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DATA VISUALIZATION PROJECT

https://github.com/nada912/Tracks-Record

Start and run the project

Prerequisites:

Have a data bricks account (If not, create an account with your professional email on_databricks).

Step 1: Create a cluster on Databricks

On the menu bar, click on "Compute" then "Create cluster" button.

Choose a name then create.

Step 2: Import the data to Databricks

On the menu bar, click on "Catalog", then "Create table" button. Name the target directory "data" then upload all the text files at once.

Step 3: Open the notebook

On the menu bar, click "New" -> "Notebook".

Once the notebook is created, click on "File" -> "Import" to import the notebook available on Github.

Now execute the cells to get and process the data then load it to the database.

Step 4: At this point everything is done, you can visualize the dashboard on PowerBi.

Connect To PowerBi and select "Get Data from other source"

Select on the left sidebar "Database" -> "PostgreSQL Database"

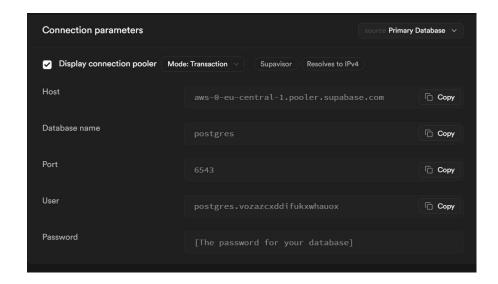
Connect with the following and open the Report Tracks.pbix file:

Host: http://aws-0-eu-central-1.pooler.supabase.com

Database name: postgre

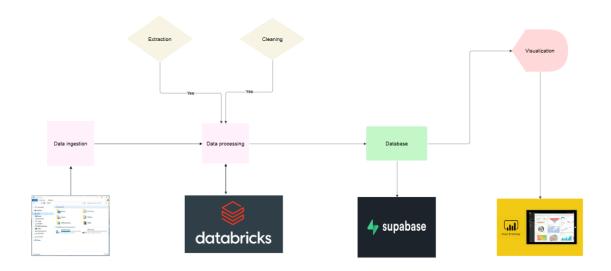
user: postgres.vozazcxddifukxwhauox password: PasswordTracks2024.

Reconnect to the database in case you have trouble visualizing the dashboard



Conception and implementation

Process and tools:



To achieve this project, we have gone through several steps and used some tools in order to first retrieve the data, clean it, load it to the database and finally visualize it.

Data ingestion:

Given the context of the project, the data was available on Teams, so we downloaded the files locally on our computers.

Data processing:

For this part we used *Databricks*, which is a powerful platform for big data processing, allowing us to clean, transform, and process our datasets efficiently.

It supports distributed computing, enabling scalability for complex operations on large amounts of data and it greatly integrates various data sources including databases.

Loading Data:

The data was initially loaded from text files into our Resilient Distributed Dataset (RDD) using the *WholeTextFiles* method. This method reads all files from a specified directory containing the .txt files and returns a pair consisting of the file path and the content of the file.

```
In [0]:

rdd =sc.wholeTextFiles("/FileStore/tables/tracks-data/AkaGambit.csv',
    "置美詩郎,,Blaze of the Soul Reaper,09 Jan 2008 06:23\n置美詩郎,,Battle Ignition,09 Jan 2008 06:21\nEvanescence,Fallen,Taking
Over Me,22 Dec 2007 07:23\nEvanescence,Fallen,Imaginary,22 Dec 2007 07:19\nEvanescence,Fallen,Tourniquet,22 Dec 2007 07:14\nE
vanescence,Fallen,Haunted,22 Dec 2007 07:11\nEvanescence,Fallen,My Immortal,22 Dec 2007 07:07\nEvanescence,Fallen,Everybody's
Fool,22 Dec 2007 07:04\nEvanescence,Fallen,Bring Me to Life,22 Dec 2007 07:00\nLikin Park,Minutes to Midnight,Given Up,21 De
c 2007 08:33\nLinkin Park,Minutes to Midnight,Wake,21 Dec 2007 03:13\nD. Gray Man (Aka Gambit Rip),,Opening 3,17 Dec 2007 20:09\nD.
Gray Man (Aka Gambit Rip),,Opening 3,17 Dec 2007 20:09\nD. Gray Man (Aka Gambit Rip),,Opening 3,17 Dec 2007 20:09\nD.
Gray Man (Aka Gambit Rip),,Opening 3,17 Dec 2007 20:09\nD. Gray Man (Aka Gambit Rip),,Opening 3,17 Dec 2007 20:09\nD. Gray Man (Aka Gambit Rip),,Opening 3,17 Dec 2007 20:09\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 20:09\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 20:09\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 20:09\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 20:09\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 19:59\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 19:59\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:28\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:24\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:12\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:14\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:16\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:10\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:10\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:10\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:10\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:10\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:10\nD. Gray Man (Aka Gambit Rip),Opening 3,17 Dec 2007 18:10\
```

Data Extraction:

The raw data within the RDD was processed to extract the listener names from the file paths. Additionally, the content was split into individual records, each representing a unique music listening event. Each record was parsed into five distinct fields: listener name, artist/band, album, track, and date.

```
n [0]:
        from pyspark.sql import SparkSession
        from pyspark.sql import Row
        rdd_{transformed} = rdd.map(lambda e: (e[0].split("/")[-1].split(".")[0], e[1])) \setminus
                               .flatMapValues(lambda v: v.split("\n"))
                                .map(lambda e: (e[0], e[1].split(","))) \
                                .map(lambda e: (
                                                                          # Listener name
                                    e[0],
                                    e[1][0].
                                                                         # Artist name
                                    e[1][1] if len(e[1]) > 1 else None, # Album name (if it exists)
                                    e[1][2] if len(e[1]) > 2 else None, # Track name (if it exists)
e[1][3] if len(e[1]) > 3 else None # Date (if it exists)
        columns = ["listener", "artist / band", "album", "track", "date"]
        # create a structured PySpark DataFrame with our specified schema for processing
        df = spark.createDataFrame(rdd transformed, schema=columns)
        display(df)
```

Here is a preview of the result we get after the first transformation:

date	track	album	artist / band	listener
09 Jan 2008 06:23	Blaze of the Soul Reaper		想美慧	AkaGambit
09 Jan 2008 06:21	Battle Ignition		和結果營	AkaGambit
22 Dec 2007 07:23	Taking Over Me	Fallen	Evanescence	AkaGambit
22 Dec 2007 07:19	Imaginary	Fallen	Evanescence	AkaGambit

Date Column Processing:

The original "date" column was split into two separate columns: listened_date and hour. This transformation allowed for more granular analysis of the listening behavior, separating the date from the time of the event.

```
In [0]:
    from pyspark.sql import functions as F

    df_formatted = df.withColumn(
        "listened_date",
        F.date_format(F.to_date(F.col("date"), "dd MMM yyyyy HH:mm"), "dd/MM/yy")
).withColumn(
        "hour",
        F.date_format(F.to_timestamp(F.col("date"), "dd MMM yyyyy HH:mm"), "HH:mm")
)
display(df_formatted)
```

This is what we get:

A ^B _C track	A ^B C date
Opening 3	17 Dec 2007 20:09
Opening 3	17 Dec 2007 20:08
Opening 3	17 Dec 2007 20:06
Opening 3	17 Dec 2007 20:05
Opening 3	17 Dec 2007 20:03
Opening 3	17 Dec 2007 20:02
Opening 3	17 Dec 2007 20:00
Opening 3	17 Dec 2007 20:00
Opening 3	17 Dec 2007 19:59
Opening 3	17 Dec 2007 19:58
Opening 3	17 Dec 2007 18:28
Opening 3	17 Dec 2007 18:27
Opening 3	17 Dec 2007 18:25
Opening 3	17 Dec 2007 18:24

$\mathbb{A}^{\mathbb{B}}_{\mathbb{C}}$ track	ABc listened_date	ABC hour
Opening 3	17/12/07	20:09
Opening 3	17/12/07	20:08
Opening 3	17/12/07	20:06
Opening 3	17/12/07	20:05
Opening 3	17/12/07	20:03
Opening 3	17/12/07	20:02
Opening 3	17/12/07	20:00
Opening 3	17/12/07	20:00
Opening 3	17/12/07	19:59
Opening 3	17/12/07	19:58
Opening 3	17/12/07	18:28
Opening 3	17/12/07	18:27
Opening 3	17/12/07	18:25
Opening 3	17/12/07	18:24

Handling Null Values:

Empty cells in the dataset were set to null, ensuring that missing values were appropriately handled. The original date column was then dropped as its information had been separated into the new listened_date and hour columns.



Removing Duplicates and Empty Rows:

As can be seen below, there were quite a bit of duplicates in the dataset. We then had to make sure that any rows that were redundant or completely empty were removed from the dataset to ensure data integrity and reduce potential noise in the analysis.



Final processed database

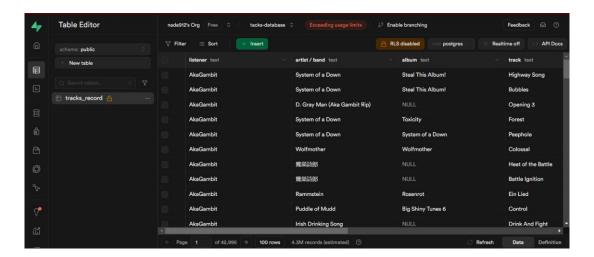


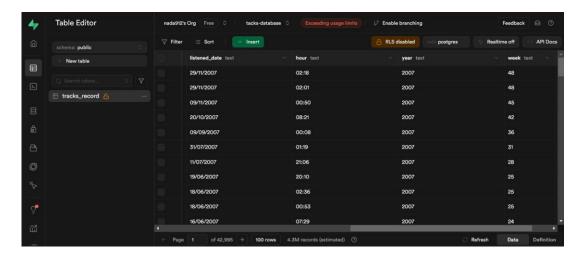
Database Integration:

The cleaned and processed data was then connected to a PostgreSQL database hosted on Supabase. Here's the Identifiers to connect to our postgre database in supabase and populate it.

```
jdbc_url = "jdbc:postgresql://aws-0-eu-central-1.pooler.supabase.com:6543/postgres"
properties = {
    "user": "postgres.dbwrzjfnxqhllunphygs",
    "password": "PasswordTracks2024.",
    "driver": "org.postgresql.Driver"
}

df_formatted.write.jdbc(url=jdbc_url, table="public.tracks_record", mode="overwrite", properties=properties)
```





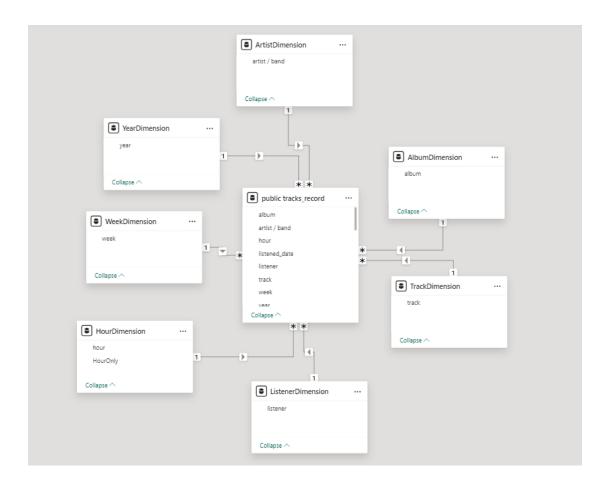
PowerBi Dashboard conception

The star model

In this project, we used a star schema to structure our data model, enhancing query performance and enabling efficient reporting. The star schema consists of a central fact table here called "public tracks_record" surrounded by multiple dimension tables, which simplifies data relationships and improves readability. Each table is smaller and easier to join, which makes it easier to filter and query our data.

Therefore, the dimensions table here are:

- Listener dimension
- Artist dimension
- Album dimension
- Track dimension
- Hour dimension
- Week dimension
- Year dimension



The requested KPIs

Most listened track of all time
Most listened track for each week
Most listened album of all time
Most listened album for each week
Cross tabulation of the number of tracks listened to by listener and by artist
Ranking the 10 biggest listeners of all time
Ranking the 10 biggest listeners for each week

Additional KPIs

For additional KPIs, we added:

- Top 10 most listened artists overall & per listener
- Top 10 most listened artists per year
- Top 10 most listened artists per hour
- Total of plays of the most listened track
- Total of plays of the most listened album
- Top 10 most active hours overall & per listener

Dashboard:

https://app.powerbi.com/groups/me/reports/02389710-c320-490c-a63a-f2f4d7b99d63/aaec84d08186c3f7547d?experience=power-bi

Principal KPIs



Additional KPIs



Troubles:

We opted for a direct connection to the database stored in supabase but in the free version we had some problems with the available resources so that we could keep our data updated.

