# **Spotify: What Makes a Top Song?**

DigHum 100 Dr. Anderson | Gloria Yang | June 10, 2021

**Description:** I've noticed that features of the most popular songs have changed a lot over the decades. While there are influences and consistencies in popular song attributes from the 20th century, the 21st century brought up new genres and styles of music. Modernization has led to different approaches and greater diversity of sounds such as louder sounds, faster tempos, and speechiness. In this project, I will be analyzing the popular songs on Spotify over time and looking for common attributes and trends.

#### Dataset:

"Spotify: All Time Top 2000 Mega Dataset" is a dataset taken from Kaggle which contains audio statistics and attributes for Spotify's top songs. The dataset is contained within a csv file and includes data on songs released from 1959 to 2019. The data set breaks down each song by the following attributes: genre, year, BPM, danceability, energy, loudness, liveness, valence, length, acousticness, speechiness, and popularity.

I will be able to cluster and manipulate the dataset's columns to make our visualizations more clear during exploration. I will be using the dataset in order to explore the attributes of Spotify's most popular songs and the changes in these attributes over time.

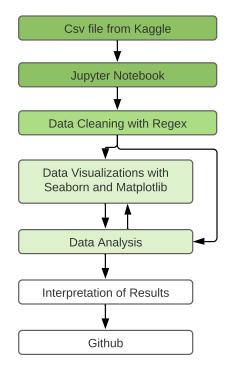
Index	Title	Artist	Top Genre	Year	Minute (BPM)	Energy	Danceability	(dB)	Liveness	Valence	(Duration)	Acousticness	Speechiness	Popularity
1	Sunrise	Norah Jones	adult standards	2004	157	30	53	-14	11	68	201	94	3	71
2	Black Night	Deep Purple	album rock	2000	135	79	50	-11	17	81	207	17	7	39
3	Clint Eastwood	Gorillaz	alternative hip hop	2001	168	69	66	-9	7	52	341	2	17	69
4	The Pretender	Foo Fighters	alternative metal	2007	173	96	43	-4	3	37	269	0	4	76
5	Waitin' On A Sunny Day	Bruce Springsteen	classic rock	2002	106	82	58	-5	10	87	256	1	3	59
1990	Heartbreak Hotel	Elvis Presley	adult standards	1958	94	21	70	-12	11	72	128	84	7	63
1991	Hound Dog	Elvis Presley	adult standards	1958	175	76	36	-8	76	95	136	73	6	69
1992	Johnny B. Goode	Chuck Berry	blues rock	1959	168	80	53	-9	31	97	162	74	7	74
1993	Take Five	The Dave Brubeck Quartet	bebop	1959	174	26	45	-13	7	60	324	54	4	65
1994	Blueberry Hill	Fats Domino	adult standards	1959	133	50	49	-10	16	83	148	74	3	56

### **Questions for EDA:**

- Can we predict a song's popularity based on attributes such as danceability, valence, loudness, and length?
- Are there any trends in the popularity of different music genres?
- How do song attributes differ according to different genres?
- How have the attributes of songs on Spotify's Top 2000 changed over time?

#### Tools:

- Matplotlib and Seaborn: tool to create data visualizations such as heatmaps, line graphs, and scatter plots
- Pandas: to manipulate, clean, and store data within dataframes
- Jupyter Notebooks: tool for data cleaning, data manipulation, and developing visualizations
- Regex: implemented to clean genre data from subgenres
- Github: tool for creating a portfolio repository that includes the original dataset, Jupyter notebook, and storyboard

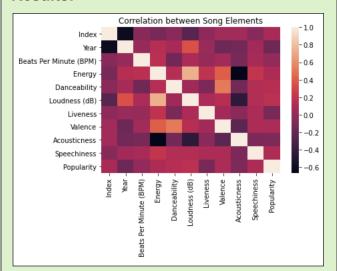


## Methodology:

- 1) I will be converting the "Spotify: All Time Top 2000 Mega Dataset" csv file into a dataset. Tools such as Matplotlib and Regex will also be imported.
- 2) Using regex, songs will be categorized from 10 main genres; songs that have other genres will be categorized as "Other". Data will be further cleaned and formatted.
- **3)** I will be exploring correlation and typical ranges between different song attributes. A heatmap will be derived from the dataset to see if there are any correlations between any two song attributes.
- **4)** Using Pandas and data visualization, I will explore possible shifts in song attributes and genre distribution. If there are shifts or existing attribute characteristics among popular songs, I will try to see such attributes are helpful in predicting a song's popularity.



#### Results:



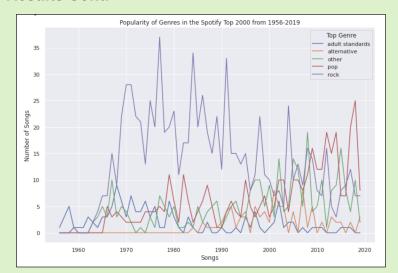
I started playing around with my dataset to get a beginning idea of existing correlations between song attributes. My heatmap indicated a strong correlation between loudness and energy. Further analysis on a scatter plot showed the correlation was positive; the more energetic a song, the more likely to the song is to feel loud.

I was also interested in seeing whether popular songs had ideal lengths and BPMs. With histograms, I was able able to see that the mean song length was approximately 4.15 minutes. The majority of songs also had a BPM with the range of 100-140.

I then used regex to categorize subgenres into the following genres: adult standards, alternative, country, electronic, folk, funk, hip pop, indie, metal, pop, rock, soul, and other. Across all songs in the dataset, the most popular genres were different types of rock and pop.

However, in terms of popularity prediction, attributes in the dataset are taken from top songs over a broad period of time and might not be representative of songs in the present day. Modernization has led to different approaches and greater diversity of sounds such as louder sounds, faster tempos, and speechiness. We must therefore also examine trends in changes to attributes over time.

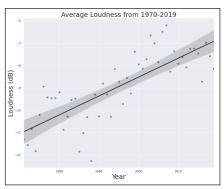
#### **Results Cont:**

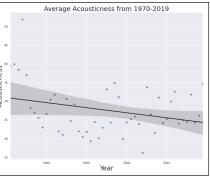


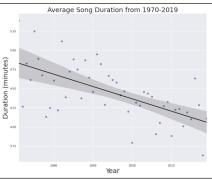
The previous graph shows the number of songs included in Spotify's Top 2000 over time in the following genres: adult standards, pop, rock, alternative, and other. While the number of of rock songs decline into the 21st century, genre diversification is shown through growth for other genres in the number of songs in Spotify's Top 2000 over time. Due to this genre diversification, I expected to see additional shifts in song attributes such as length and BPM.

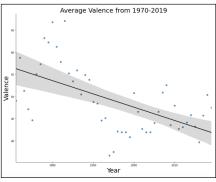
Over time, we can see that the most popular genre shifted from rock to pop. When I examined other attributes for trends over time, I also noticed that attributes such as accousticness and duration tended to decrease while attributes like loudness, danceability, and valence tended to increase. According to The Atlantic, music, particularly pop music, has become intrinsically louder due to current trends in the industry. Valence has also decreased due to widespread content for more negative and sadder content. Likewise, due to new technology, electronic sounds have become more prominent over time. Music has become more electronic and less acoustic, mirroring our modern society's integration with technology. With further research, I also learned that due to the emergence music streaming platforms like Spotify. Artists are paid by how many songs are streamed rather than physical sales, which leads to shorter songs and new song structures such as pop overture. Therefore, besides gender diversification, a song's popularity is influenced by emotional preference and technological and streaming modernization.

Graphs for song attribute trends over time can be viewed in the right column.





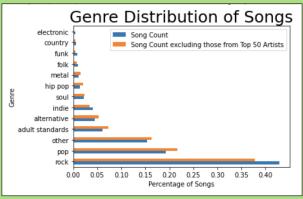


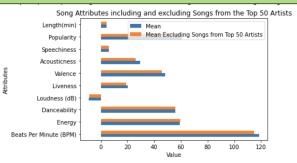


### **Addressing Potential Holes:**

There was a possibility that the my observations were due to the popularity of artists who had many songs included in the dataset rather than because of overall music shifts. If there too many popular artists who had a large number of songs listed, this could skew the dataset's overall genre distribution and other attributes due to less diverse production. In order to see if popular artists had skewed the distribution, I compared attribute and genre distribution of the overall dataset with and without the dataset's top 50 artists. 702 out of 2000 songs were from the top 50 artists and the majority of artists produced rock songs.

Genre distribution stayed relatively the same when excluding songs from the top 50 artists with a slightly lower percentage of rock songs due to exclusion of rock artists like Rolling Stones, Queen, and Bruce Springsteen.





However, if we compare song attributes when excluding and including songs from the top 50 artists, we can see that there is little difference between the song attribute means. Therefore, the inclusion of artists with multiple songs did not skew our observations regarding attribute and genre distribution and trends over time.

#### **Section for Conclusion:**

#### **Limitations:**

Spotify is also a platform that was released 2008, but songs tend to receive more listens at the beginning of its release. Older songs that are on the platform but that were released before the platform existed would receive less listens than if the platform had released at the time of song's release. This would have made older songs potentially less likely to appear in Spotify's Top 2000 songs.

According to Spotify's financial press release in 2019, Spotify had 217 million active users. However, in comparison to other streaming platforms as shown on the right, Spotify had differing demographics in terms of age, gender, and country or origin. Because different demographics might listen to different times and amounts of music, my observations are not representative of all music listeners. There are also other ways for listening to music besides streaming platforms such as radio or buying albums. In order to better guage a song's popularity and how popular songs have changed over time, I would thus also have

Demographic	Spotify	Apple	Pandora
18-24	26	17	11
25-34	29	23	28
35-44	16	22	21
45-54	11	15	17
55+	19	23	22
Male	44	56	41
Female	56	44	59

to include data taken from other streaming platforms and music services around the world.

Lastly, my dataset was only comprised of the top 2000 songs on Spotify. Although I was able to see trends over time, it would have been helpful for me to also compare the attributes of popular songs to songs outside of those in the top 2000. This way, I would be to see whether a trend in what was deemed popular was simply due to modernization or because of preference amongst music listeners.

### **Further Questions:**

- · How do the genre distributions and attributes change with different demographics such as gender and age?
- How representative are popular songs for trends in the music industry?
- Are there any specific artists or genres that are continuing to lead trends in song attributes?
- How have song attributes changed in the past year as a result of COVID 19?
- I am particularly interested in investigating attributes like valence to see if artists's songs are mirroring listener's feelings of loneliness and sadness

### **Hyperlinks:**

- Hyperlink to Jupyter: https://colab.research.google.com/drive/1DsZ4s TyKT TuBdc2z3GnCYQGoZT HXB?usp=sharing
- Hyperlink to CSV: https://drive.google.com/file/d/16BOmSQKc0WRAMaaHU9TBXTGERMbiz2uI/view?usp=sharing
- Hyperlink to Github: https://github.com/gloriiayang/DIGHUM100

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