Database Design for Spotify

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Introduction:

In a world where technology and music are inseparably reliability requires seamless access to accurate data. connected, our initiative intends to combine user information with Spotify playlists, providing a unique promote indie artists. Data analysts require real-time insight into their music choices, subscriptions, and more access to millions of songs' streams, likes, and user experience. Join us as we mix data science and music to change the way we discover and enjoy music on Spotify.

system outages. Maintaining user trust and platform

This approach is intended to transform music suggestions feedback. Using Excel, this becomes a time-consuming and curated playlists by adapting them to users' preferences operation rife with errors and delays. Spotify, on the other in genres, artists, albums, and more. Throughout this paper, hand, can immediately discover trending indie music, push we will look at the system's architecture, techniques, and them to suitable users, and track the campaign's the significant impact it has on improving the Spotify effectiveness in real-time, ensuring that new artists receive the attention they deserve.

Problem Statement:

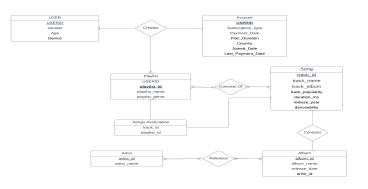
As a leading music streaming network, Spotify handles huge volumes of data, such as song details, artist profiles, user preferences, and playback history. Simpler tools, such as Excel, may have sufficed at first. However, as the platform's reach and offers developed, the limitations of Excel became clear: concerns with data size restrictions. difficulties preserving data integrity, and difficulties handling real-time concurrent access. As a result, a more robust database solution is required.

How can we ensure effective, scalable, and reliable data handling given the huge and intricate nature of Spotify's music data? Why is it critical for a music streaming behemoth like Spotify to transition from Excel to a dedicated database system?

Significance of the Problem:

Data influences user experience for a platform like Spotify. Songs Association (track_id, playlist_id) Inadequate data management can result in bad song Song (track_id, track_name, track_album, recommendations, inaccurate artist royalties, and even

ER Diagram:



Relational Schema:

User (USERID, Gender, Age, Device) Account (USERID, Subscription Type, Plan Duration, Country, Joined Date, Last Payment Date) **Playlist** (USERID, playlist id, playlist name, playlist genre)

track popularity, duration ms, release year, danceability)

Album (album id, album name, release date, artist id) Artist (artist id, artist name)

Issues Faced:

Normalization Issues: We ensured that the database is normalized to reduce redundancy and maintain data integrity was challenging. Checked if the tables were in BCNF and if not converted them into one.

Relationships Between Tables: Establishing maintaining relationships between tables, such as foreign key constraints, was complex. It's crucial to ensure that data consistency is maintained when performing operations like inserts, updates, and deletes. To combine the track id and playlist id to get track id as the primary key in the Song Table we created a separate table called Song Insert Query2: Association.

consists of track id and playlist id and doesn't have any unique value. So it is difficult to update the table as we need to update the value everywhere.

Data Validation: Validating user inputs and ensuring that the data entered into the tables adhere to the defined data types and constraints is crucial. Incomplete or inconsistent data can lead to errors and affect the reliability of your database.

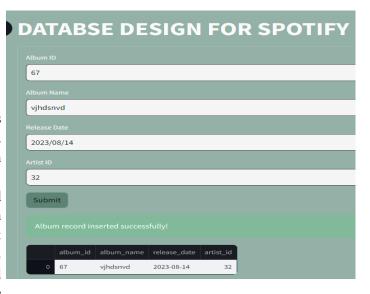
Integration Challenges: Integrating the database with the streamlit app was a bit challenging.

User Experience: Designing a database that supports an optimal user experience involves considering how data will be queried and displayed in the application. Balancing the Delete Queries: need for comprehensive data with the requirement for quick and efficient queries is often challenging.

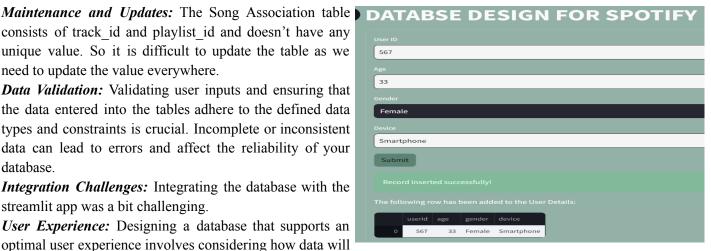
SQL Queries:

Insert Query1:

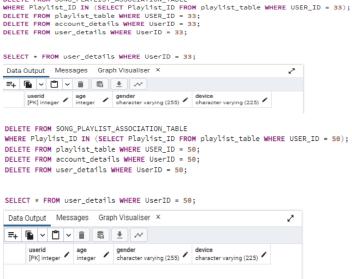
INSERT INTO album_table (Album_ID, Album_Name, Release_Date, Artist_ID) delete from Song_PlayList_Association_table VALUES (%s, %s, %s, %s)



INSERT INTO user_details (UserID, Age, Gender, Device) VALUES (%s, %s, %s, %s)



DELETE FROM SONG_PLAYLIST_ASSOCIATION_TABLE



Update Query1:

UPDATE user_details SET Age = %s WHERE UserID = %s'



Update Query2:

'UPDATE artist SET Artist_Name = %s WHERE Artist_ID = %s

DATABSE DESIGN FOR SPOTIFY Artist ID 98 New Artist Name Sree Submit Artist name updated successfully! artist_name artist_id 0 Sree 98

Select queries:

1. Retrieve all the playlists that a specific user owns, you can use the following query:

SELECT * FROM playlist_table WHERE USER_ID = '23'



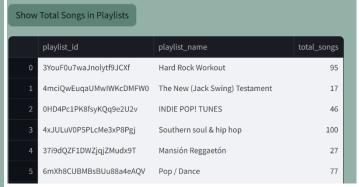
DATABSE DESIGN FOR SPOTIFY Enter User ID: 23 Search Playlists playlist_id playlist_name playlist_genre user_id 0 37i9dQZF1DWXrVH01e3PIE Trapperz Argentina rap 23 1 2YPP7fiYu5pIcp2yyHvw4A Neo Soul r&b 23

2. Retrieve the total number of songs in each playlist along with the playlist details.

SELECT p.Playlist_ID, p.Playlist_Name, COUNT(sp.Track_ID) as Total_Songs
FROM playlist_table p
JOIN SONG_PLAYLIST_ASSOCIATION_TABLE sp ON p.Playlist_ID = sp.Playlist_ID
GROUP BY p.Playlist_ID;

	playlist_id [PK] character varying (255)	playlist_name character varying (255)
1	3YouF0u7waJnolytf9JCXf	Hard Rock Workout
2	4mciQwEuqaUMwIWKcDMF	The New (Jack Swing) Testament
3	0HD4Pc1PK8fsyKQq9e2U2v	INDIE POP! TUNES
4	4xJULuV0P5PLcMe3xP8Pgj	Southern soul & hip hop
5	37i9dQZF1DWZjqjZMudx9T	Mansión Reggaetón

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3. Retrieve the top 5 most popular songs with their details.

SELECT * FROM song_details_table ORDER BY TRACK_POP DESC LIMIT 5;

	track_id [PK] character varying (255)	track_name character varying (255)	track_album_id character varying (255)	track_pop /	duration_ms /	release_year integer	danceability double precision
1	2XU0oxnq2qxCpomAAuJY8K	Dance Monkey	0UywfDKYlyiu1b38DRrzYD	100	209438	2019	0.824
2	696DnlkuDOXcMAnKITgXXK	ROXANNE	6HJDrXs0hpebaRFKA1sF90	99	163636	2019	0.621
3	7k4t7uLgt0xPwTpFmtJNTY	Tusa	7mKevNHhVnZER3BLgI80	98	200960	2019	0.803
4	2b8fOow8UzyDFAE27YhOZM	Memories	3nR9B40hYLKLcR0Eph3Goc	98	189486	2019	0.764
5	Osf12qNH5qcw8qpgymFOqD	Blinding Lights	2ZfHkwHuoAZrlz7RMj0PDz	98	201573	2019	0.513



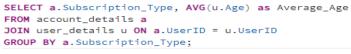
6. Retrieve Users and Their Subscription Status:

Female

5 Male



4. Retrieve the average age of users for each subscription type.



	subscription_type character varying (255)	average_age numeric
1	Basic	39.3707865168539326
2	Premium	38.0923076923076923
3	Standard	38.2500000000000000

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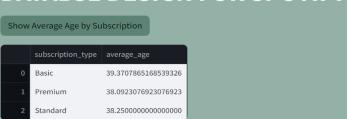
51

Standard

Basic 33



DATABSE DESIGN FOR SPOTIFY 7. Identify Artists with Most Albums Released:



SELECT ar.Artist_Name, COUNT(al.Album_ID) as Total_Albums FROM artist ar JOIN album_table al ON ar.Artist_ID = al.Artist_ID GROUP BY ar.Artist_Name ORDER BY Total_Albums DESC;

	artist_name character varying (255)	total_albums bigint
1	Martin Garrix	72
2	Dimitri Vegas & Like Mike	64
3	Hardwell	61
4	David Guetta	51
5	The Chainsmokers	51

5. Retrieve the artists who have albums released in the last year and the total number of songs in each album.



	artist_name character varying (255)	â	album_id character varying (255)	total_songs bigint	â
1	Kalli		6W1oIR6Hh8K4gWwmeMe		1
2	CLRFL		0Z3TDNVXjeTKm8KvE0F6		1
3	Critic City		20dJlmVbsCHRdqd3lNZ3Sa		1
4	NGHTMRE		1jd4dq1xm8me7AvD2EHW		1
5	Gregory Porter		045wnruBljG8IPdPbJHSVh		1

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8. Identify the Most Subscribed Country:

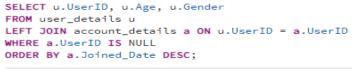
SELECT Country, COUNT (*) as Total_Subscriptions
FROM account_details
GROUP BY Country
ORDER BY Total_Subscriptions DESC
LIMIT 1;
country total subscriptions

:	country character varying (255)	total_subscriptions bigint
1	Canada	23

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9. Find the Latest Joined Users Without a Subscription:



		userid [PK] integer	age integer	•	gender character varying (255) ✔
;	1	224		30	Female

Query Execution Analysis:

OUERY PLAN

Group Key: p.playlist_id

Problematic Query 1: Retrieve the total number of songs in each playlist along with the playlist details.

Optimize query performance by addressing sequential scans on song_playlist_association_table and playlist_table through index creation on playlist_id columns. Consider switching to INNER JOIN for improved efficiency and evaluate potential gains from materialized views, especially for large datasets. Utilize EXPLAIN ANALYZE to pinpoint specific bottlenecks and fine-tune optimizations accordingly.

```
-- Explain
-- Query 2

EXPLAIN SELECT p.Playlist_ID, p.Playlist_Name, COUNT(sp.Track_ID) as Total_Songs

FROM playlist_table p

OSNG_PLAYLIST_ASSOCIATION_TABLE sp ON p.Playlist_ID = sp.Playlist_ID

GROUP BY p.Playlist ID:
```

-> Seg Scan on song_playlist_association_table sp (cost=0.00.,436.80 rows=10080 width=2.



10. List Users and Their Favorite Genre (based on most added songs):

```
SELECT u.UserID, MAX(p.Playlist_Genre) as Favorite_Genre
FROM user_details u
JOIN playlist_table p ON u.UserID = p.USER_ID
JOIN SONG_PLAYLIST_ASSOCIATION_TABLE s ON p.Playlist_ID = s.Playlist_ID
GROUP BY u.UserID;
```

	userid [PK] integer	favorite_genre text
1	18	rock
2	64	rock
3	110	latin
4	178	rock
5	55	rap

Problematic Query 2: Retrieve the artists who have albums released in the last year and the total number of songs in each album.

-> Seg Scan on playlist_table p (cost=0.00..10.60 rows=60 width=1032)

HashAggregate (cost=525.57..526.17 rows=60 width=1040)

-> Hash Join (cost=11.35..475.17 rows=10080 width=1150)

Hash Cond: ((sp.playlist_id)::text = (p.playlist_id)::text)

-> Hash (cost=10.60..10.60 rows=60 width=1032)

Query inefficiencies are indicated by sequential scans on song_details_table and album_table, highlighting the need for indexes on track_album_id and album_id. The hash join may pose a bottleneck, suggesting potential gains from converting to an INNER JOIN. The HashAggregate step, grouping by ar.artist_name and al.album_id, may be resource-intensive, warranting consideration of indexing. To address these issues, implement indexes, consider INNER JOIN optimization, and fine-tune with EXPLAIN ANALYZE for performance analysis.

```
-- Query 5

EXPLAIN SELECT ar.Artist_Name, al.Album_ID, COUNT(sd.Track_ID) as Total_Songs

FROM artist ar

JOIN album_table al ON ar.Artist_ID = al.Artist_ID

JOIN song_details_table sd ON al.Album_ID = sd.Track_Album_ID

WHERE al.Release_Date >= CURRENT_DATE - INTERVAL '5 year'

GROUP BY ar.Artist_Name, al.Album_ID;
```

DATABSE DESIGN FOR SPOTIFY

0

	QUERY PLAN text
1	HashAggregate (cost=824.30831.23 rows=693 width=1040)
2	Group Key: ar.artist_name, al.album_id
3	-> Hash Join (cost=375.00819.10 rows=693 width=1548)
4	Hash Cond: (al.artist_id = ar.artist_id)
5	-> Hash Join (cost=248.76691.03 rows=693 width=1036)
6	Hash Cond: ((sd.track_album_id)::text = (al.album_id)::text)
7	-> Seq Scan on song_details_table sd (cost=0.00436.80 rows=2080 width=10
8	-> Hash (cost=242.46242.46 rows=504 width=520)
9	-> Seq Scan on album_table al (cost=0.00242.46 rows=504 width=520)
10	Filter: (release_date >= (CURRENT_DATE - '5 years'::interval))
11	-> Hash (cost=109.44109.44 rows=1344 width=520)
12	-> Seq Scan on artist ar (cost=0.00109.44 rows=1344 width=520)

Problematic Query 3: List Users and Their Favorite Genre (based on most added songs):

The third query's execution plan reveals potential sequential inefficiencies with scans song playlist association table playlist table, and indicating a need for indexes on playlist id. The hash join might be a bottleneck, suggesting gains from converting to an INNER JOIN. The HashAggregate step, grouping by may be resource-intensive, u.userid. warranting consideration of indexing. To address these issues, implement indexes, consider INNER JOIN optimization, and fine-tune with EXPLAIN ANALYZE for performance analysis.

```
EXPLAIN SELECT u.UserID, MAX(p.Playlist_Genre) as Favorite_Genre

FROM user_details u

JOIN playlist_table p ON u.UserID = p.USER_ID

JOIN SONG_PLAYLIST_ASSOCIATION_TABLE s ON p.Playlist_ID = s.Playlist_ID

GROUP BY u.UserID;
```

	QUERY PLAN text
1	HashAggregate (cost=565.23566.03 rows=80 width=36)
2	Group Key: u.userid
3	-> Hash Join (cost=23.15514.83 rows=10080 width=122)
4	Hash Cond: (p.user_id = u.userid)
5	-> Hash Join (cost=11.35475.17 rows=10080 width=122)
6	Hash Cond: ((s.playlist_id)::text = (p.playlist_id)::text)
7	-> Seq Scan on song_playlist_association_table s (cost=0.00436.80 rows=10080 width=1
8	-> Hash (cost=10.6010.60 rows=60 width=638)
9	-> Seq Scan on playlist_table p (cost=0.0010.60 rows=60 width=638)
10	-> Hash (cost=10.8010.80 rows=80 width=4)
11	-> Seq Scan on user_details u (cost=0.0010.80 rows=80 width=4)

Future Scope:

Podcast Integration: Expand the database to support podcasts and audiobooks, catering to a broader range of audio content preferences.

Enhanced Recommendation System: Invest in machine learning algorithms to improve the recommendation system, providing more accurate and personalized music suggestions to users.

Social Features: Introduce social elements such as friend connections, shared playlists, and collaborative playlist creation to enhance user engagement and community interaction.

Offline Mode and Cross-Device Syncing: Implement offline mode, allowing users to download playlists for offline listening, and ensure seamless syncing across different devices.

User Feedback and Ratings: Incorporate a user feedback system, enabling users to rate songs and playlists, providing valuable data for refining recommendations and enhancing user experience.

Conclusion:

In summary, the database architecture devised for Spotify establishes a strong foundation for an enriched music streaming platform. The meticulously organized relational schema adeptly handles user information, subscription details, playlists, and music content. Moving forward, enhancing the user experience could involve integrating podcasts, fine-tuning the recommendation system, and introducing social functionalities. Addressing challenges like maintaining data security and privacy remains an ongoing priority. Altogether, the project shows potential in molding Spotify into a versatile and user-focused platform, ready to evolve in tandem with industry shifts and user preferences.

References:

- Music Recommendation System using Spotify
 Dataset | Kaggle
- <u>Database Design for Spotify.</u> In this series of <u>Database Designs</u>, I... | by <u>Ayush Dixit | Towards</u> <u>Data Engineering | Medium</u>