

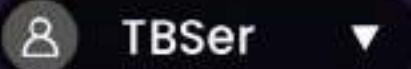


Spotify

Music Trends & Popularity Analysis



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Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Welcome

Transforming 12,000+ tracks of Spotify data into actionable business intelligence insights.

Play

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[Welcome](#)[About Us](#)[Our Team](#)[Agenda](#)[Data Preparation](#)[Dimensional Modeling](#)[Dashboard Development](#)[Insight Generation](#)[Recommendations](#)

Introduction

About Us

+8 ...

Business Description

The music streaming industry is one of the most data-driven digital sectors, where competition is strongly influenced by the ability to understand listener preferences and predict content performance.

This Business Intelligence project aims to analyze Spotify music data to identify the key factors that drive song popularity using dimensional modeling, dashboards, and insight generation.

Our Company

Spotify relies extensively on data analytics to optimize recommendations, curate playlists, and support both artists and business partners.

The objective is to transform raw music data into actionable insights that explain popularity trends and support strategic decision-making related to content curation, playlist design, and artist promotion.

[Welcome](#)[About Us](#)[Our Team](#)[Agenda](#)[Data Preparation](#)[Dimensional Modeling](#)[Dashboard Development](#)[Insight Generation](#)[Recommendations](#)

Our Team



Mariem Chammem



Nouha Boukhris



Sarra Alioua



Elaa Azouzi



Welcome



About Us



Our Team



Agenda



Agenda

Analysis Workflow



...

#	Phase	Tasks	Deliverables
1	Data preparation	Clean and transform 12,000+ tracks	Star schema model with fact and dimension tables
2	Dimensional Modeling	Design FactTracks and 4 dimension tables	Optimized data structure for analysis
3	Dashboard Development	Create Executive, Deep Dive, Artist dashboards	Interactive Power BI dashboards
4	Insight Generation	Analyze audio features vs popularity	10 key findings with business interpretation
5	Recommendations	Develop data-driven strategies	7 actionable business recommendations

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[Welcome](#)[About Us](#)[Our Team](#)[Agenda](#)

Data preparation

Project Metrics & Impact

[Data Preparation](#)[Dimensional Modeling](#)[Dashboard Development](#)[Insight Generation](#)[Recommendations](#)**12k+****Tracks Analyzed**Comprehensive
dataset coverage**6Y****Time Period**2015-2020 music
trends**8****Key Findings**Data-driven insights
identified**6****Recommendations**Actionable strategies
developed

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Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Data preparation

Project Metrics & Impact

1. Data Import

- Loaded Spotify dataset using pandas library.
- Configured with semicolon delimiter and latin1 encoding.
- Dataset contains 20 columns including track attributes and metadata.

```
df = pd.read_csv("Spotify.csv", sep=';', encoding="latin1")
```



Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Data preparation

Project Metrics & Impact

2. Data Formatting

- Rounded all numerical columns to 3 decimal places.
- Ensured consistency across all metrics.
- Improved readability for analysis.

```
float_cols = df.select_dtypes(include='float').columns  
df[float_cols] = df[float_cols].round(3)
```



Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Data preparation

Project Metrics & Impact

3. Missing Data Handling

- Identified 36 missing values in release_date column.
- Removed all rows with null values using dropna().
- Verified complete data cleanliness.

```
print(df.isna().sum())
df_clean = df.dropna()
```

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Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Data preparation

Project Metrics & Impact

4. Data Export

- `Spotify_data.csv` – Intermediate version
- `Spotify_data BI.csv` – Final cleaned version
- Ready for BI tools and visualization.

```
df_clean.to_csv('Spotify_data BI.csv', index=False)
```



Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Dimensional Modeling

A **star schema** dimensional model was implemented. The central fact table, **FactTracks**, stores quantitative measures such as popularity and audio feature values. It is connected to four dimension tables:

- **DimGenre**: genre classification
- **DimArtist**: artist information
- **DimTrack**: track-level metadata
- **DimDate**: time hierarchy (year, quarter, month)

This modeling approach improves query performance, enables flexible slicing and filtering, and aligns with BI best practices.

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Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



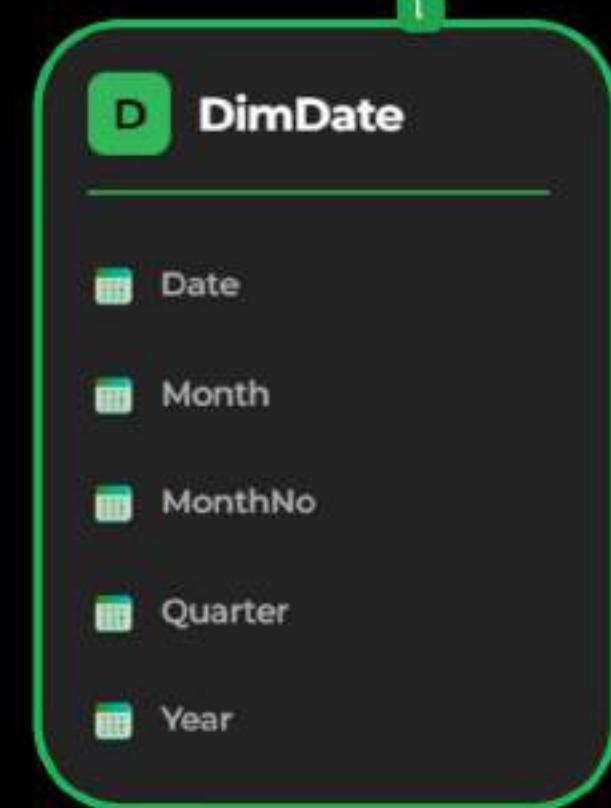
Dashboard Development



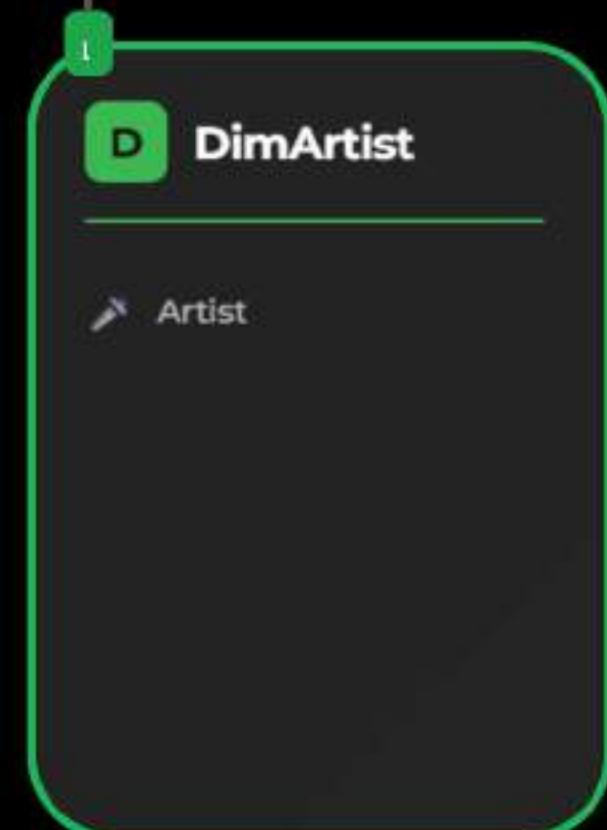
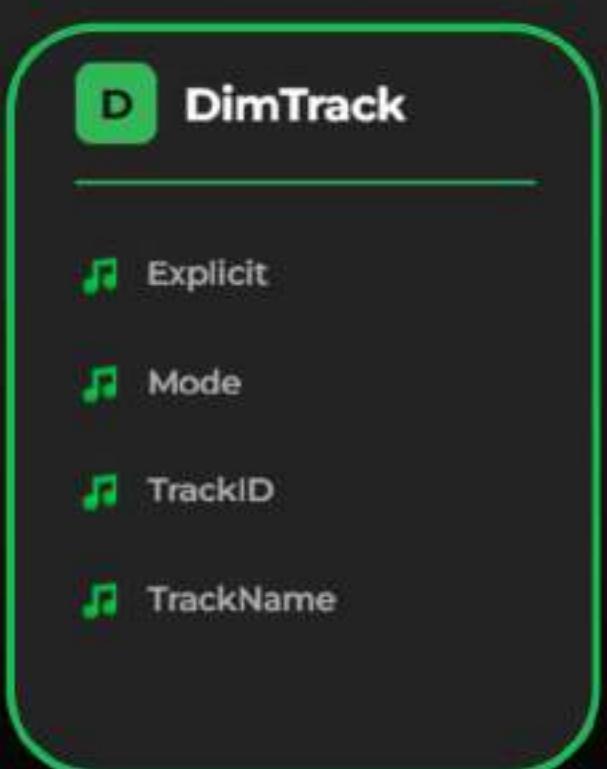
Insight Generation



Recommendations



Star schema



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Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Dashboard Development

Visual Insights from Our Analysis

Dashboard Highlights

- Executive Summary Dashboard - Overview of tracks, artists, and trends
- Deep Dive Dashboard - Audio features and genre analysis
- Artist Drill-Through Dashboard - Individual performance metrics
- Interactive visualizations with year-over-year comparisons
- Real-time filtering by genre, artist, and time period

Transforming Data into Actionable Insights

A comprehensive Power BI solution featuring three specialized dashboards, powered by custom DAX measures and interactive visualizations to explore 12,000+ tracks from 6,000 artists.





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Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

DAX Measures & Calculations

Performance Metrics

- Average Popularity
- Popularity Year-over-Year (YoY)
- Popularity Month-over-Month (MoM)

Audio Feature Metrics

- Average Energy
- Average Danceability
- Average Valence

Content Metrics

- Explicit Content Percentage
- Average Track Duration
- Total Tracks Count
- Number of Artists

Business Value

These measures enable dynamic analysis across time periods, genres, and artists, forming the foundation for actionable insights.

Technical Implementation: All measures are dynamically calculated and respond to filters and slicers, allowing users to explore data from multiple perspectives without pre-aggregation.

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Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Three-Tier Dashboard Architecture

Visual Insights from Our Analysis

Dashboard Highlights

Executive Summary Dashboard - Overview of tracks, artists, and trends

Deep Dive Dashboard - Audio features and genre analysis

Artist Drill-Through Dashboard - Individual performance metrics

Interactive visualizations with year-over-year comparisons

Real-time filtering by genre, artist, and time period



Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Executive Summary Dashboard

Key Visualizations:

- Strategic Overview Layer
- Total tracks & artist count
- Average track duration

- Genre distribution analysis
- Popularity trend lines
- High-level KPIs for decision-makers

Genre

- Acoustic Pop
- Alt Pop
- Alternative
- Alternative / Dream P...
- Alternative / Experim...
- Alternative / Indie

Artists

- ["Auli'i Cravalho", "Vai...
- ["Auli'i Cravalho"]
- ["DJ Sergey Mark'n"]
- ["Dj's Double Smile"]
- ["Dre'es", "Mia"]
- ["Freaky DJ's", "Coldw...

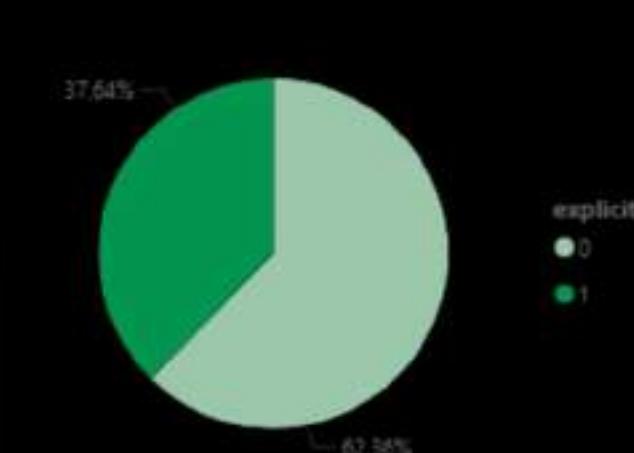
Year

- 2015
- 2016
- 2017
- 2018
- 2019
- 2020

6K

Number of Artists

Explicit Vs Non-Explicit Tracks

**12K**

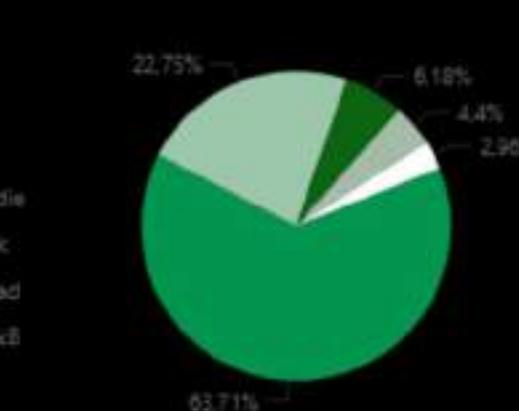
Total Tracks

Top 4 Genres by Avg Popularity

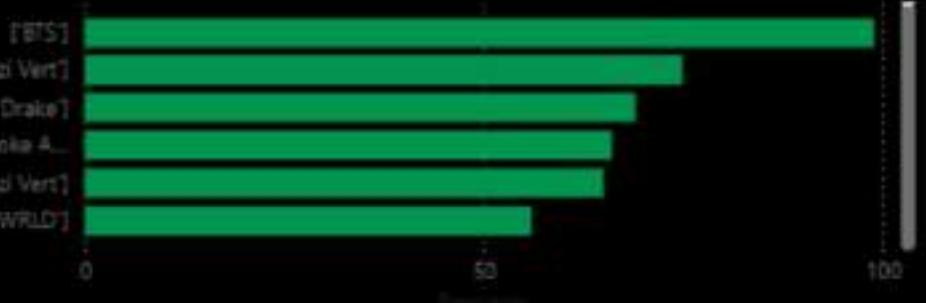
**3,46**

Avg Track Duration (min)

Share Of Tracks by Genre



Top Popular Artist





Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Deep Dive Dashboard

Correlation Analysis

Danceability vs Popularity:

Scatter plot reveals tracks with danceability between 0.6-0.8 achieve higher popularity scores (60-80 range). Sweet spot identified for streaming success.

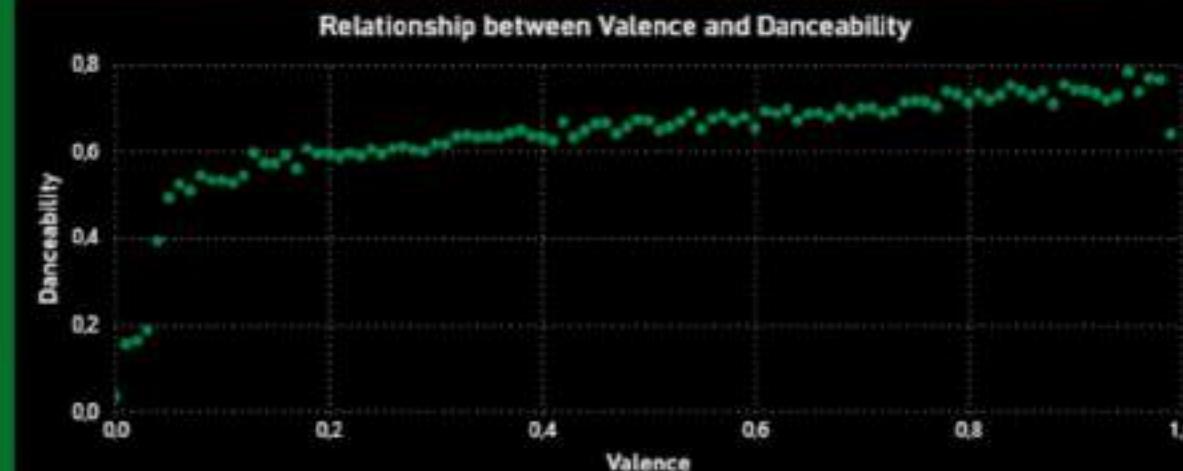
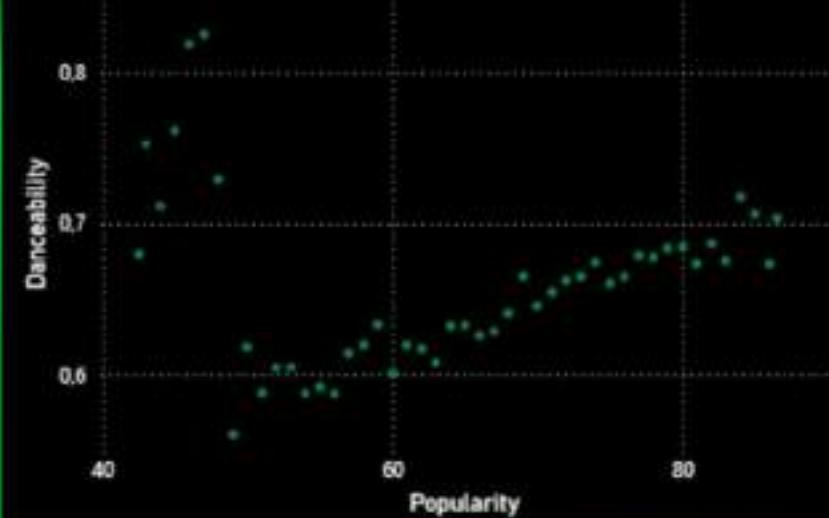
Valence vs Danceability:

Positive correlation shown - more danceable tracks tend to have higher valence (musical positiveness), indicating upbeat songs resonate better.

Artists
<input type="checkbox"/> ["Auli'i Cravalho", "Vai M..."]
<input type="checkbox"/> ["Auli'i Cravalho"]
<input type="checkbox"/> ["DJ Sergey Mark'n"]
<input type="checkbox"/> ["Dj's Double Smile"]
<input type="checkbox"/> ["Dre'es", 'Mia']
<input type="checkbox"/> ["Freaky DJ's", 'Coldway']

Genre	Average valence	Average danceability	Average energy
Worship	0.01	0.55	0.98
Tropical Pop	0.06	0.16	0.08
Trap-pop	0.04	0.37	0.00
Trap / Hip-Hop	0.60	0.62	0.72
Trap / EDM	0.35	0.72	0.72
Trap	0.41	0.74	0.60
Teen Pop	0.48	0.63	0.54
Synthwave Pop	0.71	0.69	0.62
Synth-pop	0.35	0.57	0.56
Synthpop	0.27	0.58	0.50
Soundscapes / Ambient	0.00	0.00	0.00
Rock / Pop Rock	0.18	0.44	0.41
Rock	0.39	0.60	0.71
Total	0.45	0.64	0.61

Relationship between Danceability and Track Popularity





Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Artist-Level Drill-Through Dashboard

Granular Artist Performance Analysis

Drill-Through Functionality:

Users click any artist name in Executive Summary or Deep Dive dashboards to navigate instantly to this detailed view, with all filter contexts preserved for seamless analysis.





Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Key Insights Generated

Popularity Trending Upward

Average popularity increased from ~60 to ~75 (2015-2020), reflecting improved algorithms and production standards.



Welcome

About Us

Our Team

Agenda

Key Insights Generated

Pop Genres Dominate

Pop-related genres account for 63.7% of tracks with highest average popularity scores.

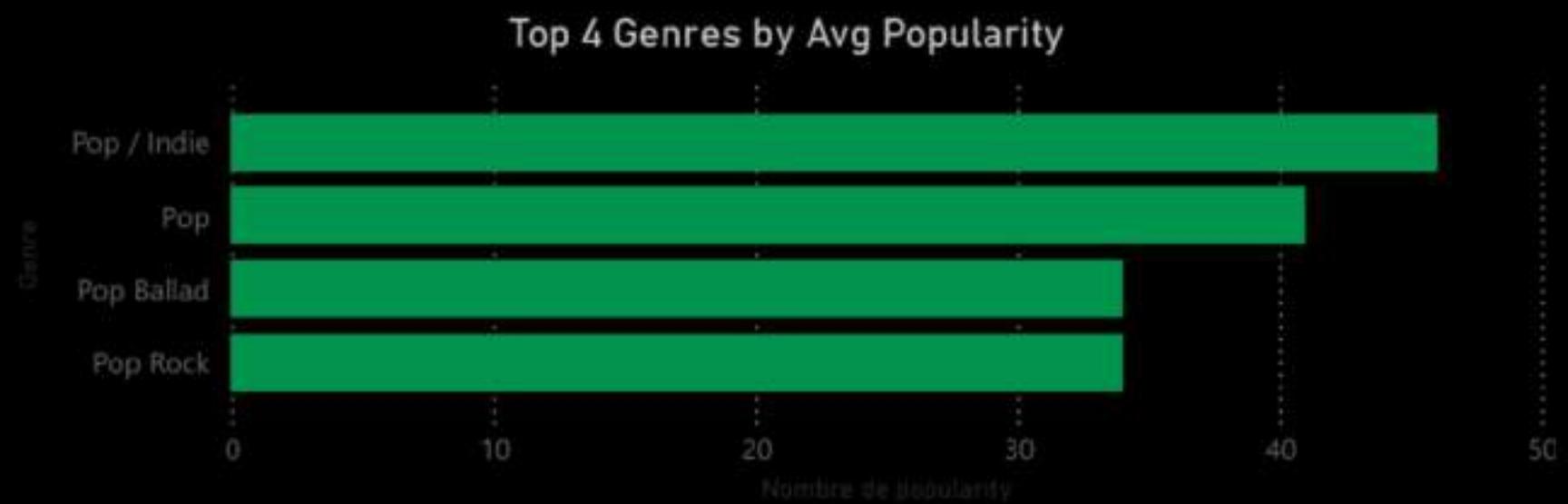
Data Preparation

Dimensional Modeling

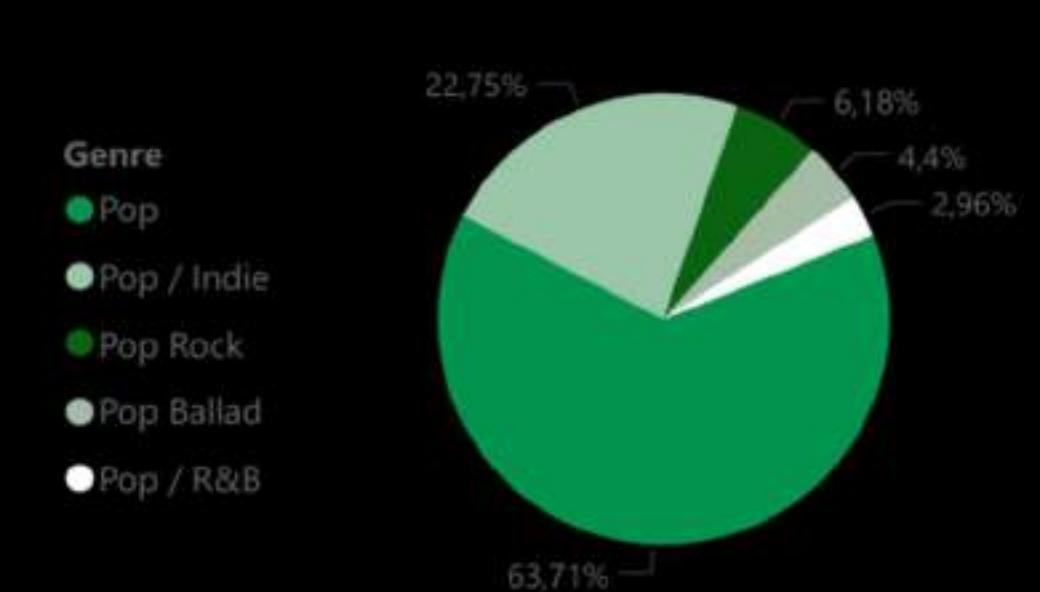
Dashboard Development

Insight Generation

Recommendations



Share Of Tracks by Genre





Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development

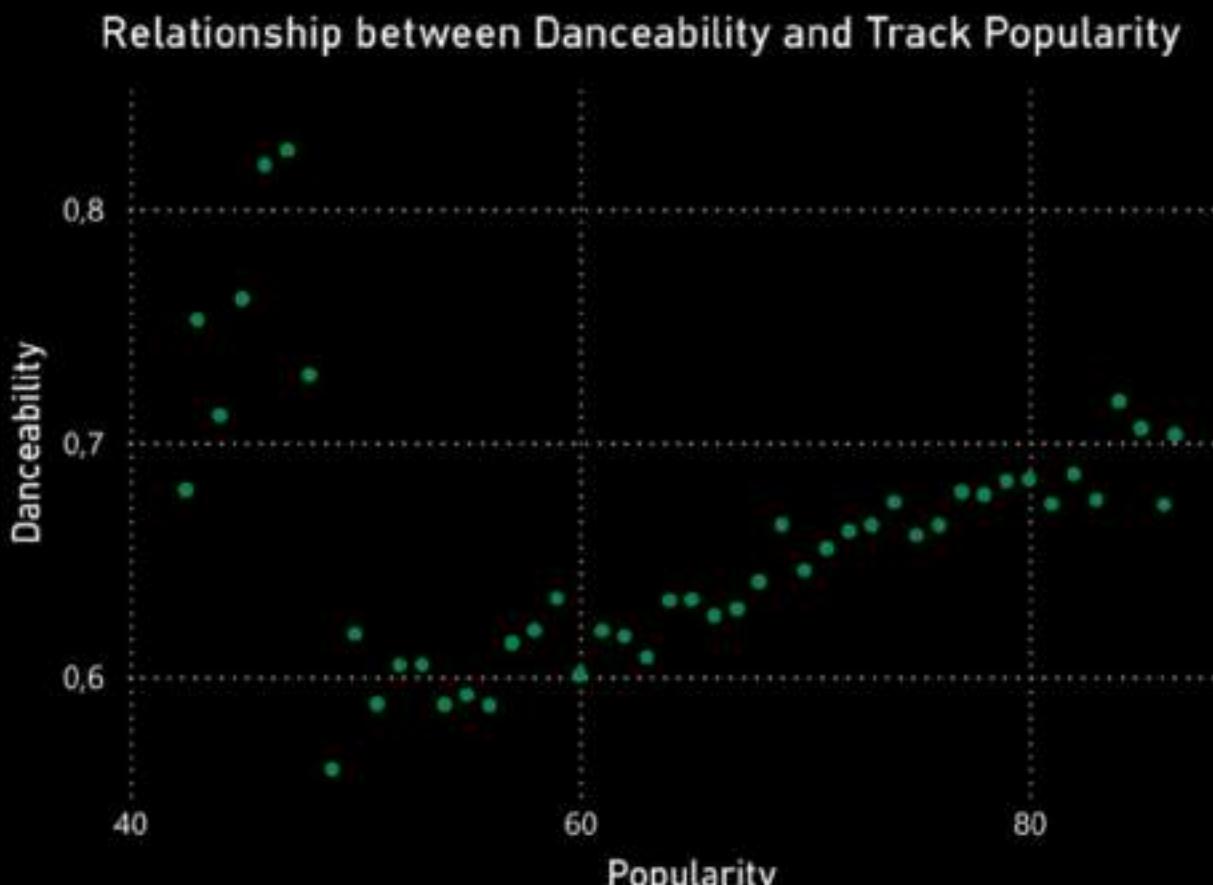


Insight Generation



Recommendations

Key Insights Generated



Danceability Drives Success

Strong positive correlation between danceability (0.6-0.8) and popularity (40-80).

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Key Insights Generated

Genre-Specific Audio Profiles

Audio features vary significantly across genres, revealing distinct musical characteristics:

Worship (0.98 energy) vs Tropical Pop (0.08 energy).

[Data Preparation](#)[Dimensional Modeling](#)[Dashboard Development](#)[Insight Generation](#)[Recommendations](#)

Genre	Average valence	Average danceability	Average energy
Worship	0,61	0,54	0,98
Tropical Pop	0,06	0,16	0,08
Trap-pop	0,04	0,07	0,00
Trap / Hip-Hop	0,60	0,62	0,72
Trap / EDM	0,35	0,73	0,72
Trap	0,41	0,74	0,60
Teen Pop	0,48	0,63	0,54
Synthwave Pop	0,71	0,69	0,62
Synth-pop	0,35	0,57	0,56
Synthpop	0,27	0,58	0,50
Soundscape / Ambient	0,00	0,00	0,00
Rock / Pop Rock	0,18	0,44	0,41
Rock	0,39	0,60	0,71
Total	0,45	0,64	0,61



Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development

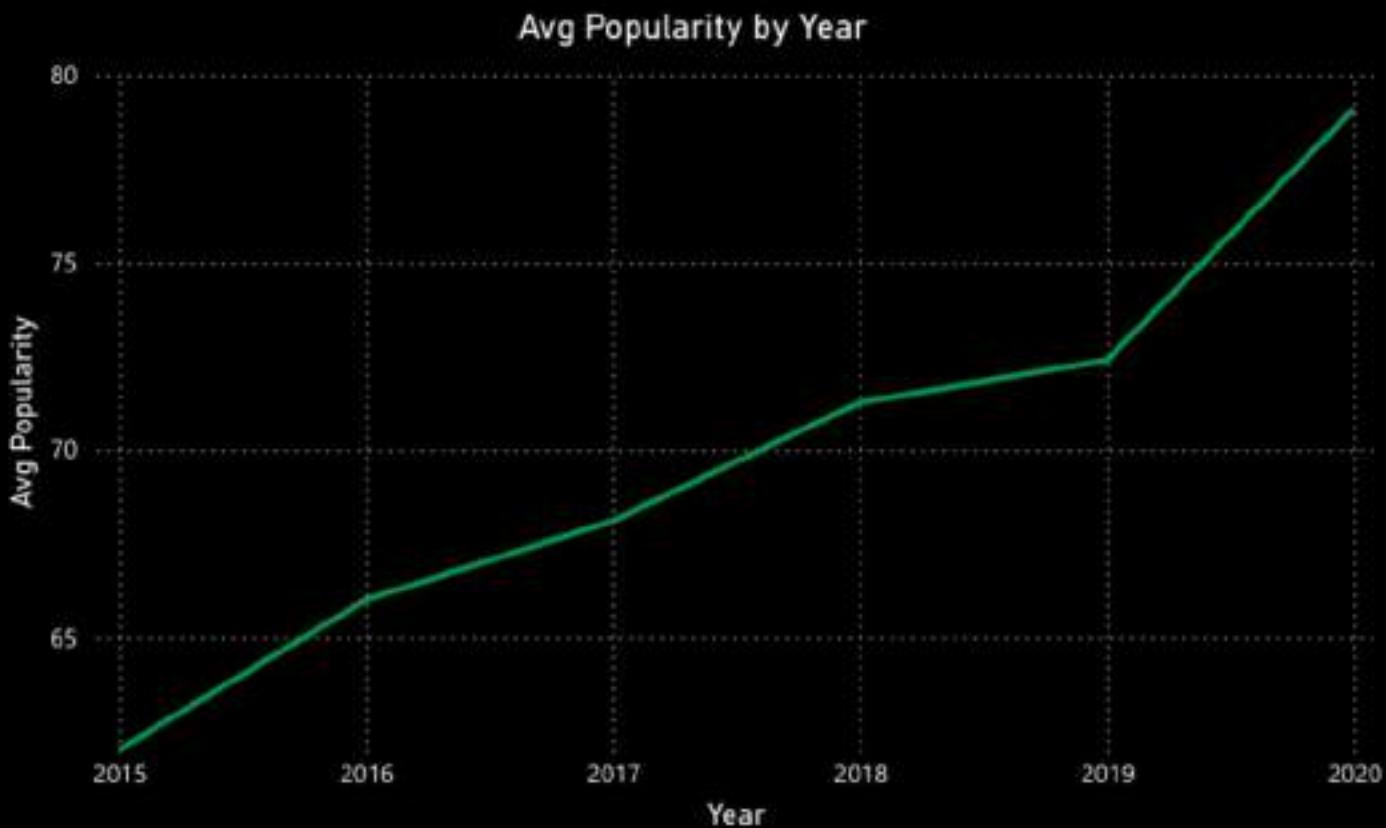
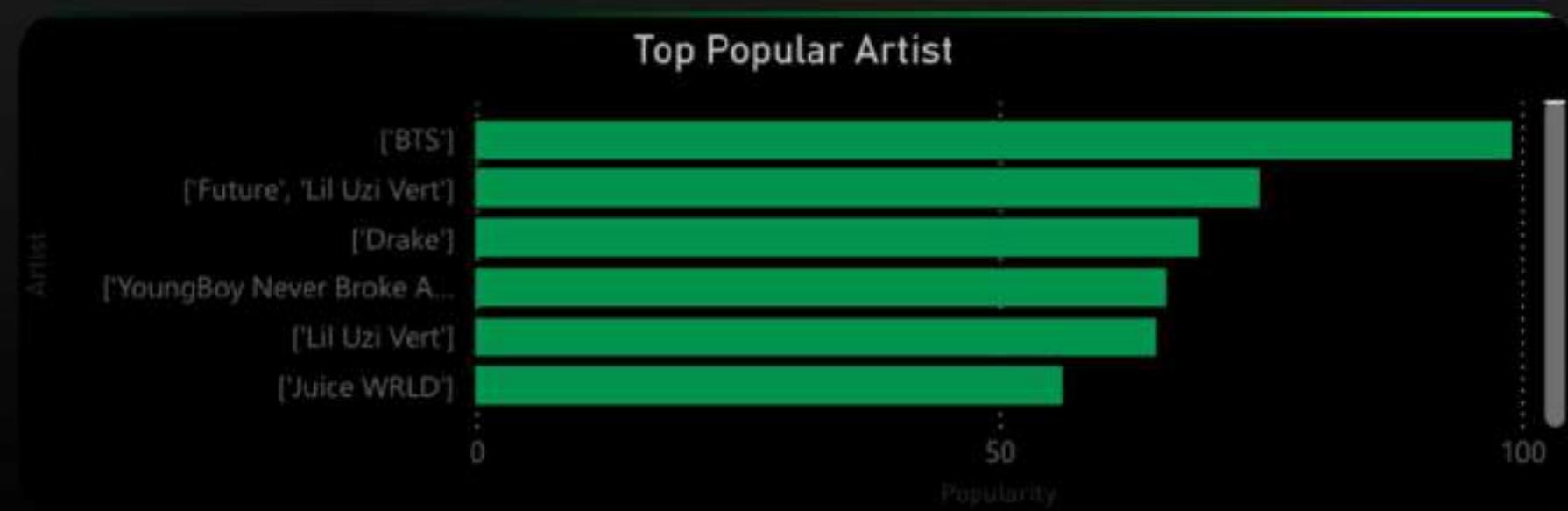


Insight Generation



Recommendations

Key Insights Generated



Top Artists Drive Engagement

BTS maintains 70.77 average popularity with consistent performance. Artist brand strength is crucial.

[Welcome](#)[About Us](#)[Our Team](#)[Agenda](#)[Data Preparation](#)[Dimensional Modeling](#)[Dashboard Development](#)[Insight Generation](#)[Recommendations](#)

Key Insights Generated

6K

Number of Artists

12K

Total Tracks

3,46

Avg Track Duration (min)



Standardized Track Duration

Average of 3.46 minutes across 12K tracks. Duration is not a differentiating factor.



Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



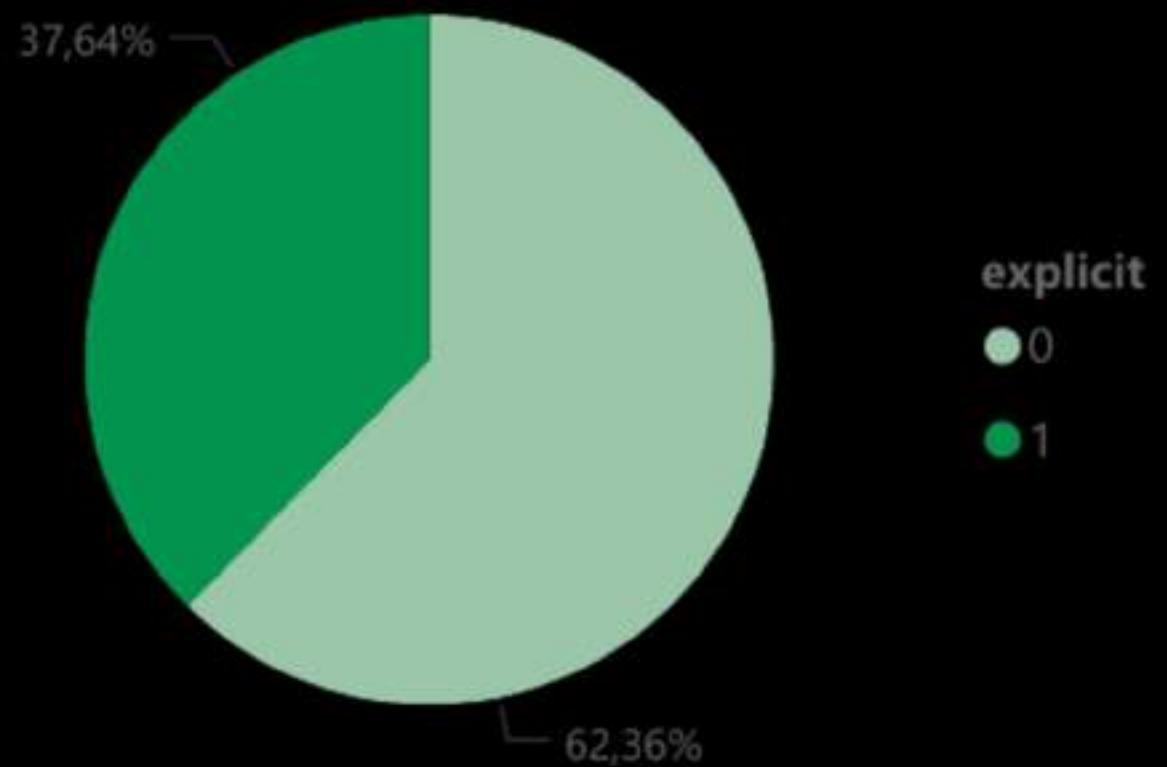
Recommendations

Key Insights Generated

Explicit Content Balance

37.64% explicit vs 62.36% clean tracks. Balance needed between freedom and accessibility.

Explicit Vs Non-Explicit Tracks





Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



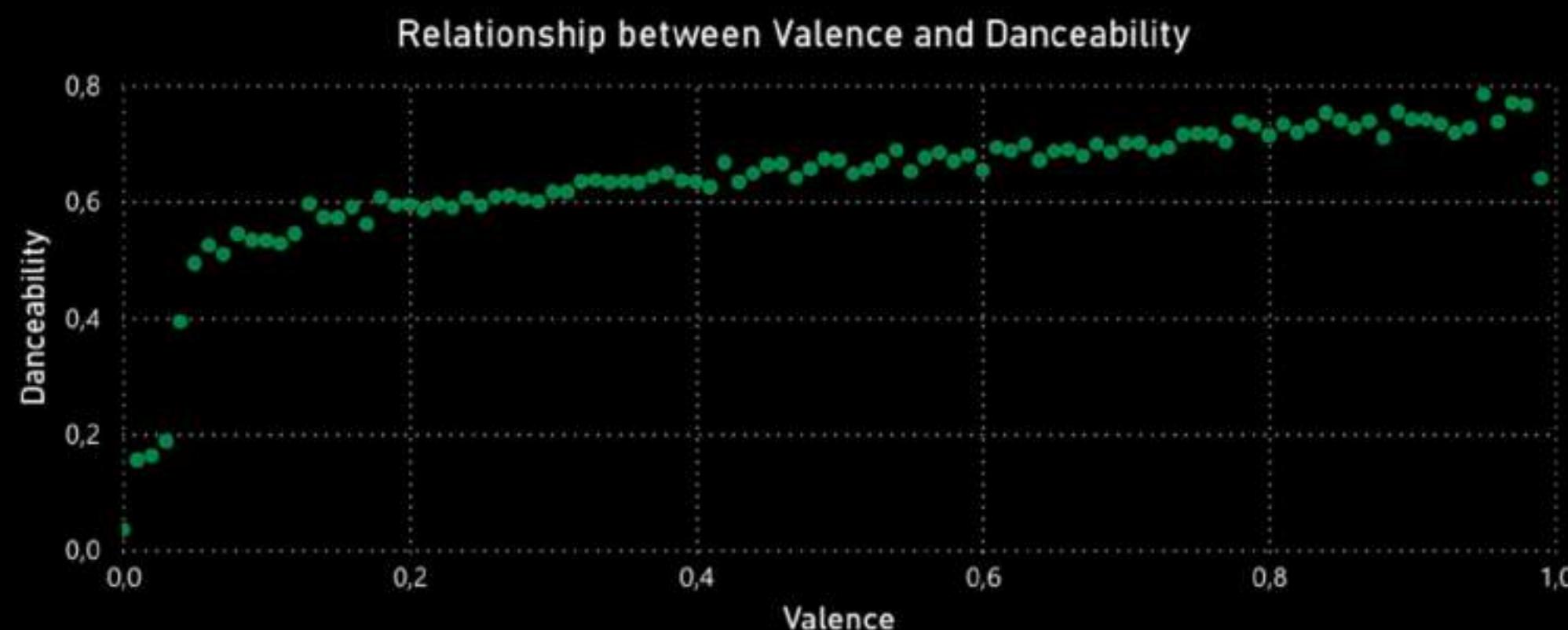
Recommendations

Key Insights Generated



Valence-Danceability Link

Moderate correlation exists. Balanced emotional tones (0.4-0.6 valence) with high danceability perform best.



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[Welcome](#)[About Us](#)[Our Team](#)[Agenda](#)[Data Preparation](#)[Dimensional Modeling](#)[Dashboard Development](#)[Insight Generation](#)[Recommendations](#)

Recommendations



Strategic Recommendations Playlist

Based on data insights

- ▶ Audio-Driven Playlist Design
- ▶ Genre-Aware Recommendation Logic
- ▶ Early Detection of High-Potential Artists
- ▶ Playlist Freshness via Artist Diversity
- ▶ Early-Trend Monitoring Dashboards
- ▶ Actionable Insights for Artists

Expected Impact

Higher engagement

Better personalization

Long-term growth





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Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



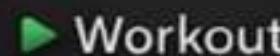
Recommendations

Recommendations

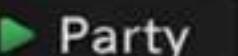


Audio-Driven Playlist Design

Apply clear audio thresholds (danceability, energy) for context-based playlists to ensure consistency and higher engagement.



Workout



Party



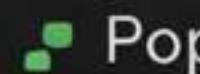
Focus



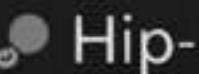
Pop
Hip-
EDM

Genre-Aware Recommendation Logic

Customize recommendation rules by genre instead of using one universal model, making suggestions more relevant and personalized.



Pop



Hip-Hop



EDM



Welcome



About Us



Our Team



Agenda



Data Preparation



Dimensional Modeling



Dashboard Development



Insight Generation



Recommendations

Recommendations



Early Detection of High-Potential Artists

Leverage audio features (energy, danceability, valence) to spot emerging artists with strong popularity potential before they trend.

Emerging Artist

Pop

Hip-Hop

EDM

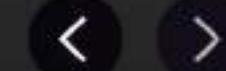


Playlist Freshness Through Artist Diversity

Limit repeated appearances of the same top artists to enhance discovery while maintaining listener engagement.



Top Artists | New Artists

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[Welcome](#)[About Us](#)[Our Team](#)[Agenda](#)[Data Preparation](#)[Dimensional Modeling](#)[Dashboard Development](#)[Insight Generation](#)[Recommendations](#)

Recommendations



Actionable Insights for Artists

Provide simple, comparative performance metrics against genre averages to support artists' creative and marketing decisions.



Artist vs Genre Avg



Early-Trend Monitoring via Dashboards

Use real-time dashboards combining audio features and popularity trends to identify and promote rising tracks faster.



Top Artists | New Artists

Thank You



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