

Imports

```
In [1]: import pandas as pd
import json
from typing import List
from os import listdir
import spotipy
import spotipy.util as util
import plotly.express as px
import plotly.io as pio
from spotipy.oauth2 import SpotifyClientCredentials
```

Getting the Data

To retrieve my Spotify data, I navigated to the privacy settings on Spotify's website at <https://www.spotify.com/> (<https://www.spotify.com/>). I initiated the data request process, and it took approximately 15 days for me to receive a zip file containing my data via email.

```
In [2]: def get_streamings(path: str = 'MyData') -> List[dict]:
        """
        Retrieves Spotify streaming history from the specified directory.

        Parameters:
            path (str): The directory path where Spotify streaming history data files

        Returns:
            List[dict]: A list of dictionaries containing streaming history data.
        """

        # List all files in the specified directory with the correct naming scheme
        files = [f'{path}/{x}' for x in listdir(path) if x.startswith('Streaming_Histo

        all_streamings = []

        for file in files:
            # Open and read the file's content
            with open(file, 'r', encoding='UTF-8') as f:
                # Load JSON content from the file
                new_streamings = json.load(f)
                all_streamings += new_streamings

        return all_streamings

# dir path in my machine
spotify_data_directory = '/Users/adrianrodriguez/Downloads/MyData 3/'

# Call the function to retrieve streaming history data
streaming_data = get_streamings(spotify_data_directory)
```

Converted 'streaming_data' into DataFrame and used the 'ts' (timestamp) column to create 'year' and 'month' columns.

```

In [3]: # Loads streaming data into a Pandas DataFrame
streaming_data = pd.DataFrame(streaming_data)

# Converts the 'ts' (timestamp) column to a datetime format
streaming_data['ts'] = pd.to_datetime(streaming_data['ts'])

# Creates a new column 'year' to extract and store the year from the 'ts' column
streaming_data['year'] = streaming_data['ts'].dt.year

# Creates a new column 'month' to extract and store the month as a string
# using the strftime method to format it to the full month name
streaming_data['month'] = streaming_data['ts'].dt.strftime('%B')

# Creates DataFrame
streaming_data = pd.DataFrame(streaming_data)

# Drop the unwanted columns
streaming_data.drop(columns=['ip_addr_decrypted', 'username'], inplace=True)

# Displays the updated DataFrame with the new 'year' and 'month' columns and remove
streaming_data

```

Out [3]:

album_name	spotify_track_uri	episode_name	...	spotify_episode_uri	reason_start	reason
st Nightmare	spotify:track:7e8utCy2JISB8dRHKi49xM	None	...	None	trackdone	tracki
st Nightmare	spotify:track:7ABWRukVQcXlrDKDx5Gek	None	...	None	trackdone	tracki
st Nightmare	spotify:track:13NCxLOlvQ4Tnexgfp03Gs	None	...	None	trackdone	tracki
st Nightmare	spotify:track:7gPd55hW5pVjTm3H9S1Wbv	None	...	None	trackdone	tracki
st Nightmare	spotify:track:0ZUXj43fteJjwvGoLMntte	None	...	None	trackdone	tracki
...
City Club	spotify:track:4WmkSPSFkXMPWrzPJOiz1H	None	...	None	fwdbtn	fw
City Club	spotify:track:2iCzZEPF8GTfQ1v4WCi7De	None	...	None	fwdbtn	fw
City Club	spotify:track:3UL1VvIrSYjTKbpgGAyyUx	None	...	None	fwdbtn	fw
City Club	spotify:track:63KqjQQ4ofFpQkmPXKTyEX	None	...	None	fwdbtn	fw
City Club	spotify:track:0EHGQN0jJ0oDtDGDTKWCBm	None	...	None	fwdbtn	fw

Data Cleaning: streaming_data:

'streaming_data' offers an extensive log of each individual streaming session, including session durations initially measured in milliseconds. To enhance clarity, these durations will be converted into minutes.

```
In [4]: # Calculates and adds a new column "minutes_played" by converting "ms_played" from
streaming_data['minutes_played'] = streaming_data['ms_played'] / (1000 * 60)

# Prints the updated DataFrame, which includes the original "ms_played" and the ne
print(streaming_data[['ms_played', 'minutes_played']])
```

```
      ms_played  minutes_played
0      173493      2.891550
1      181480      3.024667
2      205240      3.420667
3      189680      3.161333
4      274200      4.570000
...
299339      2600      0.043333
299340     140341      2.339017
299341      3436      0.057267
299342      69022      1.150367
299343     138251      2.304183
```

[299344 rows x 2 columns]

```
In [5]: # Sorts the DataFrame from greatest to lowest based on 'minutes_played'
streaming_data_sorted = streaming_data.sort_values(by='minutes_played', ascending=

# Displays the sorted DataFrame
streaming_data_sorted
```

Out[5]:

entry	user_agent_decrypted	master_metadata_track_name	master_metadata_album_artist_name	master_metadata_i
US	unknown	White Noise 3 Hour Long	Erik Eriksson	Whi
US	unknown	White Noise 3 Hour Long	Erik Eriksson	Whi
US	unknown	White Noise 2 Hour Long	Erik Eriksson	Whi
US	unknown	White Noise 3 Hour Long	Erik Eriksson	Whi
US	unknown	None	None	
...	
US	unknown	I Think I Smell A Rat	The White Stripes	
US	unknown	Pretend I'm Gay	The Growlers	
US	unknown	delicate creature	iogi	
US	unknown	I'm In The Band	The Hellacopters	
US	unknown	I'm Not Making out With You	Surf Curse	

Most Played Tracks:

When sorted in descending order, the tracks with the highest 'minutes_played' values include 'White Noise 3 Hour Long.' It's essential to focus on actual songs and exclude sound machines.

Filtering out tracks with durations exceeding 20 minutes is crucial in limiting the analysis to songs. This approach helps identify the songs that were listened to the most.

```
In [6]: import pandas as pd

# Creates a new DataFrame, songs_df, by filtering rows where the song duration are
# This helps narrow down the analysis to shorter songs for further examination.

songs_df = streaming_data[streaming_data['minutes_played'] <= 20]

# Print the updated DataFrame
songs_df
```

Out [6]:

	ts	platform	ms_played	conn_country	user_agent_decrypted	master_metadata_track_na
0	2020-02-04 01:52:32+00:00	iOS 13.3 (iPhone11,8)	173493	US	unknown	Fluorescent Adolesc
1	2020-02-04 01:55:37+00:00	iOS 13.3 (iPhone11,8)	181480	US	unknown	Only Ones Who Kn
2	2020-02-04 01:59:02+00:00	iOS 13.3 (iPhone11,8)	205240	US	unknown	Do Me a Fav
3	2020-02-04 02:02:13+00:00	iOS 13.3 (iPhone11,8)	189680	US	unknown	This House Is a Ciri
4	2020-02-04 02:06:56+00:00	iOS 13.3 (iPhone11,8)	274200	US	unknown	If You Were There, Bew
...	
299339	2018-04-18 21:14:55+00:00	iOS 11.3 (iPhone8,2)	2600	US	unknown	Rubber & Br
299340	2018-04-18	iOS 11.3	140341	US	unknown	The Daisv Ch

Data Aggregation:

Summarizes the cumulative minutes played for individual songs while preserving the initial data set for each column group and adding a 'month' and 'year' column.

```
In [7]: import pandas as pd

# Removed rows where 'master_metadata_track_name' was equal to 'None'
songs_df = songs_df[songs_df['master_metadata_track_name'] != 'None']

# Grouped the data by 'master_metadata_track_name' and calculated the sum of 'minutes_played'
grouped_df = songs_df.groupby('master_metadata_track_name').agg({
    'minutes_played': 'sum', # Sum of minutes_played
    'ts': 'count', # Count of rows (number of plays)
    'platform': 'first', # Kept the first platform in the group
    'conn_country': 'first', # Kept the first conn_country in the group
    'user_agent_decrypted': 'first', # Kept the first user_agent_decrypted in the group
    'master_metadata_album_artist_name': 'first', # Kept the first artist name in the group
    'master_metadata_album_album_name': 'first', # Kept the first album name in the group
    'spotify_track_uri': 'first', # Kept the first track URI in the group
}).reset_index()

# Sorted the grouped DataFrame by 'minutes_played' in descending order
songs_grouped_df = grouped_df.sort_values(by='minutes_played', ascending=False)

# Convert 'ts' column to datetime format
songs_grouped_df['ts'] = pd.to_datetime(streaming_data['ts'])

# Create 'year' column
songs_grouped_df['year'] = streaming_data['ts'].dt.year

# Create 'month' column using strftime
songs_grouped_df['month'] = streaming_data['ts'].dt.strftime('%B')

# Display the updated DataFrame
songs_grouped_df
```

Out[7]:

	master_metadata_track_name	minutes_played	ts	platform	conn_country	user_agent_dec
24773	Star Treatment	1483.573367	2021-04-29 22:44:32+00:00	iOS 13.3 (iPhone11,8)	US	ur
12651	I'll Come Too	1482.494117	2020-04-25 06:15:19+00:00	iOS 13.3.1 (iPhone11,8)	US	ur
24281	Someday	1176.526867	2021-04-27 21:53:35+00:00	iOS 13.3.1 (iPhone11,8)	US	ur
2589	Batphone	963.282333	2020-02-28 22:30:08+00:00	iOS 13.3.1 (iPhone11,8)	US	ur
14947	Leave It In My Dreams	949.748350	2020-05-08 22:30:26+00:00	iOS 13.3 (iPhone11,8)	US	ur
...
4637	Changing	0.000000	2020-03-10 03:14:01+00:00	iOS 14.6 (iPhone11,8)	US	ur
12638	I'll Be Somewhere	0.000000	2020-04-25 06:07:40+00:00	ios	US	ur
1168	All About That Bass	0.000000	2020-02-14 22:55:42+00:00	iOS 14.6 (iPhone11,8)	US	ur
27736	Three Times A Lady	0.000000	2021-05-24 17:42:59+00:00	ios	US	ur
3407	Blueberry Hill	0.000000	2020-03-04 21:21:47+00:00	iOS 14.2 (iPhone11,8)	US	ur

32279 rows × 11 columns

Top Songs Played:

```
In [8]: # Rank the rows based on the 'count' column
songs_grouped_df['rank'] = songs_grouped_df['minutes_played'].rank(ascending=False)

# Reorder columns to make 'rank' the first column
column_order = ['rank'] + [col for col in songs_grouped_df.columns if col != 'rank']
songs_grouped_df = songs_grouped_df[column_order]

# Sort the DataFrame based on the rank
songs_grouped_df = songs_grouped_df.sort_values(by='rank')
songs_grouped_df
```

Out [8]:

	rank	master_metadata_track_name	minutes_played	ts	platform	conn_country	user_id
24773	1.0	Star Treatment	1483.573367	2021-04-29 22:44:32+00:00	iOS 13.3 (iPhone11,8)	US	
12651	2.0	I'll Come Too	1482.494117	2020-04-25 06:15:19+00:00	iOS 13.3.1 (iPhone11,8)	US	
24281	3.0	Someday	1176.526867	2021-04-27 21:53:35+00:00	iOS 13.3.1 (iPhone11,8)	US	
2589	4.0	Batphone	963.282333	2020-02-28 22:30:08+00:00	iOS 13.3.1 (iPhone11,8)	US	
14947	5.0	Leave It In My Dreams	949.748350	2020-05-08 22:30:26+00:00	iOS 13.3 (iPhone11,8)	US	
...
5217	31395.0	Come Sail Away	0.000000	2020-03-12 15:38:43+00:00	iOS 14.8.1 (iPhone11,8)	US	
13951	31395.0	Joseph, Better You Than Me	0.000000	2020-05-01 16:58:56+00:00	OS X 10.12.5 [x86_8]	US	
6896	31395.0	Don't Carry It All	0.000000	2020-03-21 23:45:46+00:00	iOS 14.6 (iPhone11,8)	US	
25593	31395.0	Swamp Fruit	0.000000	2021-05-03 18:01:40+00:00	iOS 12.3.1 (iPhone11,8)	US	
3407	31395.0	Blueberry Hill	0.000000	2020-03-04 21:21:47+00:00	iOS 14.2 (iPhone11,8)	US	

32279 rows × 12 columns

Vizualizations:

Streams Over Time

```
In [9]: pio.renderers.default = 'notebook'

# Assuming your DataFrame is named streaming_data
# Convert the 'ts' column to a datetime object
streaming_data['ts'] = pd.to_datetime(streaming_data['ts'])

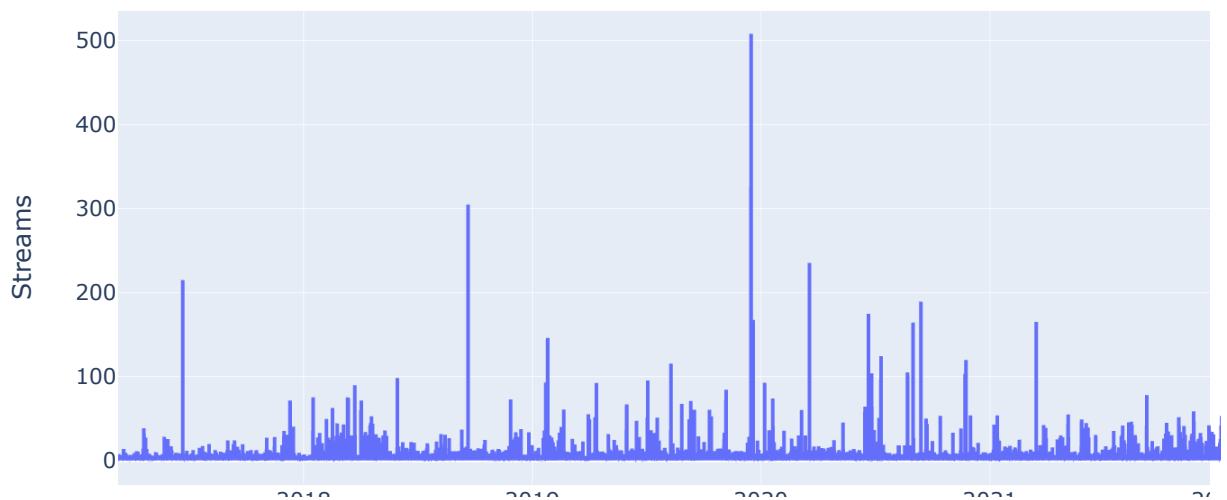
# Group the data by 'ts' and sum the 'minutes_played' for each timestamp
streams_over_time = streaming_data.groupby('ts')['minutes_played'].sum().reset_index()

# Create an interactive line plot for streams over time with Plotly
fig = px.line(streams_over_time, x='ts', y='minutes_played', title='Streams Over Time')
fig.update_xaxes(title_text='Date')
fig.update_yaxes(title_text='Streams')

# Enable zooming and panning
fig.update_xaxes(rangeslider_visible=True)

# Show the plot
fig.show()
```

Streams Over Time



Top Played Songs

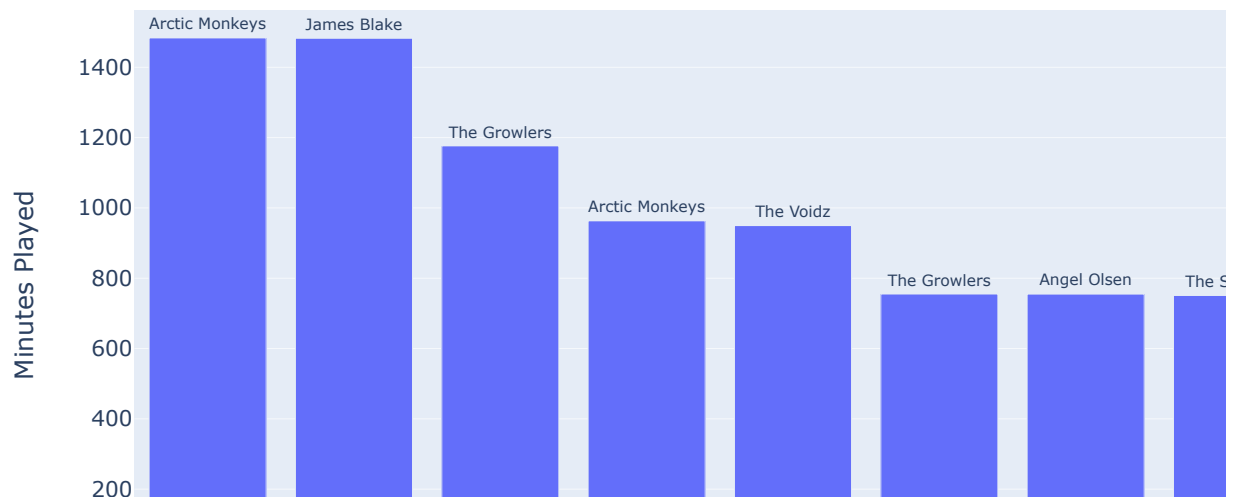
```
In [10]: # Assuming songs_grouped_df is your actual DataFrame with the columns mentioned
# Sort the DataFrame by 'minutes_played' to get the most played songs and limit to
top_songs = songs_grouped_df.sort_values('minutes_played', ascending=False).head(10)

# Create the bar chart using Plotly Express
fig = px.bar(top_songs,
             x='master_metadata_track_name',
             y='minutes_played',
             text='master_metadata_album_artist_name',
             title='Top 10 Most Played Songs')

# Customize the layout to include the artist names in the bar chart
fig.update_traces(texttemplate='%{text}', textposition='outside')
fig.update_layout(uniformtext_minsize=8, uniformtext_mode='hide',
                  xaxis_title='Song Name',
                  yaxis_title='Minutes Played')

# Show the figure
fig.show()
```

Top 10 Most Played Songs



Showing Plot Between 'rank' and 'minutes_played'

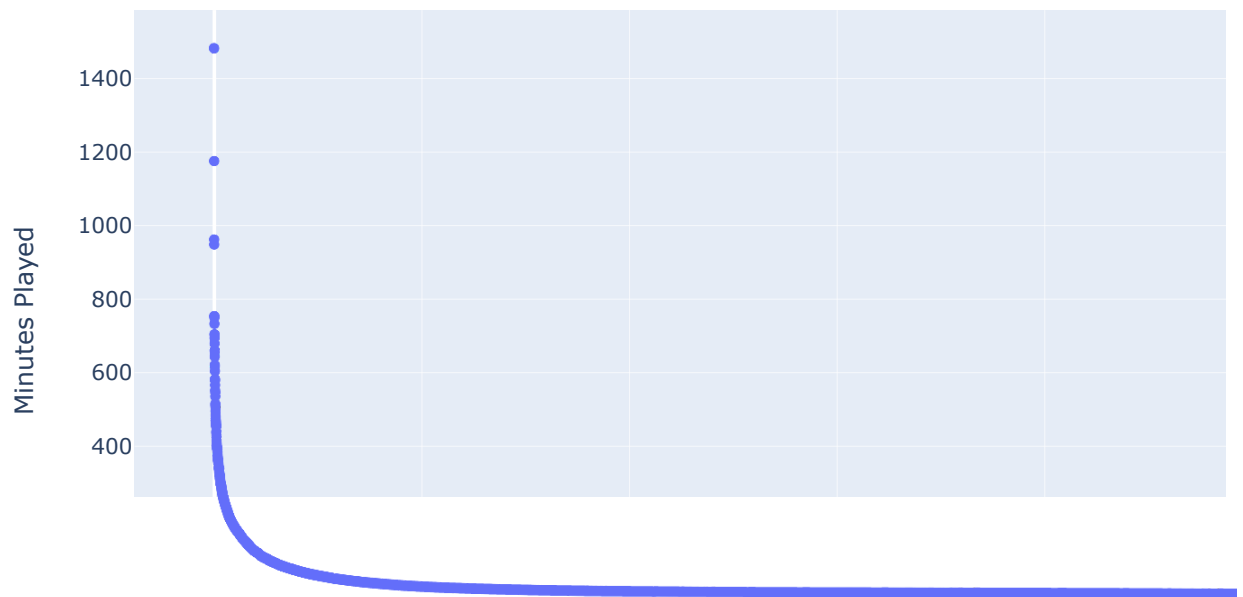
```
In [11]: # Creates an interactive scatter plot between 'rank' and 'minutes_played' using Plotly
fig = px.scatter(
    songs_grouped_df,
    x='rank',
    y='minutes_played',
    title='Scatter Plot: Rank vs. Minutes Played',
    hover_name='master_metadata_track_name', # Track name shown on hover
    hover_data=['master_metadata_album_artist_name'], # Artist name shown on hover
)

fig.update_xaxes(title_text='Rank')
fig.update_yaxes(title_text='Minutes Played')

# Enable zooming and panning
fig.update_xaxes(type='linear')
fig.update_yaxes(type='linear')

# Show the plot
fig.show()
```

Scatter Plot: Rank vs. Minutes Played



Accessing Spotify API

```
In [12]: username = 'juliorod323'
client_id = '3d2dd78837114f7ab9c5bb16c14ee9ba'
client_secret = '706144568b26419d90d293a14904396b'
redirect_uri = 'http://localhost:7777/callback'
scope = 'user-read-recently-played'

token = util.prompt_for_user_token(username=username,
                                   scope=scope,
                                   client_id=client_id,
                                   client_secret=client_secret,
                                   redirect_uri=redirect_uri)
```

```
In [13]: import requests

def get_id(track_name: str, token: str) -> str:
    headers = {
        'Accept': 'application/json',
        'Content-Type': 'application/json',
        'Authorization': f'Bearer {token}',
    }
    params = {
        'q': track_name,
        'type': 'track',
    }
    try:
        response = requests.get('https://api.spotify.com/v1/search', headers=headers, params=params)
        data = response.json()
        first_result = data['tracks']['items'][0]
        track_id = first_result['id']
        return track_id
    except:
        return None
```

```
In [14]: # Create a Spotify API client
client_credentials_manager = SpotifyClientCredentials(client_id=client_id, client_secret=client_secret)
sp = spotipy.Spotify(client_credentials_manager=client_credentials_manager)

def get_features(track_id: str, token: str) -> dict:
    """
    Get Spotify track audio features.

    Args:
        track_id (str): The Spotify track ID.
        token (str): Your Spotify access token.

    Returns:
        dict: Dictionary containing track audio features.
    """
    try:
        features = sp.audio_features(track_id)
        return features[0] if features else None
    except:
        return None
```

```
In [15]: songs_grouped_df.columns
```

```
Out[15]: Index(['rank', 'master_metadata_track_name', 'minutes_played', 'ts',
               'platform', 'conn_country', 'user_agent_decrypted',
               'master_metadata_album_artist_name', 'master_metadata_album_album_name',
               'spotify_track_uri', 'year', 'month'],
              dtype='object')
```

API Obstacles

I have run into request limitation with the Spotify API, so as a precaution, I will create the top_song_100 df that contains the top 100 most played songs. I want to see the attributes for my top 100 songs played.

```
In [16]: # Creates a condensed version of the songs_grouped_df DataFrame
top_songs_100 = songs_grouped_df.sort_values('minutes_played', ascending=False).head(100)
```

Out[16]:

	rank	master_metadata_track_name	minutes_played	ts	platform	conn_country	user_age
24773	1.0	Star Treatment	1483.573367	2021-04-29 22:44:32+00:00	iOS 13.3 (iPhone11,8)	US	
12651	2.0	I'll Come Too	1482.494117	2020-04-25 06:15:19+00:00	iOS 13.3.1 (iPhone11,8)	US	
24281	3.0	Someday	1176.526867	2021-04-27 21:53:35+00:00	iOS 13.3.1 (iPhone11,8)	US	
2589	4.0	Batphone	963.282333	2020-02-28 22:30:08+00:00	iOS 13.3.1 (iPhone11,8)	US	
14947	5.0	Leave It In My Dreams	949.748350	2020-05-08 22:30:26+00:00	iOS 13.3 (iPhone11,8)	US	
...
28359	96.0	Trouble	364.247700	2021-05-31 00:16:11+00:00	iOS 13.3.1 (iPhone11,8)	US	
10079	97.0	Going Gets Tough	362.715700	2020-04-09 02:04:29+00:00	iOS 13.3.1 (iPhone11,8)	US	
3762	98.0	British Bombs	362.609383	2020-03-05 23:58:30+00:00	OS X 10.14.6 [x86_64]	US	
30484	99.0	Why'd You Only Call Me When You're High?	359.919717	2021-06-10 00:38:03+00:00	iOS 13.3.1 (iPhone11,8)	US	
18403	100.0	Nights	359.622617	2021-03-13 08:41:57+00:00	iOS 13.3.1 (iPhone11,8)	US	

100 rows × 12 columns

```
In [17]: import pandas as pd

# Defines a list to store the audio features for each track
audio_features_list = []

# Loops through the 'master_metadata_track_name' column
for track_name in top_songs_100['master_metadata_track_name']:
    # Get the Spotify track ID for the current track
    track_id = get_id(track_name, token)

    # If a track ID is found, retrieves audio features
    if track_id:
        track_features = get_features(track_id, token)

        # If audio features are found, add them to the list
        if track_features:
            audio_features_list.append(track_features)

# Creates a DataFrame from the list of audio features
audio_features_df = pd.DataFrame(audio_features_list)

# Displays the resulting DataFrame with audio features
print(audio_features_df)
```

	danceability	energy	key	loudness	mode	speechiness	acousticness	\
0	0.581	0.767	7	-5.026	0	0.0527	0.24300	
1	0.661	0.472	7	-7.889	1	0.1640	0.75700	
2	0.539	0.521	9	-7.460	1	0.0299	0.00425	
3	0.675	0.681	5	-5.728	0	0.0356	0.31200	
4	0.600	0.746	4	-4.696	0	0.0355	0.02260	
..	
95	0.470	0.623	0	-5.655	1	0.0302	0.39200	
96	0.501	0.769	8	-6.257	1	0.0326	0.22500	
97	0.674	0.865	6	-6.167	1	0.0412	0.04620	
98	0.691	0.631	2	-6.478	1	0.0368	0.04830	
99	0.457	0.551	5	-9.360	0	0.1670	0.42700	

	instrumentalness	liveness	valence	tempo	type	\
0	0.001310	0.1410	0.673	120.070	audio_features	
1	0.002750	0.1920	0.206	73.986	audio_features	
2	0.000000	0.3670	0.679	106.035	audio_features	
3	0.000000	0.3040	0.675	106.521	audio_features	
4	0.011900	0.1090	0.614	118.551	audio_features	
..	
95	0.000439	0.0992	0.298	77.861	audio_features	
96	0.000000	0.1090	0.737	162.047	audio_features	
97	0.000000	0.3810	0.694	131.975	audio_features	
98	0.000011	0.1040	0.800	92.004	audio_features	
99	0.000001	0.1130	0.428	89.870	audio_features	

	id	uri	\
0	0b93tWwuoAC0nXe1CfR30I	spotify:track:0b93tWwuoAC0nXe1CfR30I	
1	6EvUAsfncetT0RcWshHCbl	spotify:track:6EvUAsfncetT0RcWshHCbl	
2	7hm4HTk9encxT0LYC0J6oI	spotify:track:7hm4HTk9encxT0LYC0J6oI	
3	7aiKdAM9WYW3GzWSA90XIl	spotify:track:7aiKdAM9WYW3GzWSA90XIl	
4	3lu6rUeIEXGrYVoh10U7eu	spotify:track:3lu6rUeIEXGrYVoh10U7eu	
..	
95	5n0CTysih20NYdT2S0Wpe8	spotify:track:5n0CTysih20NYdT2S0Wpe8	
96	6Et9MYUuK20kYE3ce2F0oK	spotify:track:6Et9MYUuK20kYE3ce2F0oK	
97	4ZdmTNaBTED8n9AQE0YaX	spotify:track:4ZdmTNaBTED8n9AQE0YaX	
98	086myS9r57YsLbJpU0TgK9	spotify:track:086myS9r57YsLbJpU0TgK9	
99	7eqoqGkKwg0aWNNHx90uEZ	spotify:track:7eqoqGkKwg0aWNNHx90uEZ	

	track_href	\
0	https://api.spotify.com/v1/tracks/0b93tWwuoAC0...	(https://api.spotify.com/v1/tracks/0b93tWwuoAC0...)
1	https://api.spotify.com/v1/tracks/6EvUAsfncetT...	(https://api.spotify.com/v1/tracks/6EvUAsfncetT...)
2	https://api.spotify.com/v1/tracks/7hm4HTk9encx...	(https://api.spotify.com/v1/tracks/7hm4HTk9encx...)
3	https://api.spotify.com/v1/tracks/7aiKdAM9WYW3...	(https://api.spotify.com/v1/tracks/7aiKdAM9WYW3...)
4	https://api.spotify.com/v1/tracks/3lu6rUeIEXGr...	(https://api.spotify.com/v1/tracks/3lu6rUeIEXGr...)
..
95	https://api.spotify.com/v1/tracks/5n0CTysih20N...	(https://api.spotify.com/v1/tracks/5n0CTysih20N...)
96	https://api.spotify.com/v1/tracks/6Et9MYUuK20k...	(https://api.spotify.com/v1/tracks/6Et9MYUuK20k...)
97	https://api.spotify.com/v1/tracks/4ZdmTNaBTED...	(https://api.spotify.com/v1/tracks/4ZdmTNaBTED...)
98	https://api.spotify.com/v1/tracks/086myS9r57Ys...	(https://api.spotify.com/v1/tracks/086myS9r57Ys...)
99	https://api.spotify.com/v1/tracks/7eqoqGkKwg0a...	(https://api.spotify.com/v1/tracks/7eqoqGkKwg0a...)

analysis_url	duration_ms	\
--------------	-------------	---

```

0  https://api.spotify.com/v1/audio-analysis/0b93... (https://api.spotify.com/v
1/audio-analysis/0b93...) 354640
1  https://api.spotify.com/v1/audio-analysis/6EvU... (https://api.spotify.com/v
1/audio-analysis/6EvU...) 222400
2  https://api.spotify.com/v1/audio-analysis/7hm4... (https://api.spotify.com/v
1/audio-analysis/7hm4...) 183440
3  https://api.spotify.com/v1/audio-analysis/7aiK... (https://api.spotify.com/v
1/audio-analysis/7aiK...) 271613
4  https://api.spotify.com/v1/audio-analysis/31u6... (https://api.spotify.com/v
1/audio-analysis/31u6...) 239442
..
95 https://api.spotify.com/v1/audio-analysis/5n0C... (https://api.spotify.com/v
1/audio-analysis/5n0C...) 225973
96 https://api.spotify.com/v1/audio-analysis/6Et9... (https://api.spotify.com/v
1/audio-analysis/6Et9...) 208613
97 https://api.spotify.com/v1/audio-analysis/4Zdm... (https://api.spotify.com/v
1/audio-analysis/4Zdm...) 297715
98 https://api.spotify.com/v1/audio-analysis/086m... (https://api.spotify.com/v
1/audio-analysis/086m...) 161124
99 https://api.spotify.com/v1/audio-analysis/7eqo... (https://api.spotify.com/v
1/audio-analysis/7eqo...) 307151

```

```

time_signature
0 4
1 4
2 4
3 4
4 4
..
95 4
96 4
97 4
98 4
99 4

```

[100 rows x 18 columns]

Limiting "Features" in Analysis:

```
In [18]: # Columns to drop
columns_to_drop = ['mode', 'valence', 'analysis_url', 'time_signature', 'uri', 'track_name']

# Drop the specified columns from the DataFrame, ignoring columns that do not exist
audio_features_df = audio_features_df.drop(columns=columns_to_drop, errors='ignore')

# Display the resulting DataFrame
print(audio_features_df)
```

	danceability	energy	loudness	speechiness	acousticness	\
0	0.581	0.767	-5.026	0.0527	0.24300	
1	0.661	0.472	-7.889	0.1640	0.75700	
2	0.539	0.521	-7.460	0.0299	0.00425	
3	0.675	0.681	-5.728	0.0356	0.31200	
4	0.600	0.746	-4.696	0.0355	0.02260	
..	
95	0.470	0.623	-5.655	0.0302	0.39200	
96	0.501	0.769	-6.257	0.0326	0.22500	
97	0.674	0.865	-6.167	0.0412	0.04620	
98	0.691	0.631	-6.478	0.0368	0.04830	
99	0.457	0.551	-9.360	0.1670	0.42700	

	instrumentalness	liveness	tempo	duration_ms
0	0.001310	0.1410	120.070	354640
1	0.002750	0.1920	73.986	222400
2	0.000000	0.3670	106.035	183440
3	0.000000	0.3040	106.521	271613
4	0.011900	0.1090	118.551	239442
..
95	0.000439	0.0992	77.861	225973
96	0.000000	0.1090	162.047	208613
97	0.000000	0.3810	131.975	297715
98	0.000011	0.1040	92.004	161124
99	0.000001	0.1130	89.870	307151

[100 rows x 9 columns]

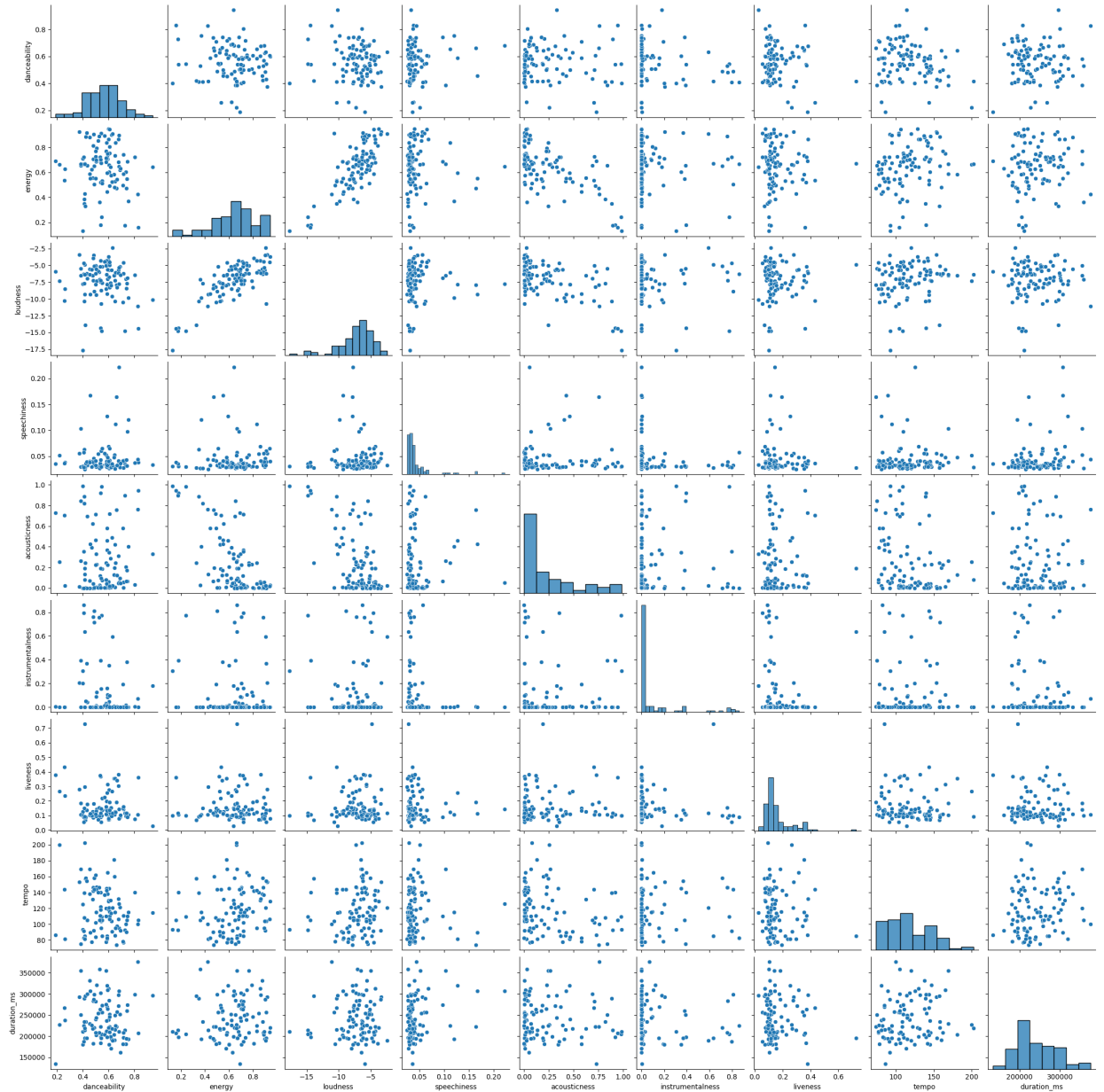
```
In [19]: audio_features_df.columns
```

```
Out[19]: Index(['danceability', 'energy', 'loudness', 'speechiness', 'acousticness',
               'instrumentalness', 'liveness', 'tempo', 'duration_ms'],
              dtype='object')
```


Pair Plot

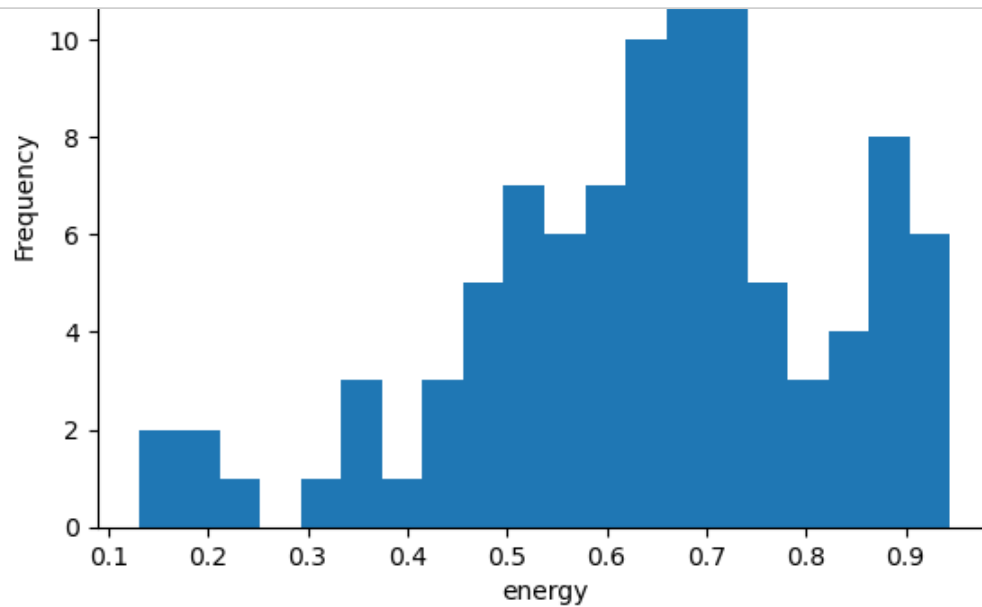
```
In [20]: import seaborn as sns
sns.pairplot(audio_features_df)
```

```
Out[20]: <seaborn.axisgrid.PairGrid at 0x17776f410>
```



Spotify Track "Features" Plotted Along Frequency

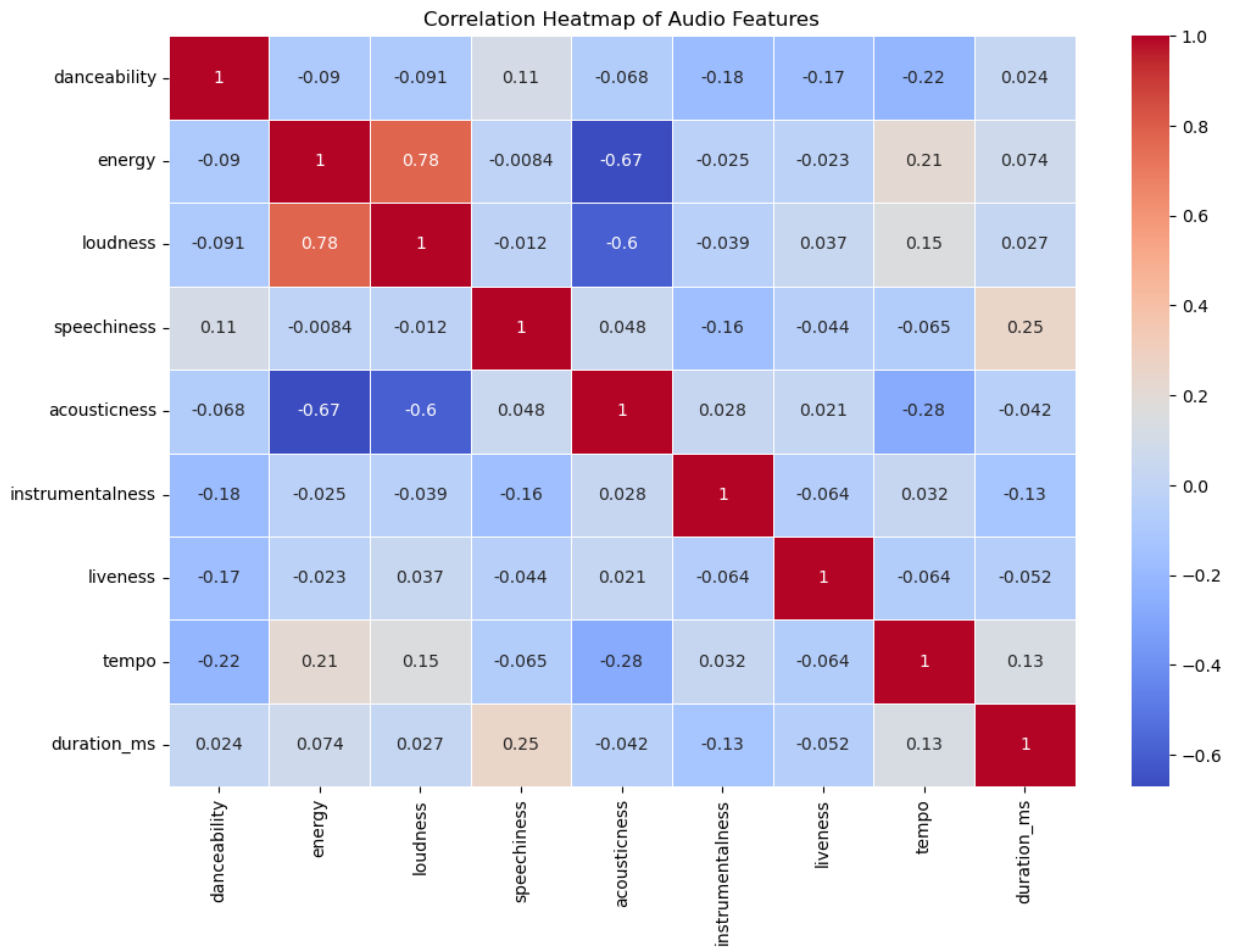
```
In [21]: import matplotlib.pyplot as plt
for feature in audio_features_df.columns:
    plt.hist(audio_features_df[feature], bins=20)
    plt.xlabel(feature)
    plt.ylabel('Frequency')
    plt.title(f'Distribution of {feature}')
    plt.show()
```



Coorelation Heatmap:

```
In [22]: import seaborn as sns
import matplotlib.pyplot as plt
corr_matrix = audio_features_df.corr()
plt.figure(figsize=(12, 8))
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', linewidths=.5)
plt.title('Correlation Heatmap of Audio Features')
```

Out[22]: Text(0.5, 1.0, 'Correlation Heatmap of Audio Features')



In []: