name:
                          "BLACKPINK"
   followers:
                          43385244
   top_track:
                            "Shut Down"
    top_track_popularity:
   top_track_release_date: "2022-09-15"
                            "['k-pop', 'k-pop girl group', 'pop']"
    genres:
    query_genre:
                            "TWICE"
    artist name:
followers:
                     18278225
    top_track:
                            "MOONLIGHT SUNRISE"
    top_track_popularity:
                            75
    top_track_release_date: "2023-03-10"
                            "['k-pop', 'k-pop girl group', 'pop']"
                            "j-pop"
    query_genre:
```

Figure 3. Cleaned data in JSON format

3. Analysis

In order to find out the most popular type of song, the first step is to load the data and identifying the top 50 tracks based on their popularity scores. Also, I sorted all the tracks by the column top_track_popularity in descending order. This means that the tracks with the highest popularity are placed at the top of the DataFrame. After sorting, the function then selects the top 50 tracks by using the head (50) method,

which takes the first 50 rows of the DataFrame. These rows represent the 50 most popular tracks according to the dataset. This filtering was crucial to pinpoint the tracks that have been most impactful to listeners. Once these tracks were selected, the next step was to standardize the genre labels. Recognizing the array of genre descriptors used on Spotify, which can often include sub-genres and overlapping terms, I implemented a normalization process that consolidated genre variations into broader, recognizable categories, specifically targeting "k-pop" and "j-pop". I attached the result of analysis step below in Figure 4.

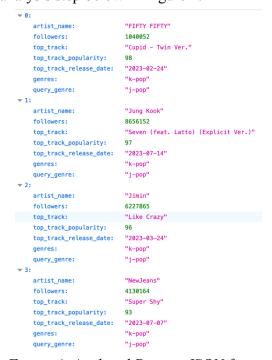


Figure 4. Analyzed Data in JSON format

4. Results

At first, I created a bar chart categorize by genres, and I set x label as 'Count Type of Genres', y label as 'Type of Genres', and the title as 'Distribution of Genres'. (Figure 5) It is obvious that many types of genres contain two kinds instead of one, and the count type of k-pop and j-pop were less than actual number.

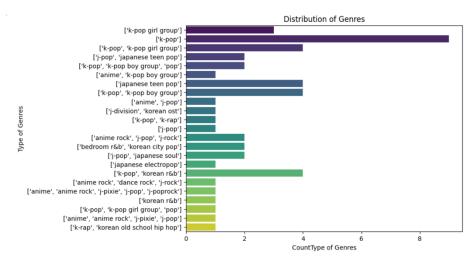


Figure 5. Distribution of Genres

We could tell that k-pop is the most popular type of song in East Asia. However, there are many songs have two types of genre classification, I need to separate the type in each genre. Except for k-pop, j-pop seems like the second popular type of song, and I will count the type of song to make sure how many k-pop and j-pop song list in the top 50. (Figure 6) As the result, there are 29 songs belongs to k-pop genre, 10 songs belong to j-pop genres, and rest of 11 songs belong to other type of genres. Based on the result, K-pop is the most popular genre of music in East Asia, followed by J-pop as the second most popular The visualization result is attached in the Figure 7.

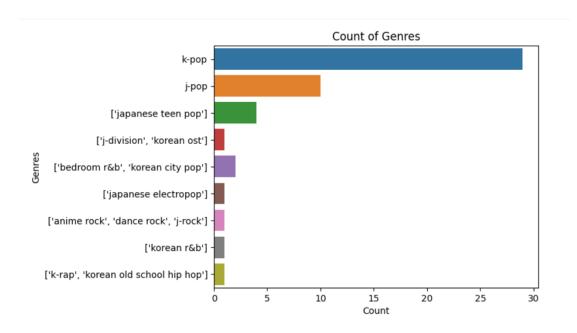


Figure 7. Count of Genres Bar Chart

I created a scatter plot to analyze the relationship between the number of artist followers and the popularity of their top tracks, setting "followers" as the x-axis, "top_track_popularity" as the y-axis, and "genres" as the hue. However, I found no correlation between the number of artist followers and the popularity of their top tracks

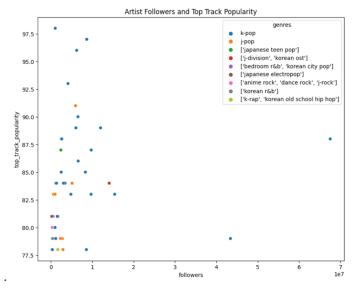


Figure 8. Count of Genres Bar Chart

5. Future Work

With additional time, I would deeper the project in temporal analysis to track the evolution and endurance of music trends. This would involve scrutinizing data over extended periods to discern long-term patterns in genre popularity and artist success, as well as seasonal variations in listening habits. Understanding these temporal dynamics could inform predictions about future industry trends and guide artists and producers in making strategic decisions about releases and marketing efforts, ultimately enabling a more proactive rather than reactive stance in a rapidly changing music landscape.