# Billboard Hot 100 Music Recommender

#### **Problem Statement**

As growth in the accessibility of technology with high computation power, along with the availability of free education, has skyrocketed, music creation is easier to get into then ever before. This has led to a boom in the number of artists putting out music on various platforms such as Apple Music, Spotify, and Soundcloud, which only compounds year after year.

As more and more new artists are added every year, older artists tend to be left in the past. Most music discovery platforms focus on current trends so they will rarely recommend anything that wouldn't' be considered somewhat recent. I want to solve this problem, by creating a recommendation model that can be used to suggest songs that made it to the Billboard Hot 100 in the years 1960-1999 given a user input song.

## Types of Recommenders

#### **Content Based Recommender**

- Built around information/features of items or users (i.e. song metadata)
- Each item is recorded in a row with features describing that item in the columns
- Idea is to score similarities/differences between all individual items in a matrix and provide recommendation based on that ranked score

#### Collaborative Based Recommender

- Built around the past interactions that have been recorded between users and items
- Data would be stored in a User-Item Interaction Matrix
- Idea is that past user-item interactions is enough to find patterns for similar users and/or recommended items

## **Project Flow**

- O 1 Web Scrape Wikipedia
- O 02 Spotify API Client
- 03 Gathering Metadata

- O 04 EDA
- 05 Music Recommender: Production
- O 06 Music Recommender: In Action

### Wikipedia Web Scrape: Billboard Hot 100

- I started with a web scrape of the Wikipedia entries for the Billboard Hot 100 number ones for each year 1960-99
- First consideration was that each song was broken up into multiple entries depending on when it was issued and when it left the charts
- Decided to keep 1 occurrence of each entry

#### List of Billboard Hot 100 number ones of 1960

From Wikipedia, the free encyclopedia

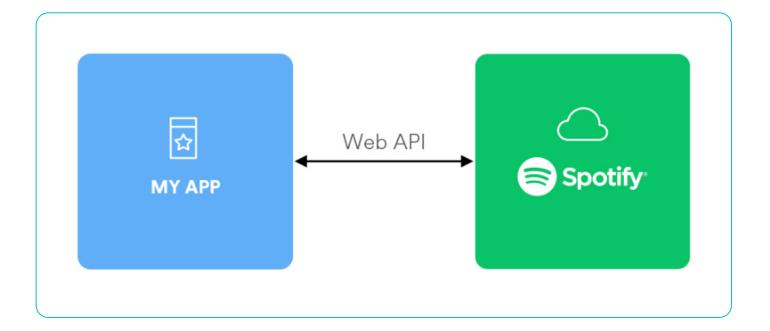
These are the Billboard Hot 100 number one hits of 1960.

That year, 12 acts achieved their first number ones, such as Marty Robbins, Johnny Preston, Mark Dinning, Connie Francis, T Larry Verne, The Drifters, Ray Charles, and Maurice Williams and the Zodiacs. Percy Faith and The Everly Brothers, despite t creation of the Hot 100, also achieved their first number one songs on the chart. Elvis Presley, Connie Francis, and Brenda Le

#### Key

The yellow background indicates the #1 song on Billboard's 1960 Year-End Chart of Pop Singles

No.	Issue date	Song	Artist(s)	Reference
24	January 4	"El Paso"	Marty Robbins	[1]
	January 11			[2]
25	January 18	"Running Bear"	Johnny Preston	[3]
	January 25			[4]
	February 1			[5]
26	February 8	"Teen Angel"	Mark Dinning	[6]
	February 15			[7]
	February 22			[8]
	February 29			[9]
	March 7			[10]
	March 14			[11]
27	M 04	"There from A Commerce Discour	Danni Faith	[12]



## Spotify WebAPI Client: BillboardHot 100

- In order to recommend songs I needed a way to measure similarity between songs
- O Spotify WebAPI provides song metadata that can be pulled, however navigating to the correct song can be a challenge
- O I wanted to build a client from the ground up to configure the search to make it easier to loop through the web scraped data

## Quick Look at Spotify Client

```
class SpotifyAPI(object):
   access token = None
    access token expires = datetime.datetime.now()
    access token did expire = True
   client id = None
   client secret = None
    token url = 'https://accounts.spotify.com/api/token'
   def __init__(self, client_id=os.environ['HOT100_REC_CID'],
                                                                               client_secret=os.environ['HOT100_REC_CS'], *args, **kwargs):
        super().__init__(*args, **kwargs)
       self.client id = client id
       self.client secret = client secret
   def get_client_credentials(self):
       Returns a base64 encoded string
       client_id = self.client_id
       client secret = self.client secret
       if client secret == None or client id == None:
           raise Exception("You must set client id and client secret")
       client creds = f'{client id}:{client_secret}
       client creds b64 = base64.b64encode(client creds.encode())
       return client creds b64.decode()
    def get token headers(self):
       client creds b64 = self.get client credentials()
    'Authorization': f'Basic {client creds b64}'
    def get token data(self):
       return {
    'grant type': 'client credentials'
```

```
def get_album(self, _id):
   return self.get resource( id, resource type='albums')
def get artist(self, id):
   return self.get resource(_id, resource_type='artists')
def get track(self, id):
    return self.get_resource(_id, resource_type='tracks')
def get features(self, id):
    return self.get resource( id, resource type='audio-features')
def get_analysis(self, _id):
   return self.get_resource(_id, resource_type='audio-analysis')
def base_search(self, query_params):
    access token = self.get access token()
   headers = self.get resource header()
   endpoint = "https://api.spotify.com/v1/search"
   lookup_url = f"{endpoint}?{query params}"
   r = requests.get(lookup url, headers=headers)
   if r.status code not in range(200, 299):
       return {}
   return r.json()
def search(self, query=None, operator=None, operator query=None, search type='artist'):
    if query == None:
       raise Exception("A query is required")
    if isinstance(query, dict):
        query = " ".join([f"{k}:{v}" for k,v in query.items()])
    if operator != None and operator query != None:
       if operator.lower() == "or" or operator.lower() == "not":
           operator = operator.upper()
            if isinstance(operator_query, str):
               query = f"{query} {operator} {operator_query}"
    query params = urlencode({"q": query, "type": search type.lower()})
   return self.base search(query params)
```

## Data Collection and Search Validation

- In order to get the right song from the Spotify API I built a validation algorithm to make sure the search returned the song with the same artist released in the same year or earlier then the year it appeared on the Hot 100
- One choice I made was to not limit the song to only 1 version and include any version of the song the search turned up with. This means live versions, re-recordings, reworks, and re-masters will be included.

```
for song in df['song'][c1:]:
    track search = spotify.search({"track": str(song)}, search type="track")
    if len(track_search['tracks']['items']) > 0:
        for i in range(len(track_search['tracks']['items'])):
            song_info = []
            if song in track_search['tracks']['items'][i]['name'].replace('"', ''):
                if (df['artist'][c1] in track_search['tracks']['items'][i]['artists'][0]['name']) and\
                (df['artist'][c1] in track_search['tracks']['items'][i]['album']['artists'][0]['name']):
                    if df['year'][c1] <= int(track_search['tracks']['items'][i]['album']['release_date'][:4]):</pre>
                        song_info.append(track_search['tracks']['items'][i]['name'])
                        song_info.append(track_search['tracks']['items'][i]['album']['name'])
                        song_info.append(track_search['tracks']['items'][i]['album']['artists'][0]['name'])
                        song_info.append(track_search['tracks']['items'][i]['popularity'])
                        song_info.append(track_search['tracks']['items'][i]['id'])
                        song info.append(track search['tracks']['items'][i]['explicit'])
                        track list.append(song info)
```

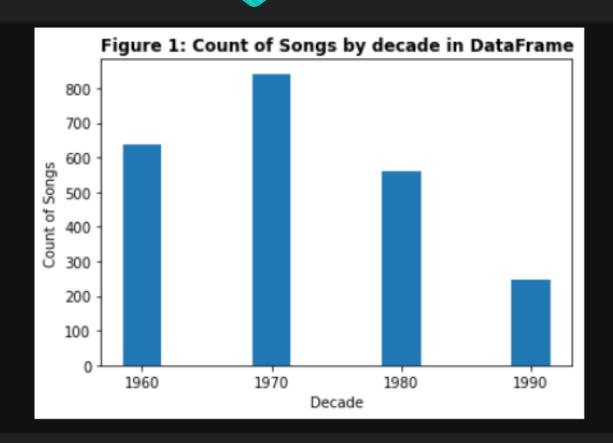
## Song Features and Analysis

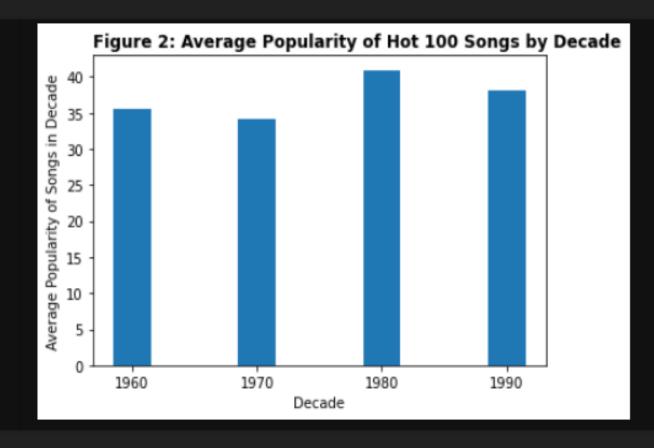
The recommender would need each song to have metadata that it could compare against itself to determine similarity. Here are the main features used by the recommender model ------

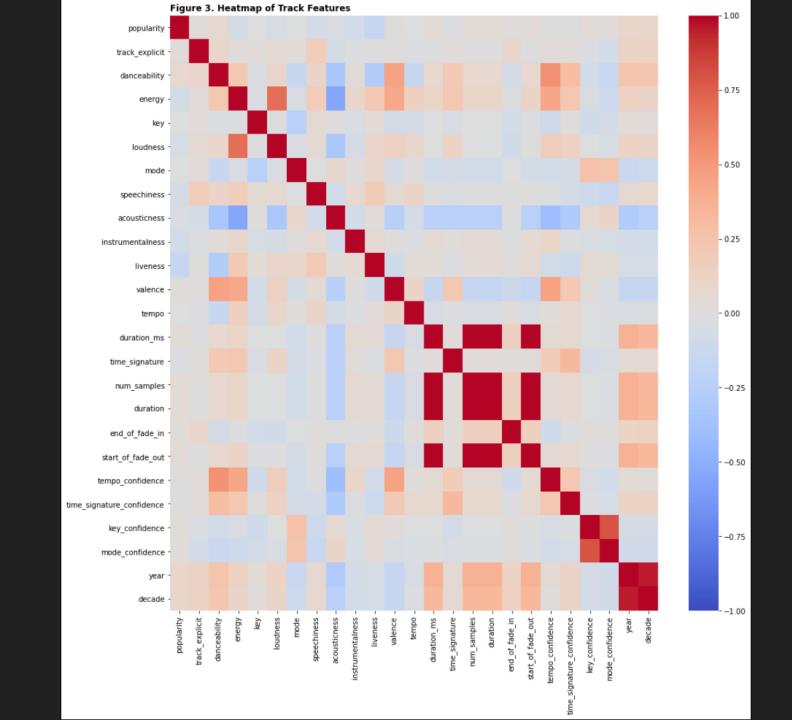
#### Song Features:

- Danceability
- Energy
- Key
- Loudness
- Mode
- Speechiness
- Acousticness
- Instrumentalness
- Liveness
- Valence
- Tempo

## **Understanding Decade Distribution**







### Recommendation Model

- Steps to Recommendation Model:
  - Find User Input song and gather metadata from Spotify WebAPI
  - Match new user song dataframe with format of Main Reference dataframe and concatenate
  - Find and sort on the cosine similarity score comparing main dataframe against itself
  - Print out top 10 recommendations with no artist being repeated

```
'song, 'album, 'artist', popularity', 'erack_did,'
'track_explicit; 'danceability', 'energy', 'key', 'loudness', 'mode',
'speckhiness', 'acousticness', 'instrumentalness', 'liveness',
'valence', 'tempo', 'duration_ms', 'time_signature', 'num_samples',
'duration', 'end_of_fade_in', 'start_of_fade_out', 'tempo_confidence',
'time_signature_confidence', 'key_confidence', 'mode_confidence',
'pd_read_csv('../data/spotify_final.csv').drop('Ummamed: 0', xxis=1)
 <u>__init__(self</u>, song_name, *args, **kwargs):
super()._init__(*args, **kwargs)
self.song_name = song_name
self.song_id = None
spotify = SpotifyAPI()
song_name = self.song_name
song_search = spotify.search({"track": song_name}, search_type="track")
if len(song_search['tracks']['items']) > 0:
      ten(song_search['tracks']['items']);
if self.song_name -- song_search['tracks']['items'][i]['name'];
                  song_id = song_search['tracks']['items'][i]['id']
      if song_id != 0:
            return song id
spotify = SpotifyAPI()
track = spotify.get_track(song_id)
self.track_data.append(track['name'])
self.track_data.append(track['album']['name'])
 self.track_data.append(track['artists'][0]['name'])
 self.track_data.append(track['p
 self.track_data.append(track['id'])
 self.track_data.append(int(track['explicit']))
 spotify - SpotifyAPI()
 track_features - spotify.get_features(song_id)
 self.track_data.append(track_features['danc
  self.track_data.append(track_features['
  self.track_data.append(track_features['key'])
 self.track_data.append(track_features['speechiness'])
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 self.track data.append(track features[
 self.track data.append(track features[
 self.track_data.append(track_features['tempo'])
 self.track_data.append(track_features['
```

```
song id = self.song id
    spotify = SpotifyAPI()
   track_analysis = spotify.get_analysis(song_id)
   self.track_data.append(track_analysis['track']['num_samples'])
   self.track_data.append(track_analysis['track']['duration'])
   self.track_data.append(track_analysis['track']['end_of_fade_
   self.track_data.append(track_analysis['track']['start_of_fad
   self.track_data.append(track_analysis['track'][
   self.track_data.append(track_analysis['track'][
   self.track_data.append(track_analysis['track']['key_confidence
   self.track_data.append(track_analysis['track']['mode_confiden
def search(self):
   self.song_id = self.find_song()
       raise Exception("Song not found.")
       self.get_track_data()
       self.get_track_features()
       self.get_track_analysis()
   return pd.DataFrame([self.track_data], columns=self.col_name)
def print_recommendations(self, indi, rec_df):
    print(f"For the song {self.song_name} by {rec_df['artist'].iloc[
   artist_already_featured = []
   for i in indi[1:]:
       if rec_df['artist'].iloc[i] in artist_already_featured:
           print(f"{c+1}. {rec_df['song'].iloc[i].title()} by {rec_
           artist already featured.append(rec df['artist'].iloc[i])
           if c >= 10:
def recommend(self):
   search df = self.search()
   rec_df = pd.concat([df, search_df], ignore_index=True)
   features = [x for x in df.columns if x not in ['song', 'album'
   cosine_similarities = cosine_similarity(rec_df[features])
   indicies = pd.Series(rec_df.index, index=rec_df['song'])
    idx = indicies[self.song_name]
   sim_scores = list(enumerate(cosine_similarities[idx]))
   sim_scores = sorted(sim_scores, key=lambda x: x[1], reverse=True
   indi = [i[0] for i in sim_scores]
   return self.print_recommendations(indi, rec_df)
```

#### ----- Begin Here ---------- Example Recommendation Searches ------Example 1: New Light by John Mayer [2]: ex\_1 = RecommendSong("New Light") ex 1.recommend() For the song New Light by John Mayer, we recommend you check out: 1. I Want You Back by The Jackson 5 2. Brand New Key by Melanie 3. Against All Odds Take A Look At Me Now 2016 Remaster by Phil Collins 4. Sweet Dreams Are Made Of This Remastered by Eurythmics 5. I Got You Babe by Sonny & Cher 6. Take On Me 2017 Acoustic by Aha 7. Believe by Cher 8. Let Your Love Flow by The Bellamy Brothers 9. Heaven by Bryan Adams 10. Together Forever by Rick Astley

#### **Example 2: Lose Yourself by Eminem**

```
[4]: ex_2 = RecommendSong("Lose Yourself")
ex 2.recommend()
```

For the song Lose Yourself by Eminem, we recommend you check out:

- 1. Every Rose Has Its Thorn by Poison
- 2. Brown Sugar by The Rolling Stones
- 3. Rhinestone Cowboy by Glen Campbell
- 4. Beat It by Michael Jackson
- 5. Boogie Oogie Oogie Remastered by A Taste Of Honey
- 6. Another One Bites The Dust 2011 Remaster by Queen
- 7. Flashdance... What A Feeling by Irene Cara
- 8. Oh Sheila by Ready For The World
- 9. Centerfold by The J. Geils Band
- 10. Money For Nothing by Dire Straits

#### Example 3: good 4 u by Olivia Rodrigo

```
[7]: ex_3 = RecommendSong("good 4 u")
ex_3.recommend()
```

For the song good 4 u by Olivia Rodrigo, we recommend you check out:

- 1. I Want To Hold Your Hand Remastered 2015 by The Beatles
- 2. Its My Party by Lesley Gore
- 3. The Locomotion by Little Eva
- 4. Hello, I Love You by The Doors
- 5. I Get Around Mono by The Beach Boys
- 6. Baby Love by The Supremes
- 7. Everyday People by Various Artists
- 8. Please Mr. Postman by The Marvelettes
- 9. Big Girls Dont Cry by Frankie Valli & The Four Seasons
- 10. Stuck On You by Elvis Presley

## Recommender Model Search Examples