

# Spotify Data Trends

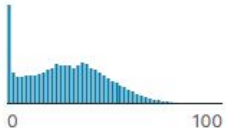



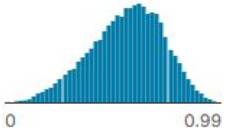
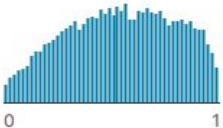
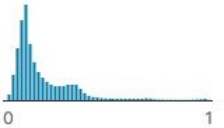
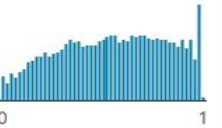

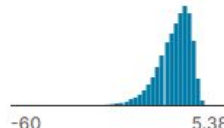


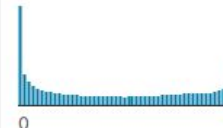


INFO330 Group 9:

Meg Balfrey, June Mi Hong, Terra  
Sumi Shrestha, Maansi Surve

# Our Dataset

- Spotify Data from Kaggle
- Covers data from the year 2000-present
- Basic track data includes song length, tempo, name, and release date
- Observational track data includes danceability, energy, speechiness, liveliness, and valence
- Also includes artist data, including name, genre, followers, and popularity



# Data Examples - Tracks

Column name →	id	name	popularity	duration_ms	explicit	instrumentalness
	586672 unique values	446475 unique values				
Example →	35iwrR4jXetI318WEWsa1Q	Carve	6	126903	0	0.744
artists	id_artists	release_date	danceability	energy	liveness	valence
114030 unique values	115062 unique values	19700 unique values				
['Uli']	['45tIt06XoI0Iio4LBEVpls']	1922-02-22	0.645	0.445	0.151	0.127
key	loudness	mode	speechiness	acousticness	tempo	time_signature
						
0	-13.338	1	0.451	0.674	104.851	3

# Data Examples - Artists

Column name →

Example →

id	# followers	genres	name	# popularity
<b>1104349</b> unique values		<div><div></div>73%</div> <div>['background piano'] 0%</div> <div>Other (298323) 27%</div>	<b>1078660</b> unique values	
0DheY5irMjBUeLybbCUE Z2	0.0	[]	Armid & Amir Zare Pashai feat. Sara Rouzbehani	0

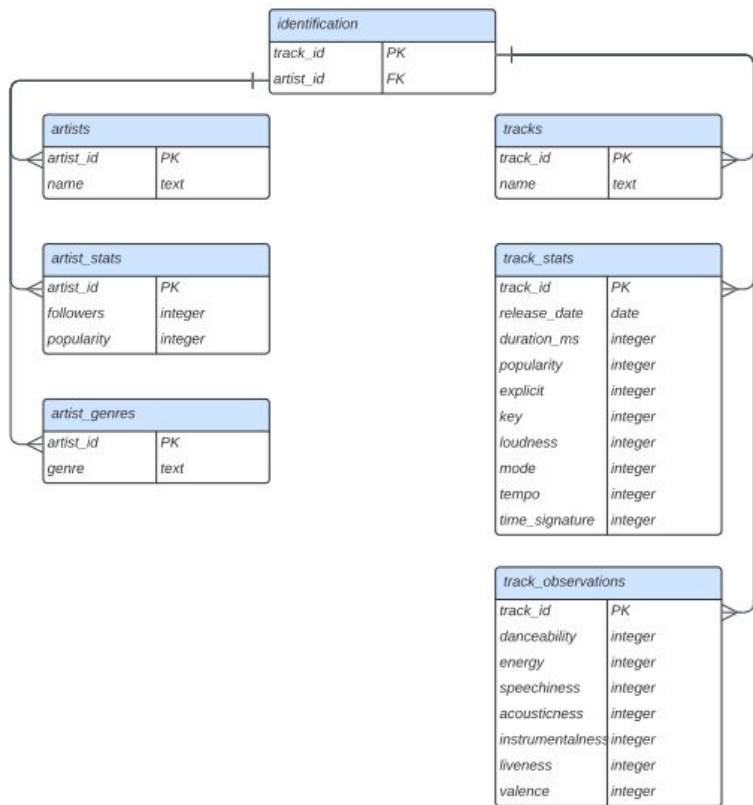
# Our Schema (Overview)

- We started with two tables
  - Artists and tracks
- Identified artist\_id as the PK in the two tables
  - Used to convert to 3NF

artists	
id	text
followers	integer
genres	text
name	text
popularity	integer

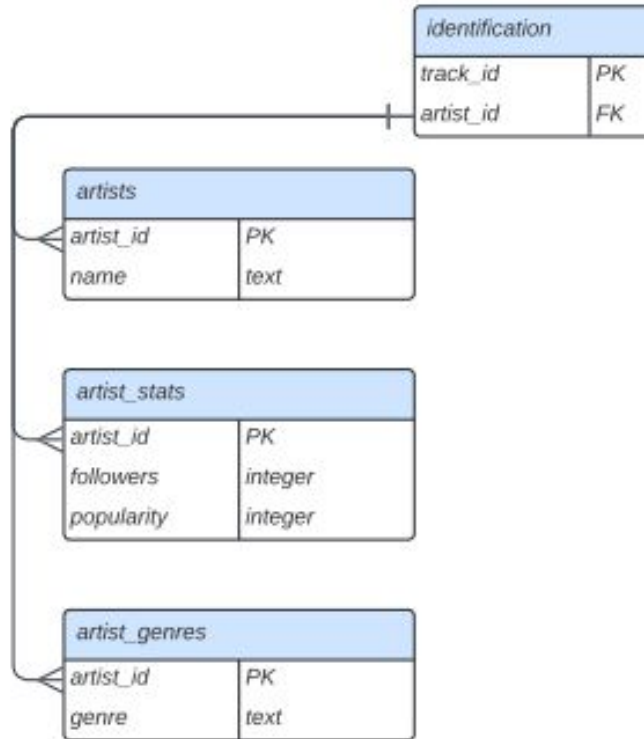
tracks	
id	text
name	text
popularity	integer
duration_ms	integer
explicit	integer
explicit	integer
artists	text
id_artists	text
release_date	date
danceability	integer
energy	integer
key	integer
loudness	integer
mode	integer
speechiness	integer
acousticness	integer
instrumentalness	integer
liveness	integer
valence	integer
tempo	integer
time_signature	integer

# Our Schema (Overview)



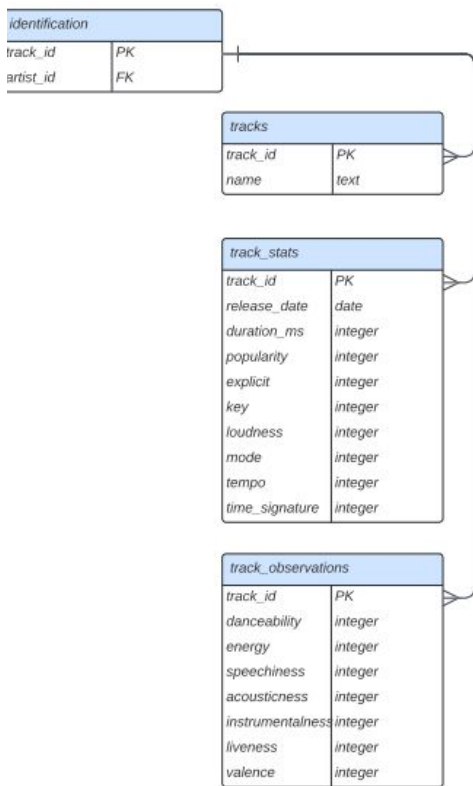
- We divided the data to separate transitive and partial dependencies
- Aimed to reduce as many redundant rows as possible
- Clarified keys

# Our Schema (Breakdown)



- Genre data was fully separated from the rest of artist data to prevent redundancy
- Statistics were separated from artist name to eliminate potential transitive dependencies

# Our Schema (Breakdown)

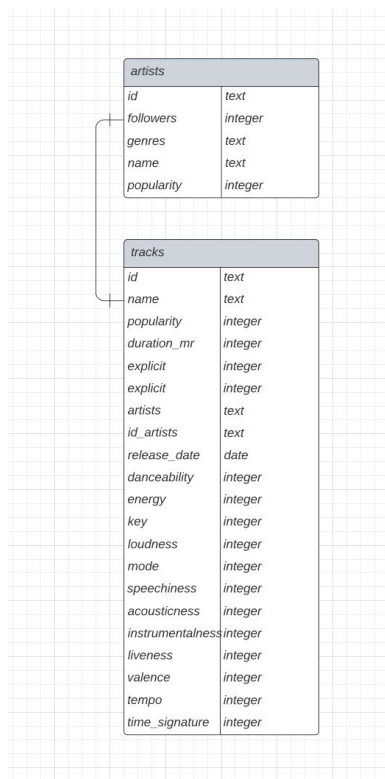


- Statistics were separated from artist name to eliminate potential transitive dependencies
- Observational and factual track statistics may be transitively dependent, so they were also separated



# Challenges

- Genres and artists
  - Separate those two columns to reach 1NF
- Creating a schema helped us visualize



# Query 1a

```
1 SELECT
2   CASE
3     WHEN LENGTH(track_stats.release_date) == 4 THEN track_stats.release_date
4     ELSE SUBSTR(track_stats.release_date, 1, 4)
5   END AS year,
6   SUM(CASE WHEN track_stats.explicit = 0 THEN 1 ELSE 0 END) AS clean_songs,
7   SUM(CASE WHEN track_stats.explicit = 1 THEN 1 ELSE 0 END) AS explicit_songs
8 FROM track_stats
9 GROUP BY year
10 ORDER BY year;
```

	year	clean_songs	explicit_songs
1	2000	1367	98
2	2001	1272	60
3	2002	1480	95
4	2003	1327	52
5	2004	1469	73
6	2005	1266	84
7	2006	1190	55
8	2007	1061	36
9	2008	970	43
10	2009	800	52
11	2010	787	49
12	2011	684	46
13	2012	616	40
14	2013	614	62
15	2014	483	29
16	2015	313	41
17	2016	147	15

- Comparing the number of clean and explicit songs released each year (since 2000)
- Sorted by year
- Can be used to find trends in song content and new norms

# Query 1b

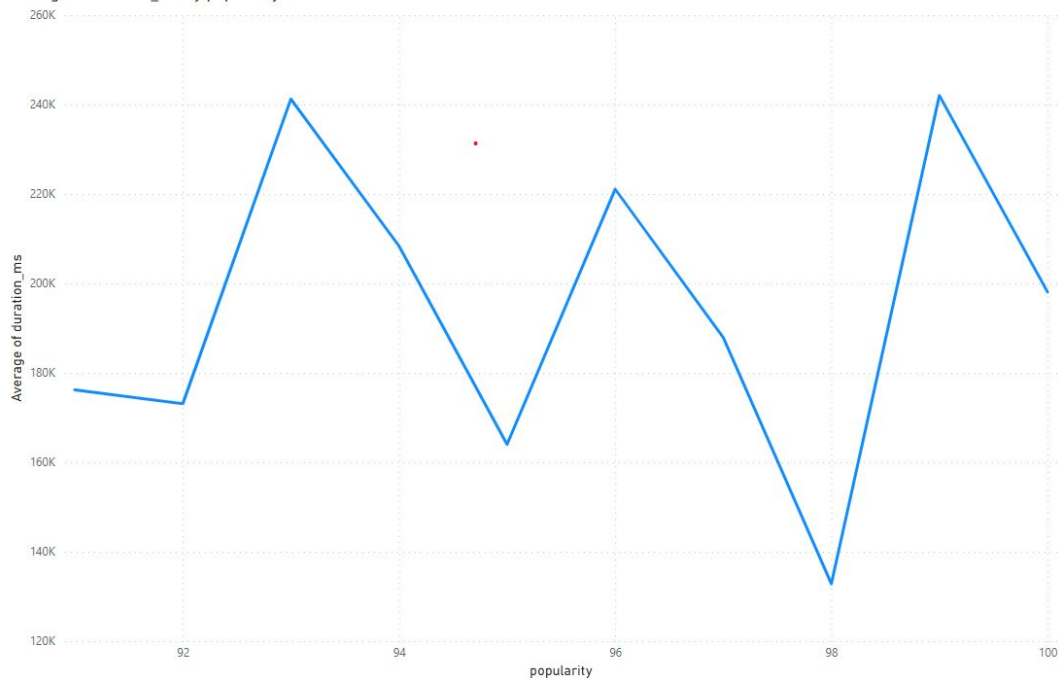
```
1 SELECT tracks.name, track_stats.duration_ms
2 FROM track_stats
3 JOIN tracks ON track_stats.track_id = tracks.track_id
4 WHERE track_stats.popularity > 90
5 ORDER BY track_stats.duration_ms ASC;
```

	name	duration_ms
1	Wellerman - Sea Shanty / 220 KID x Billen Ted ...	116750
2	Astronaut In The Ocean	132780
3	Your Love (9PM)	150053
4	Up	156945
5	What You Know Bout Love	160000
6	telepatía	160191
7	WITHOUT YOU	161385
8	Goosebumps - Remix	162803
9	The Business	164000
10	We're Good	165507
11	Head & Heart (feat. MNEK)	166028
12	Paradise (feat. Dermot Kennedy)	167903
13	Friday (feat. Mufasa & Hypeman) - Dopamine Re...	169153
14	you broke me first	169266
15	Hold On	170813

- Finding the ideal song length
- Finds tracks with popularity between 90-100, their name, and duration
- Sorted from shortest length to longest

# Visualization 1

Average of duration\_ms by popularity



- Finding ideal song length amongst most popular tracks
- Average song length amongst songs with a popularity score of over 90/100

## Query 2a

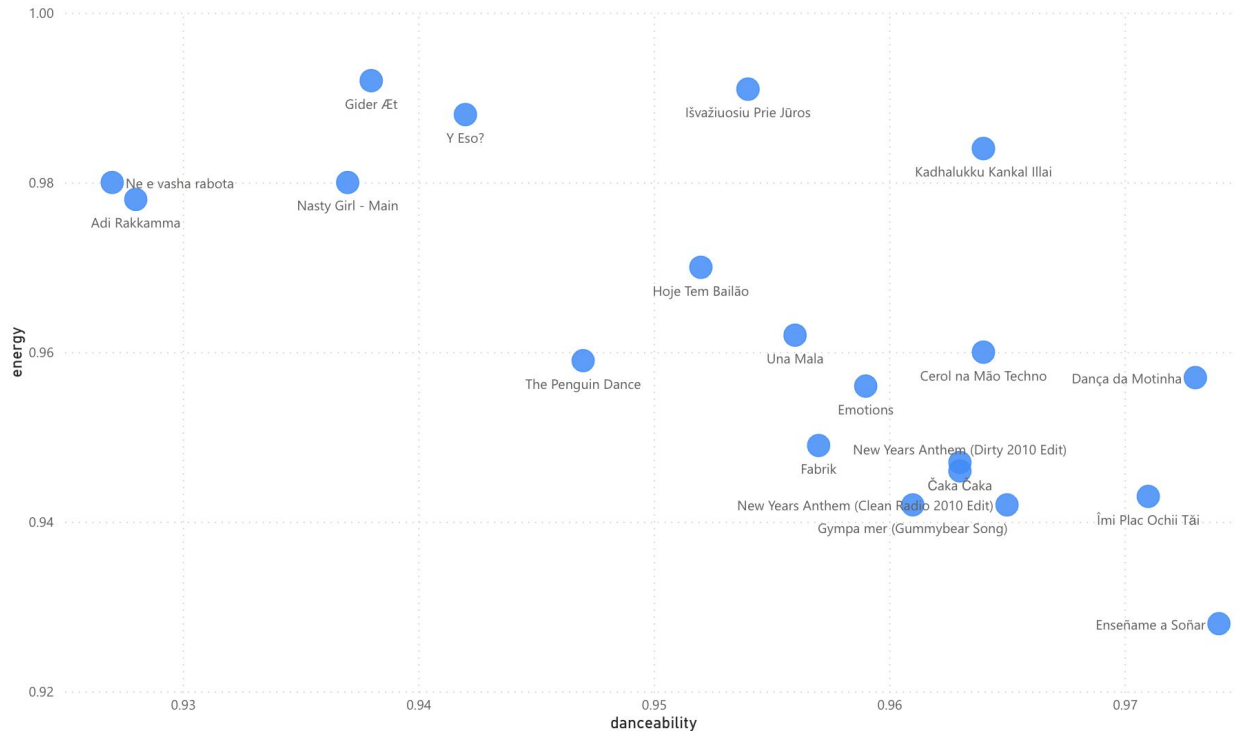
```
-- Query 1
-- Look at tracks' names, id's, and loudness
-- and sorts all the tracks by loudness in descending order
SELECT tracks.track_id, tracks.name, track_stats.loudness
FROM tracks
JOIN track_stats ON tracks.track_id = track_stats.track_id
ORDER BY track_stats.loudness DESC;
```

# Query 2b

```
-- Query 2
-- Look at tracks danceability and energy and find the sum of those two
-- for each track, and sorts the tracks in descending order by sum.
-- It finds the top 20 tracks.
SELECT tracks.name, track_observations.danceability, track_observations.energy,
       track_observations.danceability + track_observations.energy AS dance_energy_sum
FROM track_observations
JOIN tracks ON track_observations.track_id = tracks.track_id
ORDER BY dance_energy_sum DESC
LIMIT 20;
```

# Visualization 2

danceability, energy and sum by name



# Query 3a

```
SELECT t.track_id, t.name, artist_stats.followers
FROM tracks AS t
JOIN identification AS i ON t.track_id = i.track_id
JOIN artist_stats AS artist_stats ON i.artist_id = artist_stats.artist_id
WHERE artist_stats.popularity = (
    SELECT MAX(popularity)
    FROM artist_stats
);
```

```
004sbHg1qMzdrDXdxJpkj1|Heartbreaker - Single Version|44606973.0
01B1GnXpLuc08bjGuxVYZFb|Never Let You Go|44606973.0
01W9YsgZLN5aVuKWD7F4ZN|Sorry|44606973.0
01ger8hoR0K6zASTaad5Nt|Anyone|44606973.0
09CtPGIpyB4Br08qb1RGsF|Sorry|44606973.0
0BTHMXf8gwQMjph0i4211C|Sorry|44606973.0
0DonDJ3Gbf4DGB8qi0e0hc|What Do You Mean?|44606973.0
0Dya83aoBulWa3bcE50T6k|Anyone|44606973.0
0G0Wa1JArdYVidPktgf0MXM|Hold On|44606973.0
0HEoB7h42csWgDQ3H4bd7y|Anyone|44606973.0
0JBq4P9Jpp37F1PteCgGL1|Sorry|44606973.0
0JLiCAJsAULoQtijy6XMqn|What Do You Mean? - Acoustic|44606973.0
0JWNKTKyR9mxS2q3JVeV6z|Love Yourself|44606973.0
0Jnf4qSuc8ze8XD5LZktjY|Anyone|44606973.0
0K2RV0jwqQzeEfsHYZa0WMM|Hold On|44606973.0
0KDJBhhe20YnnoJtbtXy1f|U Smile|44606973.0
0NpZenGEA81m99MX1Q677C|What Do You Mean?|44606973.0
0OR73697LSVLuAauxpCcA0f|Come Around Me|44606973.0
0SNIATrCPVVL0GEPcuHSic|I'll Show You|44606973.0
0Ub799mIwtCMYYo9Mr2Q4A|As Long As You Love Me - Acoustic Version|44606973.0
0ULJuQtv43fosEJLghZJFH|Hold On|44606973.0
0Xbnr60rc8ZDKtAaWw0i6u|Be Alright|44606973.0
0aPZbnkWoWJaJ5CNVLCj8S|That Should Be Me|44606973.0
0cPUK4hukZ3V8be48WmLsY|Hold On|44606973.0
0e7eh5Lhtrypp1b1cYgbCp|Love Yourself|44606973.0
0ff0WJvYrYmtKcmJbPWuHc|Bigger|44606973.0
0gnEBCosLGgWAakmfTq3fS|Anyone|44606973.0
0hIZaDvx7EHULKphw95pgA|Hold On|44606973.0
```

- Finding the correlated track id, name, and followers from artists with the maximum popularity number
- Can be used in the music industry to determine the type of music trends based on the type of track and amount of followers an artist has



# Query 3b

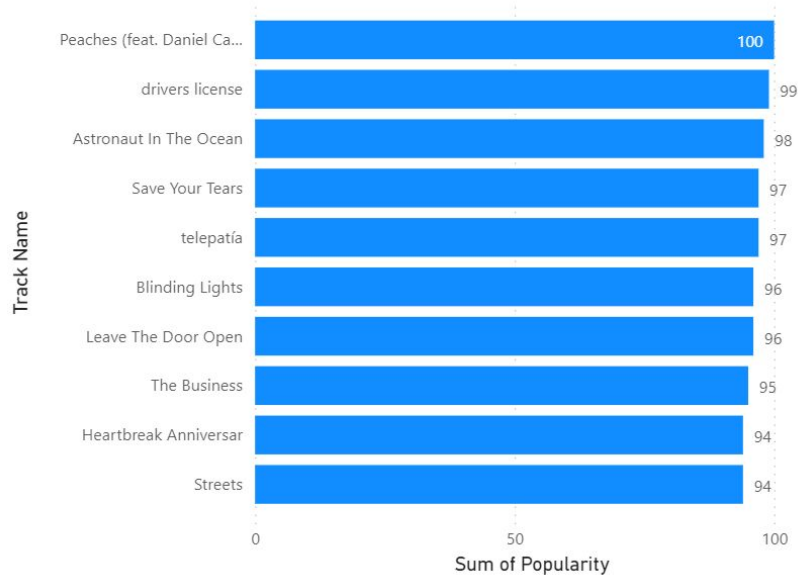
```
SELECT t.track_id, t.name, ts.popularity
FROM tracks AS t
JOIN track_stats AS ts ON t.track_id = ts.track_id
ORDER BY ts.popularity DESC
LIMIT 10;
```

```
4iJyoBOLtHqaGxP12qzhQI|Peaches (feat. Daniel Caesar & Giveon)|100
7LPN2DXiMsVn7XUKt0W1CS|drivers license|99
30fmpyhv5UAQ70mENzB277|Astronaut In The Ocean|98
5Q079kh1waicV47BqGRL3g|Save Your Tears|97
6tDDoYIxWvMLTdKpjFkc1B|telepatía|97
0VjIjW4GLUZAMYd2vXMi3b|Blinding Lights|96
7MAibcTli4IisCtbHkRGMh|Leave The Door Open|96
6f3SlT0GbA2bPZLz0aIFXN|The Business|95
60yNSPSSKe603sfwRnIBRf|Streets|94
3FAJ600NOHQV8Mc5Ri6ENp|Heartbreak Anniversary|94
```

- Finding the top 10 tracks with the highest popularity numbers
- Useful for analyzing the specific type of tracks that have the most popularity

# Visualization 3

Sum of Popularity by Track Name



- Represents the top 10 tracks, along with their sum of popularity
- Can be useful for future music producers or artists






# Query 4a

- Look at the most popular songs from 2020 and their release date
- Helpful for music producers looking at recent trends




QueryHistory

```
1 -- Query 1
2 -- Look at tracks with popularity over 80 that have been released
3 -- Since the beginning of 2020
4
5 SELECT tracks.name, track_stats.release_date
6 FROM track_stats
7 JOIN tracks ON track_stats.track_id = tracks.track_id
8 WHERE track_stats.popularity > 85
9 AND track_stats.release_date >= '2020-01-01' ;
```

Grid viewForm view



1

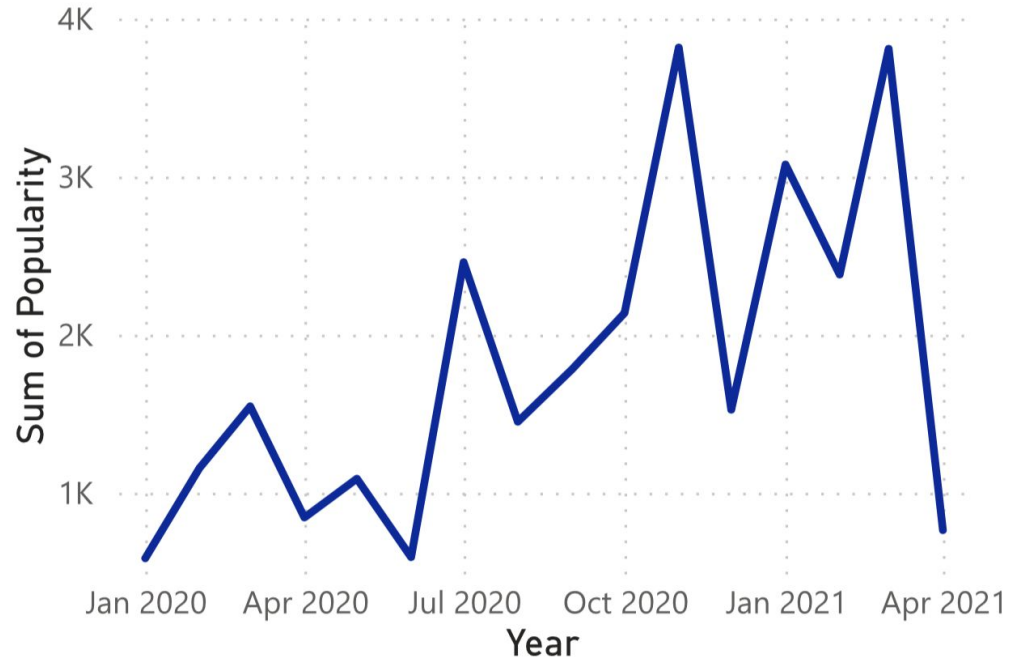


Total rows loaded: 114

	name	release_date
1	Save Your Tears	2020-03-20
2	telepatía	2020-12-04
3	Blinding Lights	2020-03-20
4	The Business	2020-09-16
5	Heartbreak Anniversary	2020-03-27
6	WITHOUT YOU	2020-11-06
7	Bandido	2020-12-10
8	LA NOCHE DE ANOCHE	2020-11-27

# Visualization 4A

Sum of Popularity by Year and Month



# Query 4b

- Look at popularity and followers on Spotify to look for a correlation
- Helps music producers look for artists who are trending but less well known

```
1 -- Query 2
2 -- Compare followers to popularity for artists who have a
3 -- popularity rating above 80
4 SELECT artist_stats.popularity, artist_stats.followers
5 FROM artist_stats
6 WHERE artist_stats.popularity >80;
7 |
```

Grid view Form view

Total rows loaded: 309

	popularity	followers
1	89	4562300
2	81	590066
3	83	4287158
4	90	1624015
5	86	7544862
6	85	4667979
7	82	4796022
8	86	3284229

# Visualization 4B

Sum of Followers by Popularity

