https://bit.ly/SxB5YA 2018 2019 2017 2020 2021 Good Vibrations? Spotify x Billboard Top 200 Albums 5-Year Analysis Project Bianca Serrano and Katie Ravenwood



CHANGES OVER TIME

Have there been any significant changes in music over the past five years?

Are there other significant features that change over time?

Do any of the features have similarities or contrasts?

GENRE PRESENCE

Have there been any changes in genre presence on the charts? Which genres are changing and how?

MOOD

Has mood of the most popular music changed over the past five years? How has it changed?

DATASET CREATION

PYTHON & SPOTIPY

Data tables of album and track information were created with Python using the SpotiPy package.

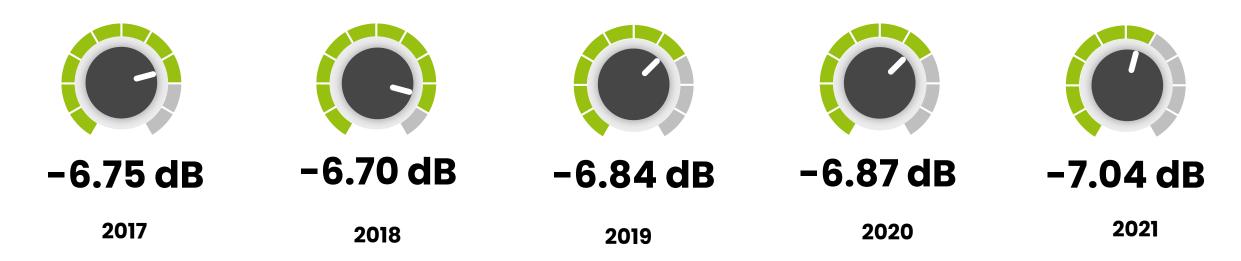
Included variables:

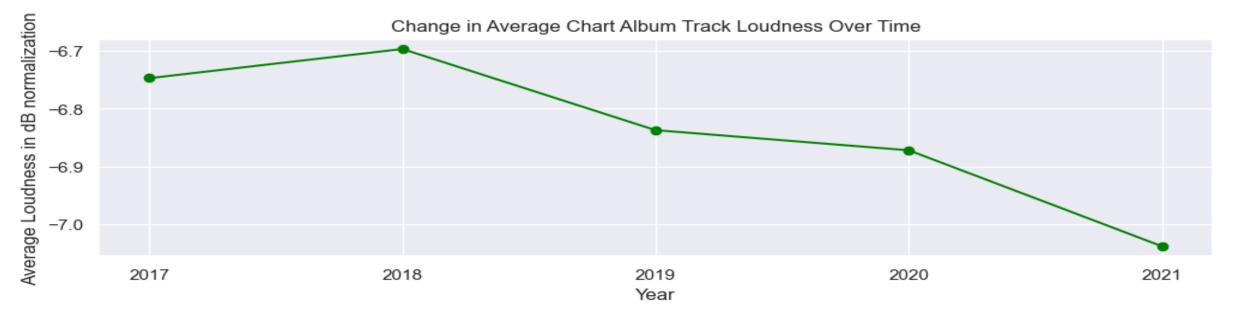
- Billboard album chart name & year
- Album ID, name, release year
- Album artists' names, IDs, popularity, and associated genres
- Track IDs, popularity, and statistics
- Explicit labeling designation
- Audio features

8996 TRACKS FROM 560 ALBUMS

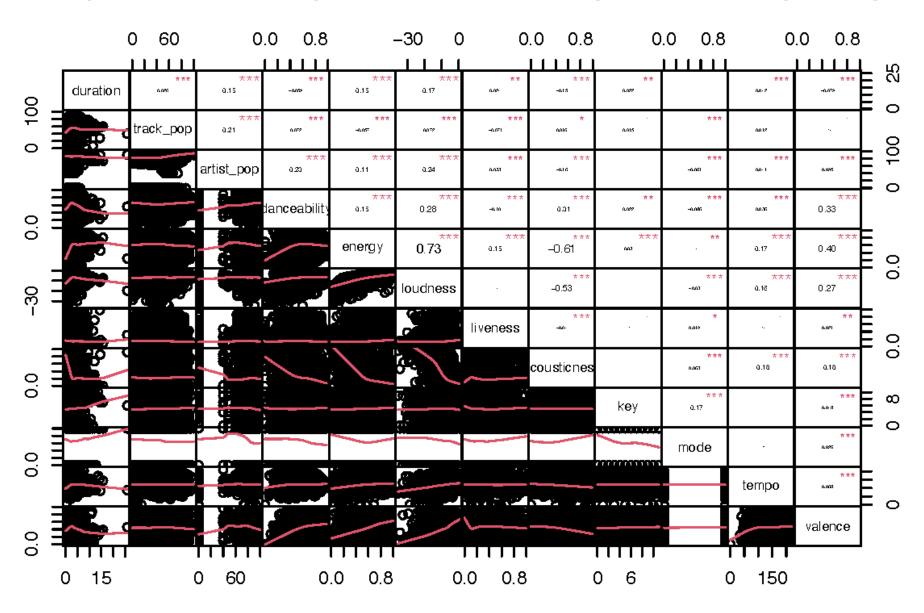
```
JUDYTER AllAlbumTrackTableCreation Last Checkpoint: 01/21/2022 (autosaved)
                                                                                                                             Logout
           Album track data return function
 In [32]: def get_all_album_tracks_more_than_100_songs(album_track_id_list):
               track_id = []
               track_popularity = []
               track_explicit = []
               track_artists = []
               track artists ids = []
               track_artists_genres = []
               track artists popularity = []
               track artists followers = []
               album_name = []
               album_id = []
               for i in tqdm(range(len(album_track_id_list))):
                   track = sp.track(album_track_id_list[i])
                   this id = track['id']
                   if i == 0:
                       track name = track['name']
                       track_id = this_id
                       track popularity = track['popularity']
                       track explicit = track['explicit']
                       track_artists = get_all_track_artists_names(this_id)
                       track_artists_ids = get_all_album_artists_ids(this_id)
                       track_artists_genres = get_all_album_artists_genres(this_id)
                       track_artists_popularity = get_all_album_artists_popularity(this_id)
                       track_artists_followers = get_all_album_artists_followers(this_id)
                       album_name = track['album']['name']
                       album_id = track['album']['id']
                       features = sp.audio features(this id)
                       features_df = pd.DataFrame(data=features, columns=features[0].keys())
                       features df['track name'] = [track name]
                       features df['track id'] = [track id]
                       features_df['track_popularity'] = [track_popularity]
                       features_df['track_explicit'] = [track_explicit]
```

CHANGE IN LOUDNESS OVER TIME





CORRELATION MATRIX OF FEATURES



KMEANS CLUSTERS

1st Cluster:

Mid-high instrumentalness Low-mid liveness Low-mid speechiness

2nd Cluster:

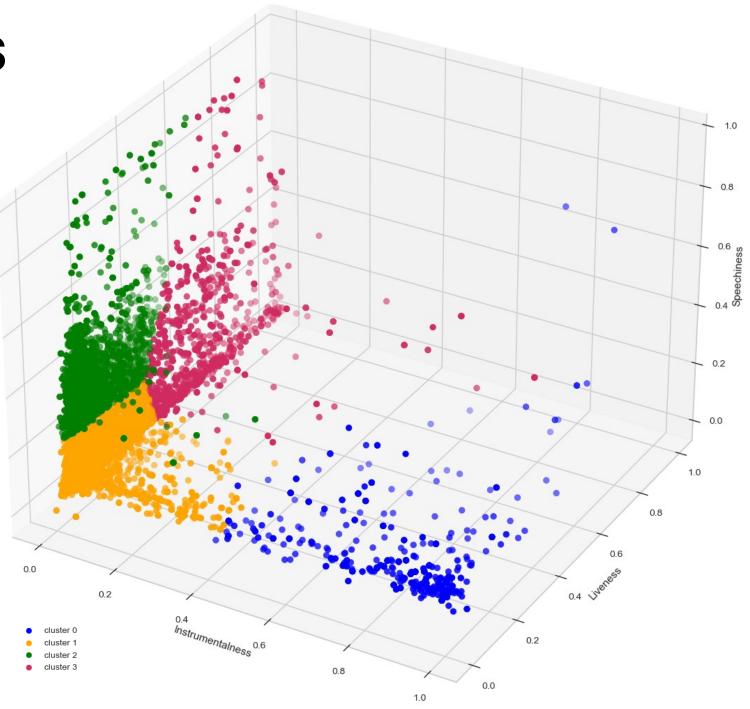
Low-mid instrumentalness Low-mid liveness Low-mid speechiness

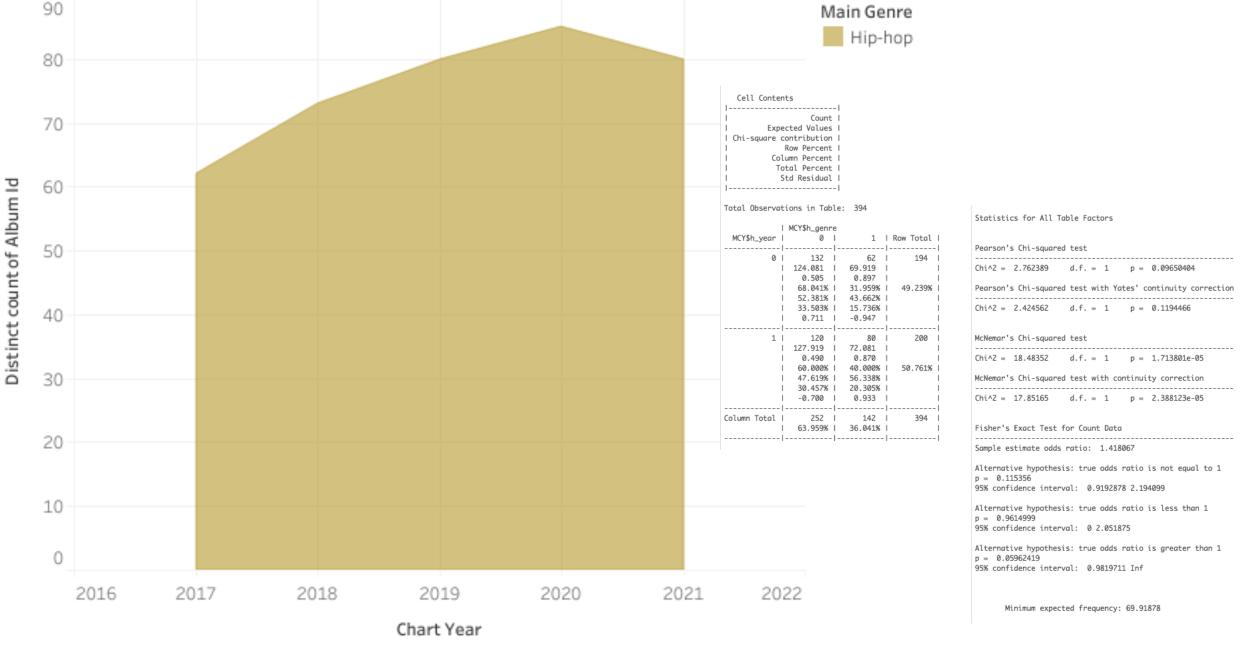
3rd Cluster:

Low instrumentalness Low-mid liveness Mid-high speechiness

4th Cluster:

Low instrumentalness Mid-high liveness Low-mid speechiness





ALBUM COUNT PER CHART YEAR FOR HIP-HOP GENRE

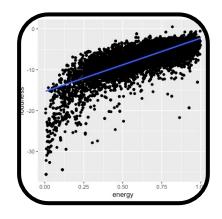
GENRE CLASSIFICATION

K Nearest Neighbors

Random Forest Classifier

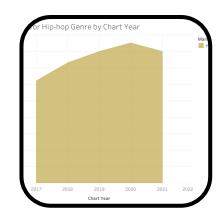
	support	f1-score	recall	precision	
Chris	14	0.59	0.57	0.62	Christian
Cor	305	0.75	0.78	0.72	Country
Easy liste	43	0.73	0.70	0.79	
					Easy_listening
	12	0.10	0.08	0.12	Edm
	2	0.57	1.00	0.40	Folk
***	5	0.29	0.20	0.50	Funk
Hi	1259	0.87	0.85	0.88	Hip-hop
	5	0.89	0.80	1.00	Indie
	3	0.00	0.00	0.00	Jazz
]	47	0.44	0.40	0.47	Latin
1	45	0.88	0.89	0.87	Metal
	53	0.72	0.70	0.74	Misc
	773	0.75	0.75	0.76	Pop
	118	0.66	0.68	0.65	R&B
Re	13	0.90	1.00	0.81	Reggae
	402	0.87	0.90	0.84	Rock
	13	0.74	0.77	0.71	Soul
Sound	189	0.75	0.75	0.76	Soundtrack
Douna					
acci	3301	0.80			accuracy
10 000 to 10 000	3301	0.64	0.66	0 65	macro avg
macro	3301	0.80	0.80	0.80	weighted avg
weighted					

	precision	recall	f1-score	support
Christian	1.00	0.48	0.65	25
Country	0.81	0.80	0.80	450
Easy_listening	0.93	0.63	0.75	62
Edm	0.00	0.00	0.00	13
Folk	1.00	1.00	1.00	8
Funk	0.33	0.12	0.18	8
Hip-hop	0.86	0.94	0.90	1896
Indie	1.00	0.44	0.62	9
Jazz	0.00	0.00	0.00	4
Latin	0.96	0.42	0.59	64
Metal	0.89	0.83	0.86	65
Misc	0.97	0.51	0.67	76
Pop	0.74	0.80	0.77	1142
R&B	0.93	0.65	0.77	176
Reggae	1.00	1.00	1.00	19
Rock	0.85	0.88	0.87	615
Soul	1.00	0.67	0.80	24
Soundtrack	0.94	0.73	0.82	295
accuracy			0.84	4951
macro avg	0,79	0.61	0.67	4951
weighted avg	0.84	0.84	0.83	4951



GENRE PRESENCE

Hip-hop presence on the charts increased while albums in the the Country genre decreased.



Change in Average Chart Album Track Valence Over Time 2018 2019 2020 Year

ENERGY CHANGE

Average track energy decreased during the five-year analysis period.

CORRELATIONS

Energy was significantly correlated with loudness and valence.

Valence generally increased during the five year analysis period, particularly in 2020.

VALENCE CHANGE

