

spotify_scraper

April 30, 2024

1 Spotify API Scraping for Creating Datasets

1.0.1 I used the code (but slightly modified it to my needs to include the `release_date`) from [this lovely article!](#)

```
[1]: # pip install spotifyscraper

[2]: # pip install pathlib

[3]: # pip install ruamel-yaml

[4]: # pip install spotipy

[5]: import spotipy
from spotipy.oauth2 import SpotifyClientCredentials
from spotipy.oauth2 import SpotifyOAuth
import pandas as pd
import time

[7]: cid = '6ff98eca336346ee942d607cc2d23879'
secret = '7d7292ad3cf0420e8f270e7d049f40ba'
client_credentials_manager = SpotifyClientCredentials(client_id=cid,
↳client_secret=secret)
sp = spotipy.Spotify(client_credentials_manager = client_credentials_manager)

[8]: # Pagination (to extract more than 100 songs at a time)
def call_playlist(creator, playlist_id):
    # Step 1: Initialize DataFrame and other variables
    playlist_features_list = ["artist", "album", "track_name", "track_id",
↳"release_date", "danceability", "energy", "key", "loudness", "mode",
↳"speechiness", "instrumentalness", "liveness", "valence", "tempo",
↳"duration_ms", "time_signature"]
    playlist_df = pd.DataFrame(columns=playlist_features_list)
    offset = 0
    total_tracks = sp.user_playlist_tracks(creator, playlist_id)["total"]
```

```

    # Step 2: Fetch tracks with pagination
    while offset < total_tracks:
        playlist = sp.user_playlist_tracks(creator, playlist_id,
        →offset=offset)["items"]
        for track in playlist:
            playlist_features = {}
            playlist_features["artist"] =
            →track["track"]["album"]["artists"][0]["name"]
            playlist_features["album"] = track["track"]["album"]["name"]
            playlist_features["track_name"] = track["track"]["name"]
            playlist_features["track_id"] = track["track"]["id"]
            playlist_features["release_date"] =
            →track["track"]["album"]["release_date"]
            audio_features = sp.audio_features(playlist_features["track_id"])[0]
            for feature in playlist_features_list[5:]:
                playlist_features[feature] = audio_features[feature]
            track_df = pd.DataFrame(playlist_features, index=[0])
            playlist_df = pd.concat([playlist_df, track_df], ignore_index=True)
            offset += 100
    # Step 3: Return DataFrame
    return playlist_df

```

```

[9]: # Function to fetch audio features with retry logic
def fetch_audio_features(track_id):
    retries = 10 # Maximum number of retry attempts
    for _ in range(retries):
        try:
            return sp.audio_features(track_id)[0]
        except spotipy.SpotifyException as e:
            if e.http_status == 429:
                # Retry after a fixed delay
                retry_after = int(e.headers.get('Retry-After', 10)) # Default
                →to 10 seconds if no Retry-After header
                print(f"Rate limited. Retrying after {retry_after} seconds...")
                time.sleep(retry_after)
            else:
                raise # Re-raise the exception if it's not a 429 error
    raise Exception("Max retries reached, unable to fetch audio features")

```

```

[10]: # https://open.spotify.com/playlist/5yAPuepGnApi5yc4QoZMDl
# "old" Playlist compiled by "emmabittinger" (this is a test)
old_playlist = call_playlist("spotify", "5yAPuepGnApi5yc4QoZMDl")

```

```

[11]: old_playlist.head()

```

```
[11]:
```

	artist	album \
0	Various Artists	Cars (VersiÃ§Ã£o de ColecciÃ§Ã£o)
1	Anthem Lights	Covers Part IV
2	Bruce Springsteen	Born In The U.S.A.
3	Lynyrd Skynyrd	Second Helping (Expanded Edition)
4	Rick Astley	Whenever You Need Somebody

	track_name	track_id \
0	Life Is A Highway - From "Cars"/Soundtrack Ver...	1QezVl06xBzPfgJ2HXST5d
1	Don't Stop Believing	0wBqAqxUygzHrUgw0MTJ6J
2	Born in the U.S.A.	0d0g1ySSI7NkpAe89Zo0b9
3	Sweet Home Alabama	7e89621JPkKaeDSTQ3avtg
4	Never Gonna Give You Up	7GhIk7I1098yCjg4BQjzvb

	release_date	danceability	energy	key	loudness	mode	speechiness \
0	2006-06-06	0.561	0.932	5	-5.475	1	0.0584
1	2015-07-17	0.516	0.391	10	-7.319	1	0.0315
2	1984-06-04	0.398	0.952	4	-6.042	1	0.0610
3	1974-04-15	0.596	0.605	7	-12.145	1	0.0255
4	1987-12-08	0.727	0.939	8	-11.855	1	0.0369

	instrumentalness	liveness	valence	tempo	duration_ms	time_signature
0	0	0.1810	0.670	103.062	275640	4
1	0	0.1440	0.395	117.873	218644	4
2	0.000077	0.1000	0.584	122.093	278680	4
3	0.000331	0.0863	0.886	97.798	283800	4
4	0.000044	0.1510	0.916	113.330	212827	4

```
[12]: # https://open.spotify.com/playlist/66kbLWdmxWuMYeByFkqADT
```

```
# "throwbaccc" Playlist compiled by "emmabittinger"
```

```
throbaccc_playlist = call_playlist("spotify", "66kbLWdmxWuMYeByFkqADT")
```

```
[14]: throbaccc_playlist.to_csv("throbaccc_playlist.csv")
```

```
[15]: # https://open.spotify.com/playlist/7dBWDKw7I8kZy0td1VYFIY
```

```
# "Songs Everyone Knows the Words To" Playlist compiled by "Ava Montgomery"
```

```
long_playlist = call_playlist("spotify", "7dBWDKw7I8kZy0td1VYFIY")
```

```
[16]: long_playlist.to_csv("long_playlist.csv")
```

```
[17]: throbaccc_playlist.head()
```

```
[17]:
```

	artist	album	track_name \
0	Miley Cyrus	The Time Of Our Lives	Party In The U.S.A.
1	Rihanna	Loud	What's My Name?
2	Train	Hey, Soul Sister	Hey, Soul Sister
3	Justin Bieber	My World	One Time

	track_id	release_date	danceability	energy	key	loudness	\
0	5Q0Nhxo0l2bP3pNjpGJwV1	2009-01-01	0.652	0.698	10	-4.667	
1	5FTCKvxzqy72ceS4Ujux4N	2010-11-16	0.692	0.786	2	-2.959	
2	OKpfYajJVVGgQ32Dby7e9i	2009-08-06	0.675	0.885	1	-4.432	
3	6eDApnV9Jdb1nYah0lbbUh	2009-01-01	0.691	0.853	1	-2.528	
4	Obg6otrW5gxNnlCqrCrXyd	2010-05-28	0.754	0.804	4	-3.177	

	mode	speechiness	instrumentalness	liveness	valence	tempo	duration_ms	\
0	0	0.0420	0.000115	0.0886	0.470	96.021	202067	
1	1	0.0690	0.000000	0.0797	0.583	100.025	263173	
2	0	0.0436	0.000000	0.0860	0.768	97.030	216667	
3	0	0.0372	0.000071	0.0820	0.762	145.999	215867	
4	1	0.0853	0.000000	0.0329	0.818	119.968	203867	

	time_signature
0	4
1	4
2	4
3	4
4	4

```
[18]: throbaccc_playlist.describe()
```

```
[18]:
```

	danceability	energy	loudness	speechiness	instrumentalness	\
count	330.000000	330.000000	330.000000	330.000000	330.000000	
mean	0.647394	0.738092	-5.048933	0.075087	0.005875	
std	0.117276	0.164847	1.792816	0.061893	0.056663	
min	0.327000	0.056500	-15.099000	0.025400	0.000000	
25%	0.578500	0.676250	-5.889250	0.037700	0.000000	
50%	0.658500	0.773000	-4.823500	0.051100	0.000000	
75%	0.724750	0.857000	-3.848500	0.088050	0.000004	
max	0.979000	0.981000	-1.644000	0.449000	0.871000	

	liveness	valence	tempo
count	330.000000	330.000000	330.000000
mean	0.177032	0.587910	121.181464
std	0.128664	0.206436	25.593651
min	0.019300	0.076500	65.043000
25%	0.088675	0.428000	101.734500
50%	0.127000	0.619000	122.624000
75%	0.248750	0.748000	132.023000
max	0.758000	0.965000	194.077000

```
[19]: long_playlist.head()
```

```
[19]:
```

	artist	album \
0	Coldplay	Viva La Vida or Death and All His Friends
1	Rihanna	Good Girl Gone Bad
2	*NSYNC	No Strings Attached
3	Train	Save Me, San Francisco (Golden Gate Edition)
4	Carrie Underwood	Some Hearts

	track_name	track_id	release_date	danceability \
0	Viva La Vida	1mea3bSkSGXuIRvnydlB5b	2008-05-26	0.486
1	Umbrella	2yPoXC7BSIUrucMdK5PzV	2007-01-01	0.583
2	Bye Bye Bye	62b0mKYxYg7dhrC6gH9vFn	2000-03-21	0.610
3	Hey, Soul Sister	4HlFJV71xXKIGcU3kRyttv	2010-12-01	0.673
4	Before He Cheats	OZUo4YjG4saFnEJhdWp9Bt	2005-11-14	0.519

	energy	key	loudness	mode	speechiness	instrumentalness	liveness \
0	0.617	5	-7.115	0	0.0287	0.000003	0.1090
1	0.829	1	-4.603	1	0.1340	0.000000	0.0426
2	0.926	8	-4.843	0	0.0479	0.001200	0.0821
3	0.886	1	-4.440	0	0.0431	0.000000	0.0826
4	0.749	6	-3.318	0	0.0405	0.000000	0.1190

	valence	tempo	duration_ms	time_signature
0	0.417	138.015	242373	4
1	0.575	174.028	275987	4
2	0.861	172.638	200400	4
3	0.795	97.012	216773	4
4	0.290	147.905	199947	4

```
[20]: long_playlist.describe()
```

```
[20]:
```

	danceability	energy	loudness	speechiness	instrumentalness \
count	345.000000	345.000000	345.000000	345.000000	345.000000
mean	0.614093	0.704965	-5.682510	0.079451	0.007984
std	0.139569	0.171480	2.219701	0.069564	0.049434
min	0.209000	0.111000	-18.064000	0.024900	0.000000
25%	0.522000	0.594000	-6.682000	0.038200	0.000000
50%	0.614000	0.728000	-5.223000	0.054500	0.000000
75%	0.717000	0.839000	-4.165000	0.085300	0.000090
max	0.967000	0.986000	-1.848000	0.449000	0.616000

	liveness	valence	tempo
count	345.000000	345.000000	345.000000
mean	0.17513	0.508641	125.870136
std	0.13550	0.228687	26.866936
min	0.02100	0.038500	65.997000
25%	0.09010	0.336000	106.970000
50%	0.12000	0.506000	125.072000

75%	0.23700	0.691000	142.673000
max	0.77000	0.969000	199.935000

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