DATA ENGINEERING INDIVIDUAL COURSEWORK

SPOTIFY PLAYLIST DATABASE:

A SENTIMENT ANALYSIS



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Word Count:

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1.0 INTRODUCTION

"Music can heal the wounds which medicine cannot touch", says Debasish Mridha. From many decades ago, music has already been recognised as an opportunity to address mental health challenges (Schriewer and Bulaj, 2016). Nowadays, audience can enjoy the music more conveniently via music streaming services, instead of downloading the original audio file of a song. As one of the biggest music streaming platforms, Spotify had over 365 million users by February 2022 (Caddy, 2022).

One important feature of Spotify is the editorial playlists. Spotify's in-house teams curate these playlists by selecting collections of songs that somehow have some similarities so that meaningful playlists are created. Many playlists have emotion-related tags such as "happy", "moody" and "chilling", and each of them is designed to match the audience's emotions.

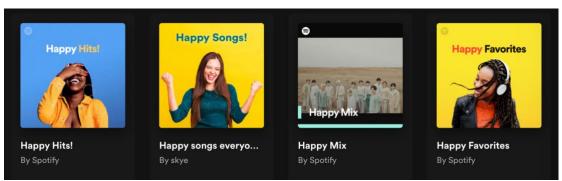


Figure 1: Spotify Curated Playlists Related to "Happy"

While the Spotify playlist is an important feature of the platform in helping the artists reach more of the target audiences, it is worthy of investigation that the exact criterion of how a song is featured on a certain playlist. This project aims to create a database of a specific Spotify playlist that contains a number of different attributes of the songs within that playlist and conduct further sentiment analyses of these songs based on the attributes.

In achieving the objective of the project, the following steps are executed and will be explained in detail throughout this report: First, real-time streaming data will be extracted through API scraping from Spotify and several other relevant platforms. The data will then be processed and stored in a suitable way so that further analyses are allowed. This project will then conduct sentiment analyses on these processed data so that we can evaluate whether the "emotion" of the songs in the playlist match with the playlist title. It is believed that successful execution of the project with provide precious guidelines for sentiment related research on a bigger database beyond the scope of this project.

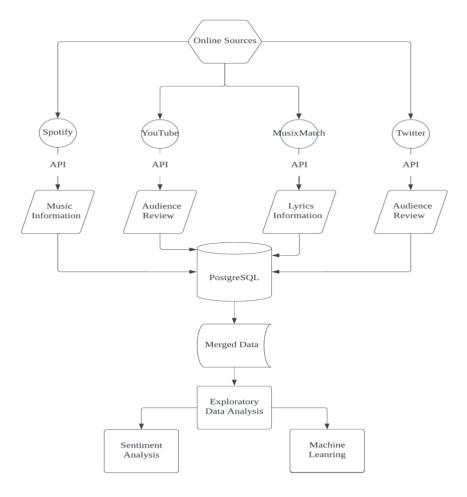


Figure 2: Project Workflow

2.0 DATA COLLECTION

This project mainly used the Application Programming Interface (API) techniques in the Python language to extract data from four different platforms including Spotify, MusixMatch, YouTube and Twitter. The API is an important tool to obtain data from dynamic websites and allows some level of customisations (Medium, 2022).

2.1 SPOTIFY DATA

As the main subject matter of this project, data from Spotify was scraped first. Spotify provides API services on its developer website and a number of functions can be achieved under the free user plan. On the developer website, a Python library called Spotipy (https://github.com/plamere/spotipy) is recommended for executing the API calls. This package was utilised by this project throughout the data extraction from Spotify.

First of all, the project wanted to retrieve a list of featured playlists (Editor's Picks) on Spotify to decide which playlist to extract and evaluate. By using the featured_playlists() method, a list of playlists which were the Editor's picks on the day when the data was scraped was returned:

```
Editor's picks

0 New Music Friday

1 Feel Good Friday

2 RapCaviar

3 Main Stage

4 I Love My '90s Hip-Hop

5 Mood Booster

6 Dance Hits

7 Today's Top Hits

8 just hits

9 Dance Party

10 Happy 80s

11 young & free
```

Figure 3: Featured Playlists on 8th April

The sixth result was a playlist named "Mood Booster", which was highly related to the research objective - to analyse the emotion of songs with sentimental analysis techniques and evaluate whether the emotion of the song match with the overall playlist genre.



Figure 4: The *Mood Booster* Playlist

According to Spotify's description of this playlist, the songs in the playlist are supposed to make the audience "feel good" and "get happy". This project is interested in the if the emotions of the songs in this playlist actually match that purpose. To dig deeper into the tracks in this playlist, the playlist's ID on Spotify needs to be known. Unfortunately, at the moment, the only feasible way to get a playlist's ID is through getting a user's current playlist. Therefore, this playlist was manually followed on Spotify and added to the profile using a personal Spotify account.

With the playlist id being known, the project was able to retrieve more information about the tracks in this the playlist. A list 76 song names were first retrieved via the playlist items() method, and the 76 songs' corresponding Spotify IDs were retrieved via similar ways. The songs' Spotify IDs allowed the project to retrieve a number of attributes of the track by using the Spotify API's track() method, including the artist information. By calling the method and looping into the nested dictionary (see outcome example in Figure 5) returned by the method, this project was able to get the artist's name, album, Spotify popularity, release date, duration information of the track. The method could also tell whether a track contains explicit content or not: most music streaming platforms distinguish and differentiate between tracks that is suitable for mainstream consumption, and those songs that may contain a parental advisory or may be considered explicit content (Soundplate, 2022). On Spotify, a track with explicit content will have a "E" or "Explicit" symbol next to its title. With the API, boolean values of True or False was returned regarding the "explicit" attribute.

```
{'album': {'album_type': 'single',
            artists': [{'external_urls': {'spotify': 'https://open.spotify.com/artist/2ZmXexIJAD7PgABrj0qQRb'},
                         'href': 'https://api.spotify.com/vl/artists/2ZmXexIJAD7PgABrj0qQRb'
                        'id': '2ZmXexIJAD7PgABrj0qQRb',
                        'name': 'N.Flying'
                        'type': 'artist',
                        'uri': 'spotify:artist:2ZmXexIJAD7PgABrj0qQRb'}],
           'available markets': ['AD',
                                  'AE',
                                  'AG',
                                  'AL',
'disc number': 1,
'duration_ms': 210652,
'explicit': False,
'external_ids': {'isrc': 'KRA381900017'},
'external_urls': {'spotify': 'https://open.spotify.com/track/2LwH6T39A5IODRgPv9XitR'},
'href': 'https://api.spotify.com/v1/tracks/2LwH6T39A5IODRgPv9XitR',
'id': '2LwH6T39A5IODRgPv9XitR',
'is_local': False,
'name': 'Rooftop'
'popularity': 61,
'preview url': 'https://p.scdn.co/mp3-preview/a22310aa8b97d93e7e850c35a6e04f1165b11419?cid=7b1fa7a7eb25461f8d3a4a66e1966de5',
track_number': 1,
'type': 'track',
'uri': 'spotify:track:2LwH6T39A5IODRgPv9XitR'}
```

Figure 5: Output Example of the track() Method

2.2 MUSIXMATCH DATA

MusixMatch is an Italian music data company which has the world's largest database of 14 million lyrics items in various different languages (Baydeer, 2021). With the lyrics data provided by MusixMatch, this project would be able to conduct further sentiment analysis on the lyric strings.

With the free API plan provided by MusixMatch, the account created was limited to 2,000 API calls per day, and only 30% of the lyrics of a song was accessible. The project would have to assume most songs would have their emotions set in stone in the very first bit.

By inputting the title and the artist's name of the song as the parameters of the MusixMatch request call, JSON styled results were pulled. In Figure 6, this project used *Halsey's* song "*Drive*" as an example to examine the output.

Figure 6: Output Example a Random Song

2.3 YOUTUBE DATA

Other than emotion-related attributes, this project is also interested in the popularity of the official music videos and user-uploaded lyrics videos of the tracks in the playlist on YouTube.

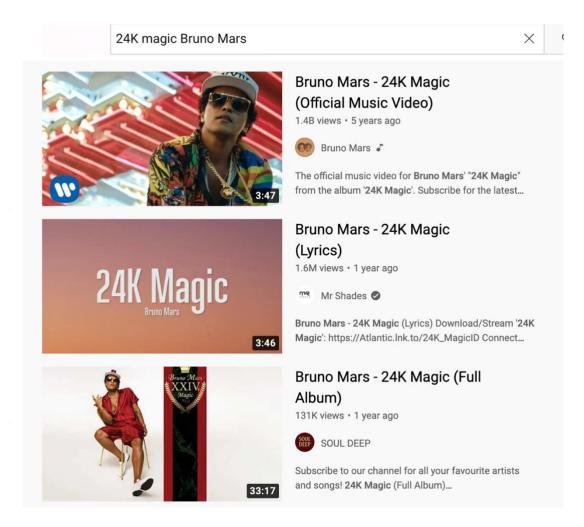


Figure 7: Top3 YouTube Results of Searching "24K Magic" + Bruno Mars

Figure 7 displays the top results of querying the keywords "24K Magic" (song title) and "Bruno Mars" (artist's name) on YouTube, which includes the official music video of the song uploaded by the artist's own channel as well as user-uploaded lyrics and album videos.

By using the YouTube V3 API's method *Search()*, this project first got the top 10 query results' YouTube video IDs for each of the 76 songs and stored them in a 76*10 dictionary. The number 10 was decided because they must be the most relevant videos to the song. To get the attributes like the view and like count for each video, the YouTube V3 *statistics()* method was used. This project retrieved a total of four statistics of each of the 760 videos: the view, like, favourite and comment count, and calculates the mean of each count for each of the 76 songs. The four attributes should reflect the popularity of the songs on YouTube.

2.4 TWITTER DATA

Twitter is one of the biggest social networking services in the world where users can post and interact with messages known as "tweets" (Igi-global, 2022). This project is also interested in the emotions associated with the tweets that discuss the tracks in the *Mood Booster* playlist.

To retrieve the tweet strings, the Twitter API was used, and a simple Python wrapper called *Twitter API* (https://github.com/geduldig/TwitterAPI) suggested by the Twitter developer's website was used for making the requests in Python.

The standard result of searching a random query ("University College London" in this case) is displayed in Figure 8.

Figure 8: Output Example of the Twitter API

The project is only interested in the "text" section in the requesting result.

Similar to the previous two requests from MusixMatch and YouTube, this project also used the combination of the "song title" and the "artist's name" as the input parameter. By querying the 76 combinations and only retaining the "text" output, the last part of the database was successfully retrieved.

3.0 DATA PROCESSING

Most of the API result are in JSON formats or in a nested dictionary. By enumerating through these formats in Python, this project was able to generate four dataframes for four of the different platforms. Each dataframe has 76 rows and has the track name and artist's name columns as the foreign keys.

	track_name	spotify_id	artist_name	album	spotify_popularity	release_date	duration	explicit_content
0	One Right Now (with The Weeknd)	00Blm7zeNqgYLPtW6zg8cj	Post Malone	One Right Now	92	2021-11-05	193506	True
1	dancing in the kitchen	0ohcCrxZkBfFbkuRPOZQZX	LANY	dancing in the kitchen	76	2021-06-25	208599	False
2	Sheesh!	3ddNKnYpVx0ul8vcwbTQ5Y	Surfaces	Sheesh!	75	2021-08-20	148846	False
3	Can I Get It	6w8ZPYdnGajyfPddTWdthN	Adele	30	82	2021-11-19	210384	False
4	Black And White	7rpNuuoMbid56XkDsx2FjE	Niall Horan	Heartbreak Weather	78	2020-03-13	193089	False

Figure 9: Spotify_df

The main dataframe, *Spotify_df*, has 8 columns in total and 6 of them are unique Spotify attributes: the Spotify ID, album, Spotify popularity, release date, duration information of a song and whether it contains explicit content.

	track_name	artist_name	lyrics
0	One Right Now (with The Weeknd)	Post Malone	Na-na-na-na, na-na Na-na-na-na, oh no Yeah, ye
1	dancing in the kitchen	LANY	City lights looking like ice underneath the st
2	Sheesh!	Surfaces	You know what I'm sayin'? (Sheesh) I be like

Figure 10: Musixmatch_df

The *Musixmatch_df* dataframe has 3 columns and only the lyrics column is unique which contains the lyric string of each song.



Figure 11: Youtube_df

The Youtube_df dataframe has 4 unique columns which are the average view, like, favourite and comment counts for the 10 most relevant videos of each of the 76 songs on YouTube.

	track_name	artist_name	tweets
0	One Right Now (with The Weeknd)	Post Malone	I'm obsessed with this bop by The Weeknd and P
1	dancing in the kitchen	LANY	Hi everyone! One of my favorite songs is danci

Figure 12: Twitter_df

Similar to the *Musixmatch_df* dataframe, the *Twitter_df* only contains one unique column, which is the tweet strings that discuss each of the 76 songs in the playlist.

4.0 DATA STORAGE

4.1 LOCAL STORAGE

The four dataframes are first exported to 4 csv files for local storage. However, for more flexible, affordable, and scalable data management, the dataframes need to be stored in a more reliable cloud storage database.

4.2 CLOUD DATABASE

This project chose the PostgreSQL as the database management system. PostgreSQL is a powerful open-source object-relational database system with a solid reputation for active development and stability, functional robustness, and good performances for over 30 years (PostgreSQL, 2022). This system would also allow the project to conduct analyses via SQL queries via the Postgres connection and a relational database that had meaningful linkages could be created.

Initialises the db_engine using my own credentials db_engine = create_engine('postgresql://doratian18:qwerty123@depgdb.crhso94tou3n.eu-west-2.rds.amazonaws.com:5432/doratian18')

Figure 13: Initialising the Database Connection

With the user, host name and port number being initialised, the *db_engine* was created for future connections to this database.

doratian18-> \dt List of relations Schema Name Type Owner					
public public public public public public public public (7 rows)	+ Company_stock_sql PARA_stock_news_sql PARA_stock_sql musixmatch_df spotify_df twitter_df youtube_df	+	doratian18 doratian18 doratian18 doratian18 doratian18 doratian18 doratian18		

Figure 14: Dataframes Stored in PostgreSQL

By connecting to the database in the terminal and using the command line prompts to check the tables, the last four rows in Figure 14 indicates that the four dataframes were successfully stored in the database and had the correct ownership.

5.0 RELATIONAL DATABASE

5.1 SCHEMA

Now that

5.2 SQL QUERIES

6.0 EXPLORATORY DATA ANALYSIS

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7.0 SENTIMENT ANALYSIS

7.1 WORD CLOUD

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7.2 NLTK SENTIMENT ANALYSIS

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8.0 REGRESSION MODEL

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9.0 FUTURE OPPORTUNITIES

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10.0 CONCLUSION

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11.0 BIBLIOGRAPHIES

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12.0 APPENDIX

HEADING 2