

# SPOTIFY DATA ANALYSIS USING SQL



Data Cleaning and Insights on  
Song Popularity

# INTRODUCTION

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This project involves cleaning and analyzing Spotify data from two datasets. The datasets used are `high_popularity_spotify_data` and `low_popularity_spotify_data`. Music is universal, and this analysis aims to provide insights that resonate with anyone interested in Spotify's rich database of songs and artists.

# DATASET OVERVIEW

## Datasets Used:

1. High Popularity Spotify Data
2. Low Popularity Spotify Data

### Key Columns:

track\_name (Song Name)  
track\_artist (Artist Name)  
track\_popularity (Popularity Score)  
track\_album\_release\_date (Release Date)  
energy, danceability, key, etc.



# STEP 1 - DATA CLEANING

## Removing Duplicates

```
SELECT
    track_name, COUNT(*) AS count_name
FROM
    music.high_popularity_spotify_data
GROUP BY track_name
HAVING COUNT(*) > 1
ORDER BY count_name DESC;
```

| Result Grid     Filter Rows: <input type="text"/>   <a href="#">Expo</a> |  |            |
|--|--|------------|
|  | track_name                                     | count_name |
| ▶  | Not Like Us                                    | 3          |
|  | Timeless (with Playboi Carti)                  | 2          |
|  | Sticky (feat. GloRilla, Sexyy Red & Lil Wayne) | 2          |
|  | QuÃ© PasarÃ-a...                               | 2          |
|  | NEW DROP                                       | 2          |
|  | Tu Boda  | 2          |
|  | Si Antes Te Hubiera Conocido                   | 2          |
|  | QUE VASHACER HOY?                              | 2          |
|  | +57  | 2          |





# STEP 2 - DATA ANALYSIS

## Basic Queries

1. Find the total number of Pop songs in the High Popularity dataset

```
SELECT
    playlist_genre, COUNT(*) AS Total_popsong
FROM
    music.high_popularity_spotify_data
WHERE
    playlist_genre LIKE '%pop%'
GROUP BY playlist_genre;
```

Result Grid



Filter Rows:

|   | playlist_genre | Total_popsong |
|---|----------------|---------------|
| ▶ | pop            | 47            |



## 2. Identify the most common genre in the Low Popularity dataset

```
SELECT
    playlist_genre, COUNT(*) AS Common_genre
FROM
    music.low_popularity_spotify_data
GROUP BY playlist_genre
ORDER BY Common_genre DESC
LIMIT 1;
```

| Result Grid |                |              | Filter Rows: |
|-------------|----------------|--------------|--------------|
|             | playlist_genre | Common_genre |              |
| ▶           | lofi           | 121          |              |





### 3. Calculate the average popularity score of Rock song

```
SELECT
    AVG(track_popularity) AS avg_score
FROM
    music.high_popularity_spotify_data
WHERE
    playlist_genre = 'Rock';
```

| Result Grid |           |
|-------------|-----------|
|             | avg_score |
| ▶           | 75.1538   |



# Intermediate Queries

1. List artists who have produced songs in both Pop and hip-hop genres

```
SELECT
    a.track_artist
FROM
    music.high_popularity_spotify_data AS a
    JOIN
    music.high_popularity_spotify_data AS b ON a.track_artist = b.track_artist
WHERE
    a.playlist_genre LIKE '%pop%'
    AND b.playlist_genre LIKE '%hip_hop%';
```

|   | track_artist                                       |
|---|--|
| ▶ | Tyler, The Creator, GloRilla, Sexyy Red, Lil Wayne |
|   | The Weeknd, Playboi Carti                          |
|   | GloRilla, Sexyy Red                                |
|   | Don Toliver  |
|   | Kendrick Lamar                                     |
|   | Kendrick Lamar                                     |



## 2. Find Artists with the Highest Average Popularity Across Both Tables

```
SELECT
    track_artist, AVG(track_popularity) AS Average_popularity
FROM
    (SELECT
        track_artist, track_popularity
    FROM
        music.high_popularity_spotify_data UNION ALL SELECT
        track_artist, track_popularity
    FROM
        music.low_popularity_spotify_data) AS Combined_data
GROUP BY track_artist
ORDER BY Average_popularity DESC
LIMIT 1;
```

| Result Grid |                       |                    | Filter Rows: |
|-------------|-----------------------|--------------------|--------------|
|             | track_artist          | Average_popularity |              |
| ▶           | Lady Gaga, Bruno Mars | 100.0000           |              |



### 3 .Find the Overlapping Artists in Both Datasets

```
SELECT DISTINCT
```

```
    a.track_artist
```

```
FROM
```

```
    music.high_popularity_spotify_data AS a
```

```
    JOIN
```

```
    music.low_popularity_spotify_data AS b ON a.track_artist = b.track_artist;
```

| track_artist                      |
|-----------------------------------|
| Creedence Clearwater Revival      |
| Van Halen                         |
| BabyChiefDoit                     |
| 41, Kyle Richh, Jenn Carter, TaTa |
| Pop Smoke                         |
| Teto, MatuÃa                      |
| ArcÃingel, Bad Bunny              |
| Youka                             |

### 4 .list songs that appear in both high and low popularity datasets

```
SELECT DISTINCT
```

```
    c.track_name
```

```
FROM
```

```
    music.high_popularity_spotify_data AS c
```

```
    JOIN
```

```
    music.low_popularity_spotify_data AS d ON c.track_name = d.track_name;
```

| track_name                                 |
|--|
| The Viper                                  |
| Bent (with Kyle Richh, Jenn Carter & TaTa) |
| M4   |
| Me Acostumbre (feat. Bad Bunny)            |
| Hm Hm hm                                   |

## 5 .identify artists who improved their average song popularity over time

```
SELECT
    track_artist,
    AVG(track_popularity) AS song_popularity,
    track_album_release_date
FROM
    music.high_popularity_spotify_data
GROUP BY track_artist , track_album_release_date
ORDER BY song_popularity ASC
LIMIT 5;
```

| Result Grid  |                                   |                 |                          |
|--|-----------------------------------|-----------------|--------------------------|
| Filter Rows: <input type="text"/>                      |                                   |                 |                          |
| Export: <input type="button" value="Export"/>          |                                   |                 |                          |
| Wrap Cell Content: <input type="button" value="Wrap"/> |                                   |                 |                          |
|  | track_artist                      | song_popularity | track_album_release_date |
|  | Van Halen                         | 68.0000         | 1978-02-10               |
|  | Arc̃angel, Bad Bunny              | 68.0000         | 2017-04-11               |
|  | Bryant Myers, Bad Bunny           | 68.0000         | 2017-06-05               |
|  | Filipe Ret, Dallass, Hunter       | 68.0000         | 2023-03-21               |
| ▶  | 41, Kyle Richh, Jenn Carter, TaTa | 68.0000         | 2023-11-17               |





## 6 .Which song has the highest energy score in the High Popularity dataset

```
SELECT
    track_album_name, MAX(energy) AS Highest_energy_score
FROM
    music.high_popularity_spotify_data
GROUP BY track_album_name
ORDER BY Highest_energy_score DESC
LIMIT 1;
```

Result Grid

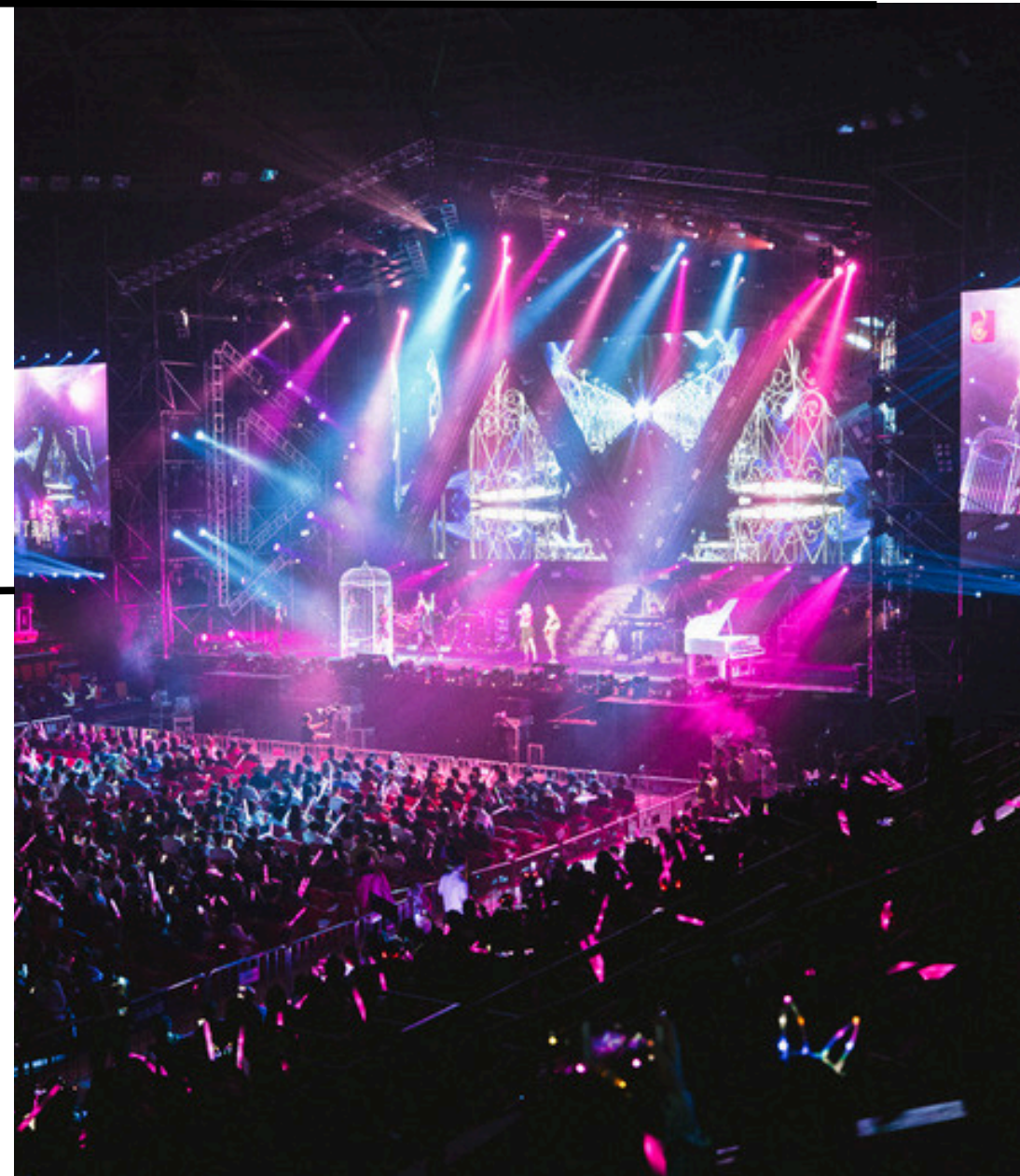


Filter Rows:

Export:



|   | track_album_name  | Highest_energy_score |
|---|-------------------|----------------------|
| ▶ | 1984 (Remastered) | 0.978                |



# Conclusion

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This portfolio demonstrates the process of cleaning, transforming, and analyzing Spotify data. By asking engaging and relatable questions, we provide insights into popular genres, and artists' performance. The queries are designed to appeal to a broad audience, showcasing the power of SQL in analyzing real-world datasets. Whether you're a music enthusiast or a data professional, these findings offer a glimpse into the fascinating world of Spotify data.