

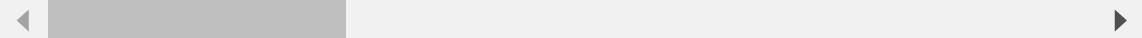
```
In [34]: ┌─ import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
    from sklearn.cluster import KMeans
    from sklearn.preprocessing import StandardScaler
    from sklearn.decomposition import PCA
```

In [35]: ⏷ data=pd.read_csv("spotify.csv")
data

Out[35]:

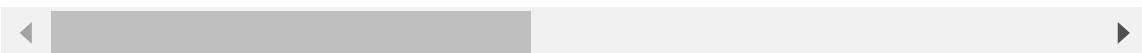
		track_id	track_name	track_artist	track_popularity	
0	6f807x0ima9a1j3VPbc7VN	I Don't Care (with Justin Bieber) - Loud Luxur...	Ed Sheeran	66	2oCs0DGTs	
1	0r7CVbZTWZgbTCYdfa2P31	Memories - Dillon Francis Remix	Maroon 5	67	63rPSO264	
2	1z1Hg7Vb0AhHDiEmnDE79I	All the Time - Don Diablo Remix	Zara Larsson	70	1HoSmj2i	
3	75FpbthrwQmzHIBJLuGdC7	Call You Mine - Keanu Silva Remix	The Chainsmokers	60	1nqYsOef1	
4	1e8PAfcKUYoKkxPhrHqw4x	Someone You Loved - Future Humans Remix	Lewis Capaldi	69	7m7vv9w	
...	
32828	7bxnKAamR3snQ1VGLuVfC1	City Of Lights - Official Radio Edit	Lush & Simon	42	2azRoBBWl	
32829	5Aevni09Em4575077nkWHz	Closer - Sultan & Ned Shepard Remix	Tegan and Sara	20	6kD6KLxj	
32830	7ImMqPP3Q1yfUHvsdn7wEo	Sweet Surrender - Radio Edit	Starkillers	14	0ltWNSY9	
32831	2m69mhnfQ1Oq6IGtXuYhgX	Only For You - Maor Levi Remix	Mat Zo	15	1fGrOkH	
32832	29zWqhca3zt5NsckZqDf6c	Typhoon - Original Mix	Julian Calor	27	0X3mUOm6N	

32833 rows × 23 columns



In [36]: ➔ data.describe()

Out[36]:		track_popularity	danceability	energy	key	loudness	m
	count	32833.000000	32833.000000	32833.000000	32833.000000	32833.000000	32833.000000
	mean	42.477081	0.654850	0.698619	5.374471	-6.719499	0.565
	std	24.984074	0.145085	0.180910	3.611657	2.988436	0.495
	min	0.000000	0.000000	0.000175	0.000000	-46.448000	0.000
	25%	24.000000	0.563000	0.581000	2.000000	-8.171000	0.000
	50%	45.000000	0.672000	0.721000	6.000000	-6.166000	1.000
	75%	62.000000	0.761000	0.840000	9.000000	-4.645000	1.000
	max	100.000000	0.983000	1.000000	11.000000	1.275000	1.000



In [37]: ➔ data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32833 entries, 0 to 32832
Data columns (total 23 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   track_id         32833 non-null   object 
 1   track_name       32828 non-null   object 
 2   track_artist     32828 non-null   object 
 3   track_popularity 32833 non-null   int64  
 4   track_album_id   32833 non-null   object 
 5   track_album_name 32828 non-null   object 
 6   track_album_release_date 32833 non-null   object 
 7   playlist_name    32833 non-null   object 
 8   playlist_id      32833 non-null   object 
 9   playlist_genre   32833 non-null   object 
 10  playlist_subgenre 32833 non-null   object 
 11  danceability     32833 non-null   float64
 12  energy           32833 non-null   float64
 13  key              32833 non-null   int64  
 14  loudness          32833 non-null   float64
 15  mode              32833 non-null   int64  
 16  speechiness       32833 non-null   float64
 17  acousticness      32833 non-null   float64
 18  instrumentalness 32833 non-null   float64
 19  liveness          32833 non-null   float64
 20  valence           32833 non-null   float64
 21  tempo              32833 non-null   float64
 22  duration_ms       32833 non-null   int64 

dtypes: float64(9), int64(4), object(10)
memory usage: 5.8+ MB
```

In [38]: ⏷ data.isnull().sum()

```
Out[38]: track_id          0
track_name         5
track_artist        5
track_popularity    0
track_album_id      0
track_album_name     5
track_album_release_date  0
playlist_name       0
playlist_id          0
playlist_genre        0
playlist_subgenre     0
danceability         0
energy                 0
key                   0
loudness               0
mode                  0
speechiness           0
acousticness          0
instrumentalness      0
liveness                0
valence                 0
tempo                  0
duration_ms            0
dtype: int64
```

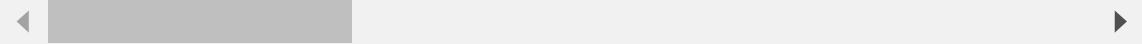
In [39]:

```
data=data.dropna()
data.head()
```

Out[39]:

		track_id	track_name	track_artist	track_popularity	track
0	6f807x0ima9a1j3VPbc7VN	I Don't Care (with Justin Bieber) - Loud Luxur...	Ed Sheeran	66	2oCs0DGTsR098C	
1	0r7CVbZTWZgbTCYdfa2P31	Memories - Dillon Francis Remix	Maroon 5	67	63rPSO264uRjW1;	
2	1z1Hg7Vb0AhHDiEmnDE79I	All the Time - Don Diablo Remix	Zara Larsson	70	1HoSmj2eLcsrRl	
3	75FpbthrwQmzHIBJLuGdC7	Call You Mine - Keanu Silva Remix	The Chainsmokers	60	1nqYsOef1yKKuG	
4	1e8PAfcKUYoKkxPhrHqw4x	Someone You Loved - Future Humans Remix	Lewis Capaldi	69	7m7vv9wlQ4i0LI	

5 rows × 23 columns



```
In [40]: ┌─ duplicates = data.duplicated(subset=["track_name", "track_artist"])
  ┌─ print("Duplicate Rows:")
  ┌─ print(data[duplicates].head())
```

Duplicate Rows:

	track_id	track_name	track_artist			
193	3BOcWxFUURAomDXRSDFve4	Something Real	Armin van Buuren			
209	4TIkSdsNSfqpuq6ZYvCjAz	All You Need To Know (feat. Calle Lehmann)	Gryffin			
232	3sHuIjfAzluc6S9cXoqfqC	Let It Be Me	Steve Aoki			
272	36orMWv2PgvnzXsd5CJ0yL	Post Malone (feat. RANI)	Sam Feldt			
294	7rpyHKSH3dkrsKEgv1eNgv	Woke Up Late (feat. Hailee Steinfeld) - Sam Fe...	Drax Project			
	track_popularity	track_album_id	track_album_name			
193	58	5cqwXF2j9LkvFInBFlnQd3	Balence			
209	68	2IAVHJdaRPFA6MQqXHog75	Gravity			
232	23	5ocW53VBnOpr16EAMOLGet	Let It Be Me			
272	75	45nsubB5EsRVWlx0ED1ET	Post Malone (feat. RANI) [Joe Stone Remix]			
294	56	5VW1WffQj2SqKUhwnNq1xJ	Woke Up Late (feat. Hailee Steinfeld) [Sam Fel...			
	track_album_release_date	playlist_name	playlist_id			
193	25-10-2019	Dance Room	37i9dQZF1DX2ENAPP1Tyed			
209	24-10-2019	Dance Room	37i9dQZF1DX2ENAPP1Tyed			
232	06-09-2019	Cardio	37i9dQZF1DWSJHnPb1f0X3			
272	16-08-2019	Dance Pop Hits	37i9dQZF1DX6pH08wMhkaI			
294	02-05-2019	Dance Pop Hits	37i9dQZF1DX6pH08wMhkaI			
	playlist_genre	key	loudness	mode	speechiness	acousticness
193	pop	...	1	-4.578	1	0.0439
209	pop	...	0	-6.019	1	0.0376
232	pop	...	7	-5.299	1	0.0864
272	pop	...	7	-3.870	1	0.1220
294	pop	...	0	-3.753	1	0.0978
	instrumentalness	liveness	valence	tempo	duration_ms	
193	0.0	0.0693	0.232	127.922	179531	
209	0.0	0.1030	0.219	139.929	238338	
232	0.0	0.1060	0.387	114.098	224061	
272	0.0	0.1050	0.651	107.356	174444	
294	0.0	0.1080	0.499	122.055	196475	

[5 rows x 23 columns]

```
In [41]: # Drop duplicates and keep the first occurrence
data = data.drop_duplicates(subset=["track_name", "track_artist"])
print("\nDataFrame after removing duplicates:")
print(data.head())
```

DataFrame after removing duplicates:

	track_id	track_name	track_artist	track_popularity	track_album_id	track_album_name	track_album_release_date	mode	speechiness	acousticness	instrumentalness	liveness	valence	tempo	duration_ms
0	6f807x0ima9a1j3VPbc7VN	I Don't Care (with Justin Bieber) - Loud Luxury...	Ed Sheeran	66	2oCs0DGTsR098Gh5ZSl2Cx	I Don't Care (with Justin Bieber) [Loud Luxury...]	14-06-2019	1	0.0583	0.1020	0.000000	0.0653	0.518		
1	0r7CVbZTWZgbTCYdfa2P31	Memories - Dillon Francis R emix	Maroon 5	67	63rPS0264uRjW1X5E6cWv6	Memories (Dillon Francis Remix)	13-12-2019	1	0.0373	0.0724	0.004210	0.3570	0.693		
2	1z1Hg7Vb0AhHDiEmnDE791	All the Time - Don Diablo R emix	Zara Larsson	70	1HoSmj2eLcsrR0vE9gThr4	All the Time (Don Diablo Remix)	05-07-2019	0	0.0742	0.0794	0.000023	0.1100	0.613		
3	75FpbthrwQmzHlBJLuGdC7	Call You Mine - Keanu Silva R emix	The Chainsmokers	60	1nqYsOef1yKKuGOVchbsk6	Call You Mine - The Remixes	19-07-2019	1	0.1020	0.0287	0.000009	0.2040	0.277		
4	1e8PAfcKUYoKkxPhrHqw4x	Someone You Loved - Future Humans R emix	Lewis Capaldi	69	7m7vv9wlQ4i0LFuJiE2zsQ	Someone You Loved (Future Humans Remix)	05-03-2019	2	0.0359	0.0803	0.000000	0.0833	0.725		

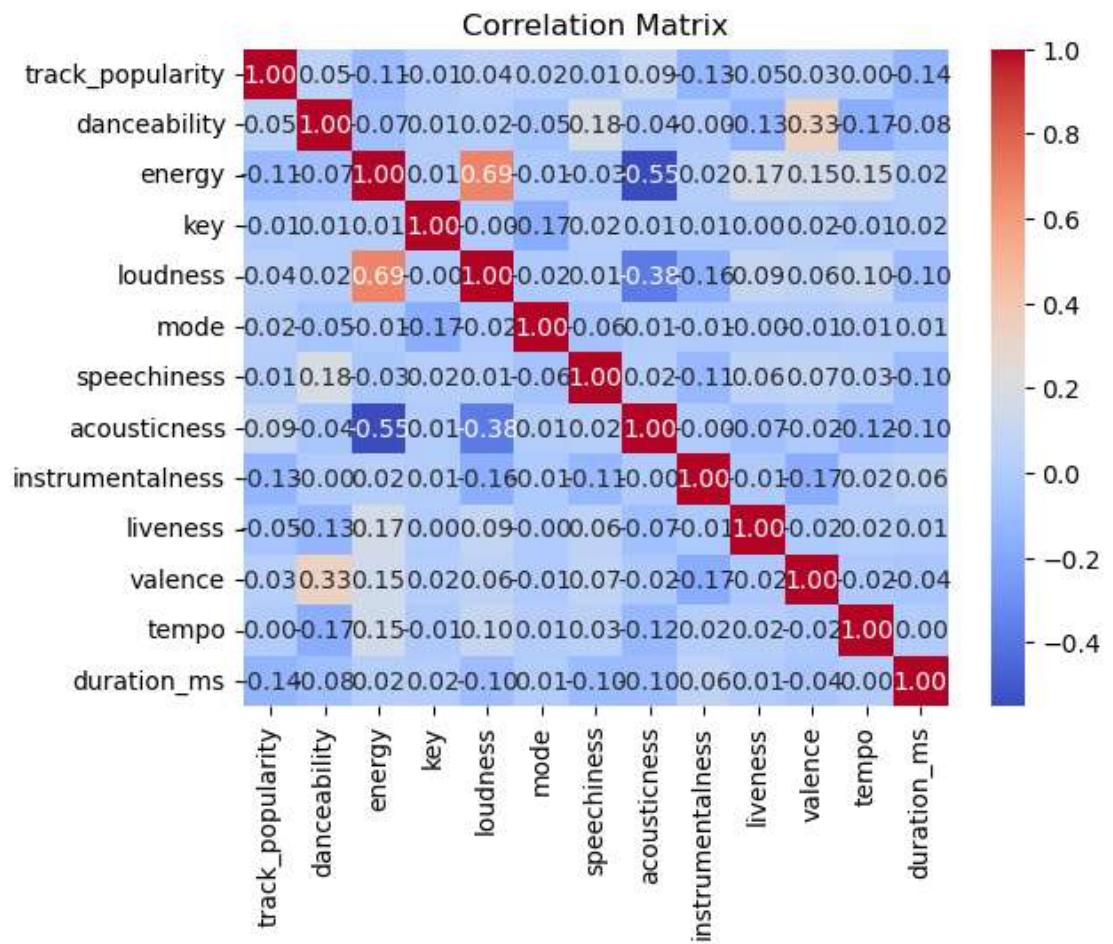
```
0 122.036      194754
1 99.972       162600
2 124.008      176616
3 121.956      169093
4 123.976      189052
```

[5 rows x 23 columns]

EXPLORATORY DATA ANALYSIS (EDA)

In [42]:

```
#correlation matrix
correlation_matrix = data.corr()
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt=".2f")
plt.title("Correlation Matrix")
plt.show()
```

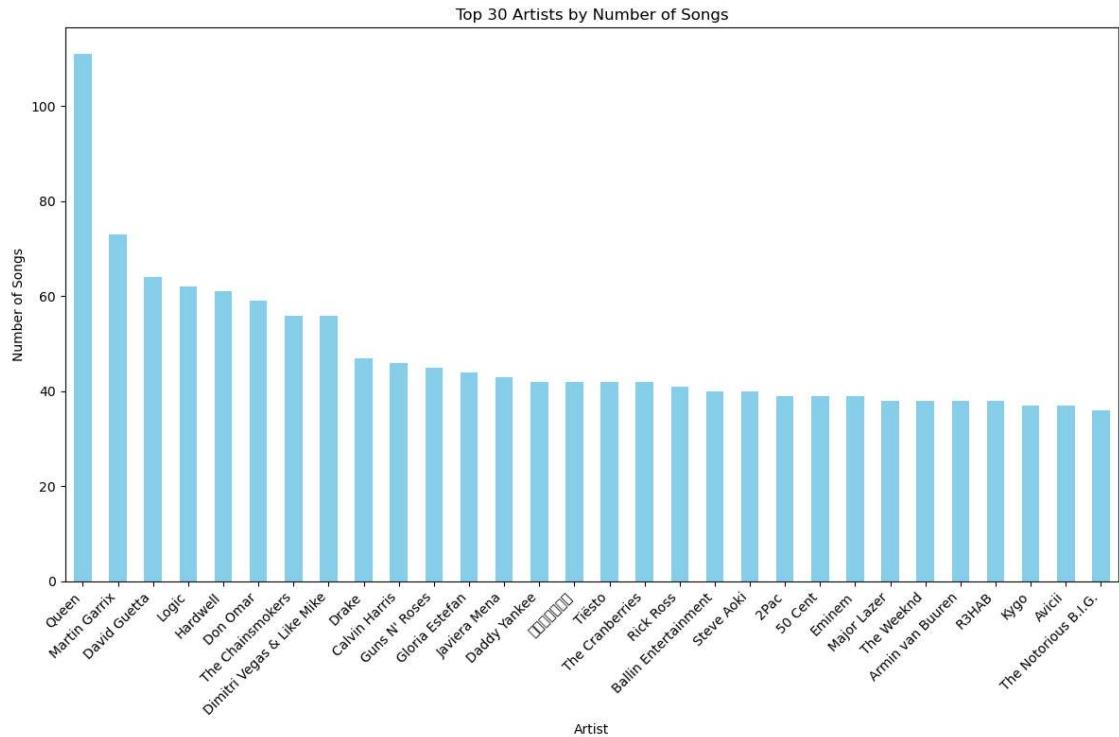


```
In [43]: ┌─ artist_counts = data['track_artist'].value_counts()  
└─ print(artist_counts.head(30))
```

Queen	111
Martin Garrix	73
David Guetta	64
Logic	62
Hardwell	61
Don Omar	59
The Chainsmokers	56
Dimitri Vegas & Like Mike	56
Drake	47
Calvin Harris	46
Guns N' Roses	45
Gloria Estefan	44
Javiera Mena	43
Daddy Yankee	42
オメガトライブ	42
Tiësto	42
The Cranberries	42
Rick Ross	41
Ballin Entertainment	40
Steve Aoki	40
2Pac	39
50 Cent	39
Eminem	39
Major Lazer	38
The Weeknd	38
Armin van Buuren	38
R3HAB	38
Kygo	37
Avicii	37
The Notorious B.I.G.	36
Name: track_artist, dtype: int64	

```
In [44]: ┌─ top_30_artists = artist_counts.head(30)
#bar graph
plt.figure(figsize=(12, 8))
top_30_artists.plot(kind='bar', color='skyblue')
plt.title('Top 30 Artists by Number of Songs')
plt.xlabel('Artist')
plt.ylabel('Number of Songs')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```

```
C:\Users\HP\AppData\Local\Temp\ipykernel_8836\2882317381.py:9: UserWarning: Glyph 12458 (\N{KATAKANA LETTER O}) missing from current font.
    plt.tight_layout()
C:\Users\HP\AppData\Local\Temp\ipykernel_8836\2882317381.py:9: UserWarning: Glyph 12513 (\N{KATAKANA LETTER ME}) missing from current font.
    plt.tight_layout()
C:\Users\HP\AppData\Local\Temp\ipykernel_8836\2882317381.py:9: UserWarning: Glyph 12460 (\N{KATAKANA LETTER GA}) missing from current font.
    plt.tight_layout()
C:\Users\HP\AppData\Local\Temp\ipykernel_8836\2882317381.py:9: UserWarning: Glyph 12488 (\N{KATAKANA LETTER TO}) missing from current font.
    plt.tight_layout()
C:\Users\HP\AppData\Local\Temp\ipykernel_8836\2882317381.py:9: UserWarning: Glyph 12521 (\N{KATAKANA LETTER RA}) missing from current font.
    plt.tight_layout()
C:\Users\HP\AppData\Local\Temp\ipykernel_8836\2882317381.py:9: UserWarning: Glyph 12452 (\N{KATAKANA LETTER I}) missing from current font.
    plt.tight_layout()
C:\Users\HP\AppData\Local\Temp\ipykernel_8836\2882317381.py:9: UserWarning: Glyph 12502 (\N{KATAKANA LETTER BU}) missing from current font.
    plt.tight_layout()
C:\Users\HP\anaconda3\lib\site-packages\IPython\core\pylabtools.py:151: UserWarning: Glyph 12458 (\N{KATAKANA LETTER O}) missing from current font.
    fig.canvas.print_figure(bytes_io, **kw)
C:\Users\HP\anaconda3\lib\site-packages\IPython\core\pylabtools.py:151: UserWarning: Glyph 12513 (\N{KATAKANA LETTER ME}) missing from current font.
    fig.canvas.print_figure(bytes_io, **kw)
C:\Users\HP\anaconda3\lib\site-packages\IPython\core\pylabtools.py:151: UserWarning: Glyph 12460 (\N{KATAKANA LETTER GA}) missing from current font.
    fig.canvas.print_figure(bytes_io, **kw)
C:\Users\HP\anaconda3\lib\site-packages\IPython\core\pylabtools.py:151: UserWarning: Glyph 12488 (\N{KATAKANA LETTER TO}) missing from current font.
    fig.canvas.print_figure(bytes_io, **kw)
C:\Users\HP\anaconda3\lib\site-packages\IPython\core\pylabtools.py:151: UserWarning: Glyph 12521 (\N{KATAKANA LETTER RA}) missing from current font.
    fig.canvas.print_figure(bytes_io, **kw)
C:\Users\HP\anaconda3\lib\site-packages\IPython\core\pylabtools.py:151: UserWarning: Glyph 12452 (\N{KATAKANA LETTER I}) missing from current font.
    fig.canvas.print_figure(bytes_io, **kw)
C:\Users\HP\anaconda3\lib\site-packages\IPython\core\pylabtools.py:151: UserWarning: Glyph 12502 (\N{KATAKANA LETTER BU}) missing from current font.
    fig.canvas.print_figure(bytes_io, **kw)
```



In [45]: ➔ data['playlist_genre'].value_counts()

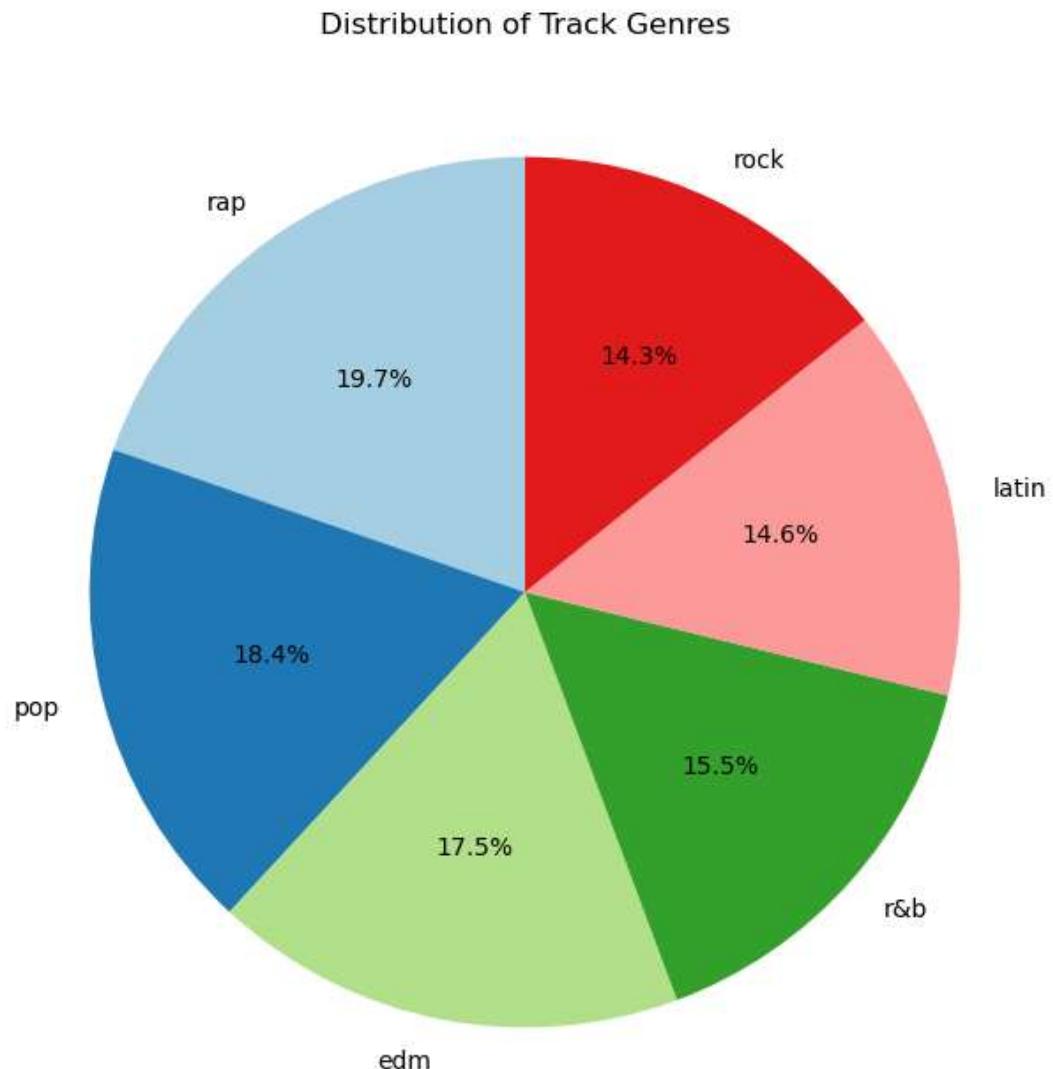
Out[45]:

rap	5165
pop	4836
edm	4603
r&b	4067
latin	3819
rock	3739

Name: playlist_genre, dtype: int64

```
In [46]: # Count the number of occurrences for each track genre
genre_counts = data['playlist_genre'].value_counts()

# Plot the pie chart
plt.figure(figsize=(8, 8))
plt.pie(genre_counts, labels=genre_counts.index, autopct='%1.1f%%', startangle=90)
plt.title('Distribution of Track Genres')
plt.show()
```

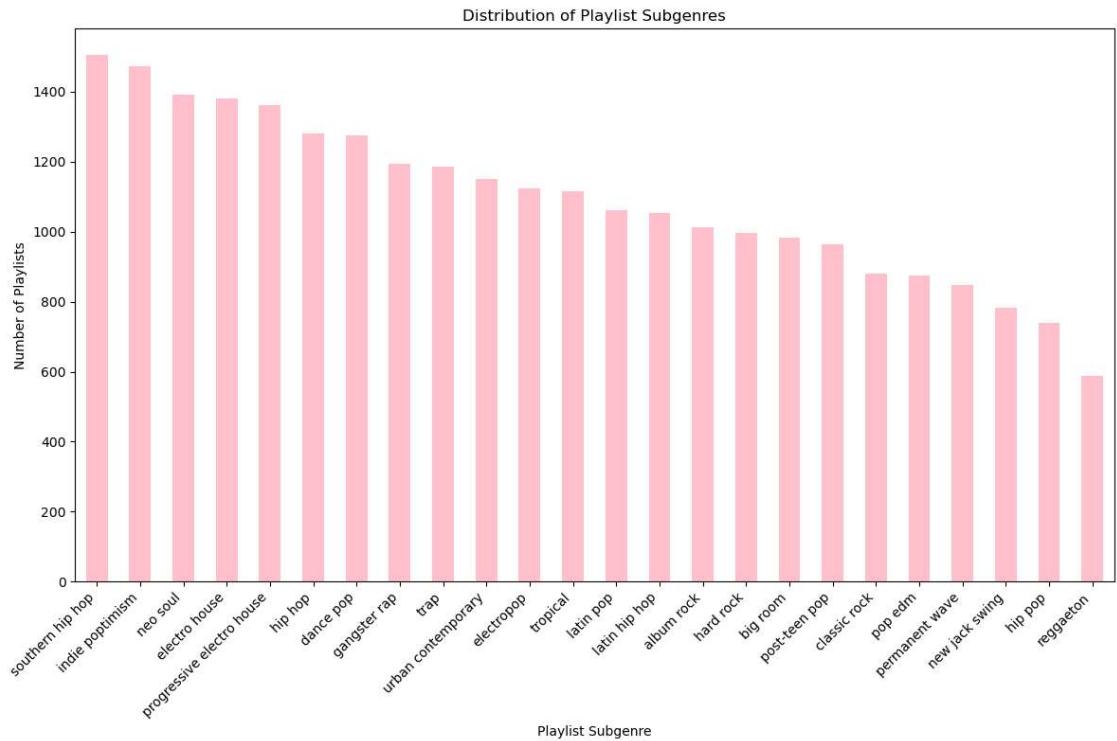


```
In [47]: ┌─ data['playlist_subgenre'].value_counts()
```

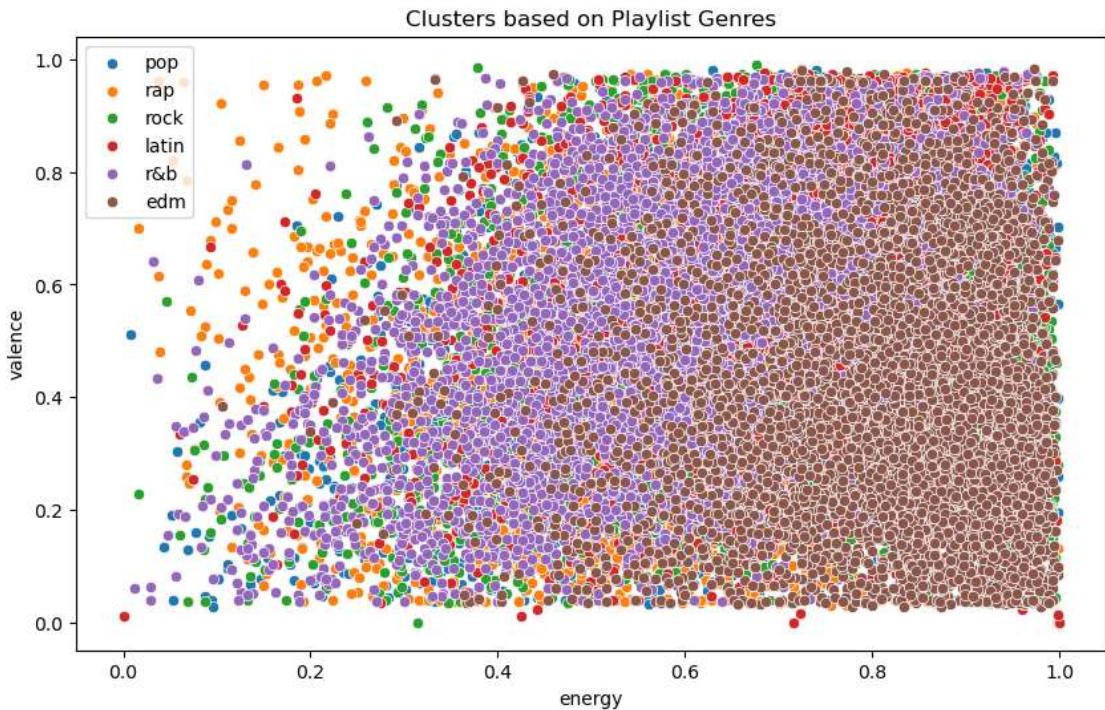
```
Out[47]: southern hip hop      1505
          indie poptimism       1474
          neo soul              1393
          electro house         1382
          progressive electro house 1363
          hip hop                1281
          dance pop              1276
          gangster rap           1194
          trap                   1185
          urban contemporary     1151
          electropop             1123
          tropical               1117
          latin pop              1061
          latin hip hop          1053
          album rock              1013
          hard rock               997
          big room                982
          post-teen pop           963
          classic rock            880
          pop edm                 876
          permanent wave          849
          new jack swing           784
          hip pop                 739
          reggaeton               588
Name: playlist_subgenre, dtype: int64
```

```
In [48]: # Count the number of occurrences for each playlist subgenre
subgenre_counts = data['playlist_subgenre'].value_counts()

# Plot the bar graph
plt.figure(figsize=(12, 8))
subgenre_counts.plot(kind='bar', color='pink')
plt.title('Distribution of Playlist Subgenres')
plt.xlabel('Playlist Subgenre')
plt.ylabel('Number of Playlists')
plt.xticks(rotation=45, ha='right')
plt.tight_layout()
plt.show()
```



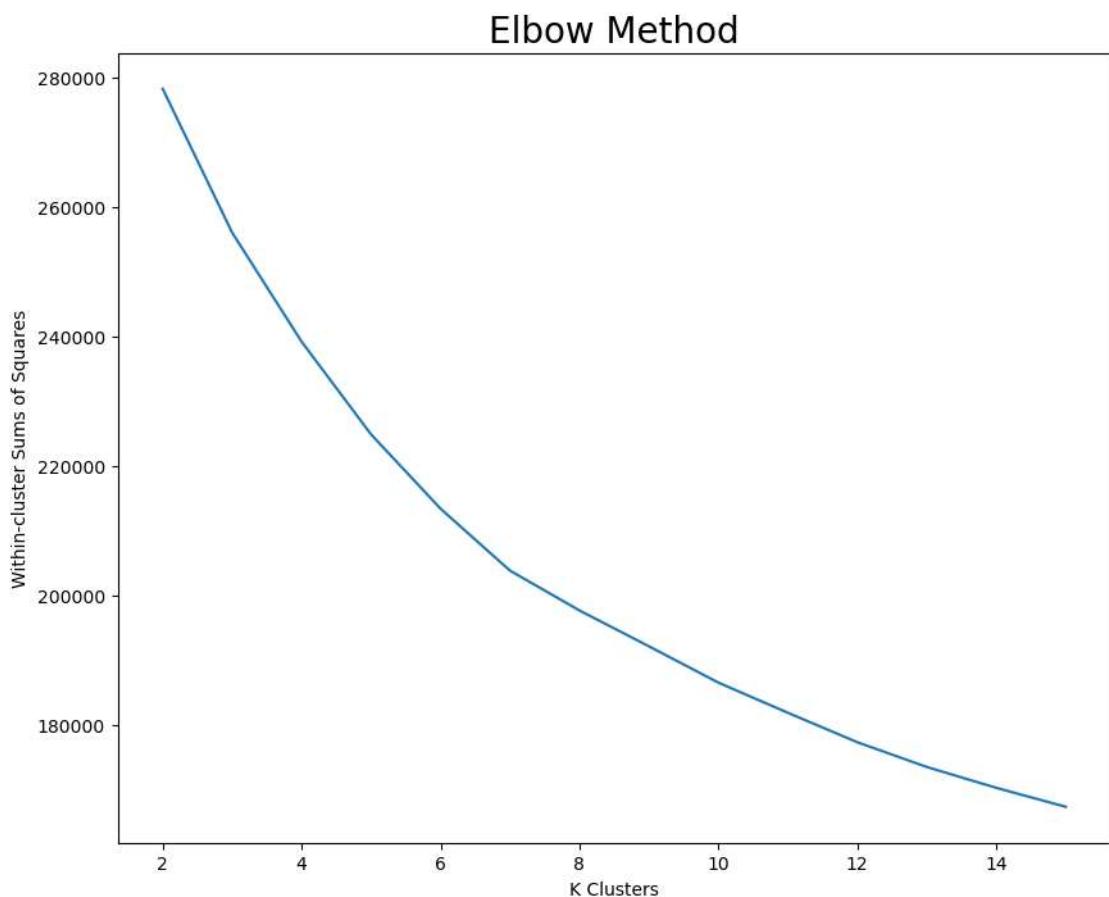
```
In [49]: # Clustering based on playlist genres
plt.figure(figsize=(10, 6))
playlist_genres = data['playlist_genre'].unique()
for genre in playlist_genres:
    genre_data = data[data['playlist_genre'] == genre]
    sns.scatterplot(x='energy', y='valence', data=genre_data, label=genre)
plt.title('Clusters based on Playlist Genres')
plt.legend()
plt.show()
```



```
In [ ]: 
```

```
In [50]: from sklearn.cluster import KMeans
from sklearn.decomposition import PCA
from sklearn.metrics import silhouette_score
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import StandardScaler
cols = data.iloc[:, 9: ].select_dtypes(['uint8', 'int64', 'float64']).columns
```

```
In [51]: ┆ wcss = []
for i in range(2,16):
    km = KMeans(n_clusters=i, init='k-means++', max_iter=500, n_init=10, r
    d = StandardScaler().fit_transform(data[cols])
    km.fit(d)
    wcss.append(km.inertia_)
fig = plt.figure(figsize=(10,8))
ax = sns.lineplot(x=range(2,16), y=wcss)
ax.set_title('Elbow Method')
ax.title.set_size(20)
plt.xlabel('K Clusters')
plt.ylabel('Within-cluster Sums of Squares')
plt.show()
```



```
In [52]: ┆ new_data=data[cols].dropna()
scaler = StandardScaler()
scaled_data = scaler.fit_transform(new_data)
k=4
kmeans = KMeans(n_clusters=k, random_state=1)
cluster_labels = kmeans.fit_predict(scaled_data)
silhouette_avg = silhouette_score(scaled_data, cluster_labels)
print("Silhouette Score:", silhouette_avg)
```

Silhouette Score: 0.10559443877113443

In [53]: cols

```
Out[53]: Index(['danceability', 'energy', 'key', 'loudness', 'mode', 'speechiness',
   'acousticness', 'instrumentalness', 'liveness', 'valence', 'tempo',
   'duration_ms'],
  dtype='object')
```

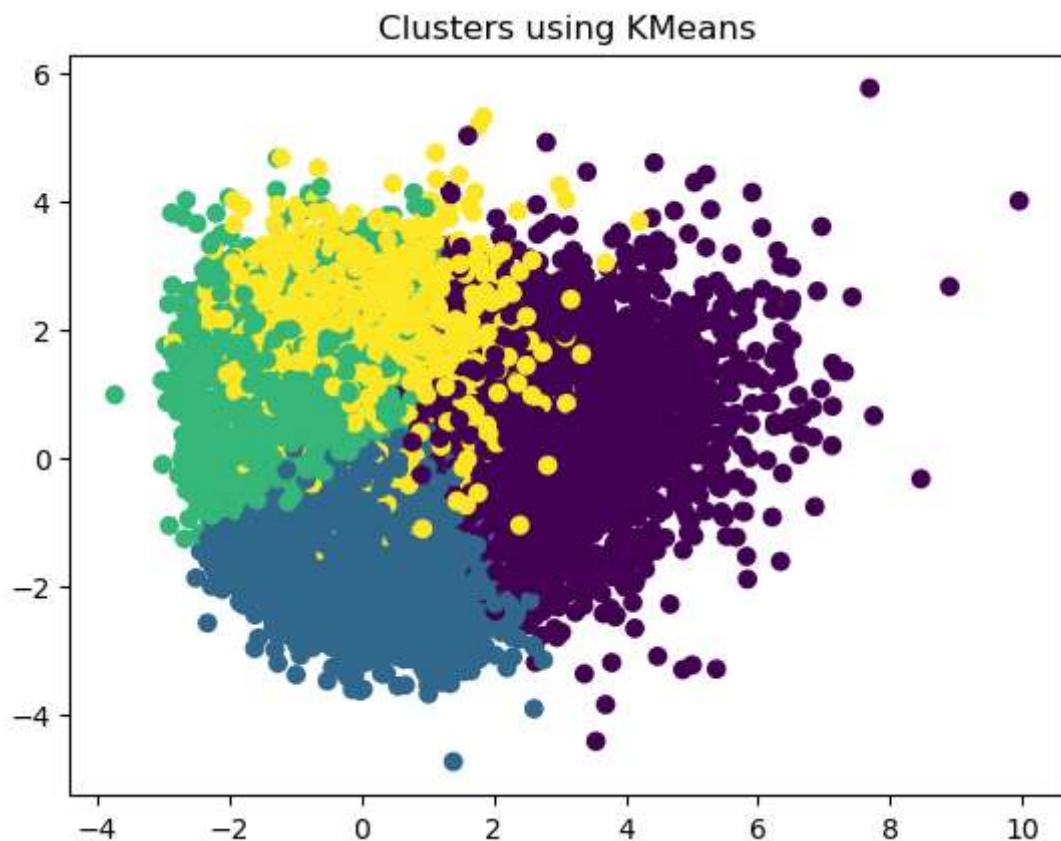
```
In [54]: # Build Model (Example using KMeans)
features_for_clustering = ['danceability', 'energy', 'key', 'loudness', 'mode',
                           'speechiness', 'acousticness', 'instrumentalness', 'liveness',
                           'valence', 'tempo', 'duration_ms']

X = data[features_for_clustering]

# Standardize the features
scaler = StandardScaler()
X_standardized = scaler.fit_transform(X)

# Use KMeans for clustering
kmeans = KMeans(n_clusters=4, random_state=42)
data['cluster'] = kmeans.fit_predict(X_standardized)

# Plot clusters in 2D using PCA
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X_standardized)
plt.scatter(X_pca[:, 0], X_pca[:, 1], c=data['cluster'], cmap='viridis')
plt.title('Clusters using KMeans')
plt.show()
```



```
In [55]: ┌─ top_tracks_cluster_0 = data[data['cluster'] == 0].nlargest(15, 'track_popularity')
    ┌─ print("Top 15 tracks from Cluster 0:")
    ┌─ print(top_tracks_cluster_0[['track_name', 'track_artist', 'track_popularity']])
```

Top 15 tracks from Cluster 0:

	track_name	track_artist	track_popularity
711	Memories	Maroon 5	98
689	everything i wanted	Billie Eilish	97
3155	Falling	Trevor Daniel	97
344	bad guy	Billie Eilish	95
718	Someone You Loved	Lewis Capaldi	94
1303	HIGHEST IN THE ROOM	Travis Scott	94
739	Lose You To Love Me	Selena Gomez	93
922	Dance Monkey	Tones and I	92
5533	No Idea	Don Toliver	92
742	Before You Go	Lewis Capaldi	90
1319	Trampoline (with ZAYN)	SHAED	90
1782	7 rings	Ariana Grande	90
5517	Suicidal	YNW Melly	90
5929	Ballin' (with Roddy Ricch)	Mustard	90
22142	lovely (with Khalid)	Billie Eilish	89

```
In [56]: ┌─ top_tracks_cluster_1 = data[data['cluster'] == 1].nlargest(15, 'track_popularity')
    ┌─ print("Top 15 tracks from Cluster 1:")
    ┌─ print(top_tracks_cluster_1[['track_name', 'track_artist', 'track_popularity']])
```

Top 15 tracks from Cluster 1:

	track_name	track_artist	track_popularity
687	Tusa	KAROL G	98
5508	The Box	Roddy Ricch	98
1302	Don't Start Now	Dua Lipa	97
139	RITMO (Bad Boys For Life)	The Black Eyed Peas	96
694	Yummy	Justin Bieber	95
646	Ride It	Regard	94
1306	hot girl bummer	blackbear	94
1463	My Oh My (feat. DaBaby)	Camila Cabello	94
702	China	Anuel AA	93
709	Señorita	Shawn Mendes	93
5507	Life Is Good (feat. Drake)	Future	93
5514	BOP	DaBaby	93
11115	Vete	Bad Bunny	93
1339	Fantasias	Rauw Alejandro	92
179	Lose Control	MEDUZA	91

```
In [57]: ┌─ top_tracks_cluster_2 = data[data['cluster'] == 2].nlargest(15, 'track_popularity')
  ┌─ print("Top 15 tracks from Cluster 2:")
  ┌─ print(top_tracks_cluster_2[['track_name', 'track_artist', 'track_popularity']])
```

Top 15 tracks from Cluster 2:

	track_name	track_artist
\		
716	Blinding Lights	The Weeknd
730	Heartless	The Weeknd
720	Bandit (with YoungBoy Never Broke Again)	Juice WRLD
1314	How Do You Sleep?	Sam Smith
1333	Callaita	Bad Bunny
1611	Watermelon Sugar	Harry Styles
18360	Goodbyes (feat. Young Thug)	Post Malone
25788	All I Want for Christmas Is You	Mariah Carey
1327	Good as Hell (feat. Ariana Grande) - Remix	Lizzo
5540	HIGHEST IN THE ROOM (feat. ROSALÍA & Lil Baby)...	Travis Scott
1341	Loco Contigo (feat. J. Balvin & Tyga)	DJ Snake
1344	Si Te Vas	Sech
8980	rockstar (feat. 21 Savage)	Post Malone
10785	Robbery	Juice WRLD
18868	Quizas	Rich Music LTD
	track_popularity	
716	98	
730	93	
720	92	
1314	91	
1333	90	
1611	90	
18360	90	
25788	90	
1327	89	
5540	89	
1341	88	
1344	88	
8980	88	
10785	88	
18868	88	

```
In [58]: ┌─ top_tracks_cluster_3 = data[data['cluster'] == 3].nlargest(15, 'track_popularity')
  ┌─ print("Top 15 tracks from Cluster 3:")
  ┌─ print(top_tracks_cluster_3[['track_name', 'track_artist', 'track_popularity']])
```

Top 15 tracks from Cluster 3:

	track_name	track_artist
18562	Baila Conmigo (feat. Kelly Ruiz)	Dayvi
22997	In My Room	Frank Ocean
634	Baianá	Bakermat
1110	Feels Like We Only Go Backwards	Tame Impala
1223	Instant Crush	Daft Punk
5054	On Melancholy Hill	Gorillaz
14448	Heroes - 2017 Remaster	David Bowie
13622	1979 - Remastered 2012	The Smashing Pumpkins
23031	A Palé	ROSALÍA
11762	Born To Be Wild	Steppenwolf
13672	Reptilia	The Strokes
14051	Blue Monday - 2016 Remaster	New Order
14575	Chamber Of Reflection	Mac DeMarco
4147	Intro	The xx
10846	Look at Me Now	Brennan Savage

	track_popularity
18562	83
22997	79
634	76
1110	75
1223	75
5054	75
14448	75
13622	74
23031	74
11762	73
13672	73
14051	73
14575	73
4147	72
10846	72

```
In [59]: ┌─ data['track_popularity'].describe()
```

```
Out[59]: count    26229.000000
mean      39.508178
std       23.265382
min       0.000000
25%      22.000000
50%      42.000000
75%      58.000000
max      98.000000
Name: track_popularity, dtype: float64
```

```
In [60]: ► data['cluster'].value_counts()
```

```
Out[60]: 1    10898  
2    8740  
0    4285  
3    2306  
Name: cluster, dtype: int64
```

```
In [ ]: ►
```

```
In [61]: ┌─▶ from sklearn.metrics.pairwise import cosine_similarity

data['new_index'] = range(1, len(data) + 1)
data.set_index('new_index', inplace=True)

# Create keywords column by combining track_name and track_artist
data['keywords'] = data['track_name'] + ' ' + data['track_artist']

# Extract genres and create a genre matrix
genres = data['playlist_genre'].str.split(',').explode().str.strip()
genre_matrix = pd.get_dummies(genres).groupby(level=0).sum()

# Combine keywords and genre_matrix
feature_matrix = pd.concat([data['keywords'], genre_matrix], axis=1)

# Create user profile based on favorite genres
favorite_genres = ['edm', 'pop']
user_profile = pd.DataFrame({'genre_matrix': np.isin(genre_matrix.columns,

# Calculate similarity scores
similarity_scores = cosine_similarity(feature_matrix.drop(columns='keywords'))

# Combine similarity scores with the original data
data_recommendations = pd.concat([data, pd.DataFrame({'similarity': similarity_scores}), axis=1]

# Get top N recommendations
N = 25
sorted_recommendations = data_recommendations.sort_values(by='similarity', ascending=False).head(N)
top_songs = sorted_recommendations[['track_name', 'track_artist', 'similarity']]

# Display top recommendations
print(top_songs)
```



		track_name	track_arti
st \			
1	I Don't Care (with Justin Bieber) - Loud Luxur...		Ed Sheer
an			
23095		The End	SpinR
ox			
23079		Haunted	Dave M
ak			
23080		Mad Echoes	Swede Drea
ms			
23081		Raise	VITI
ZE			
23082		Gringos	Ke
vu			
23083		Don't Stop	H2
HB			
23084		Omega	Ka
Zo			
23085		Never	Widemo
de			
23086		Mantra	NAE
MS			
23087		Outta Control	Winning Te
am			
23088		Dominica	NAE
MS			
23089		No Fear	Latw
er			
23090		Shotgun	RudeLi
es			
23091		Watch Your Back	FineRefin
ed			
23092		Don't Wanna Be Cold	Krimso
nn			
23093		Take Me	Casti
on			
23078		Back To Life	KRI
SM			
23077	Symphony - Dr Phunk Remix	Sick Individua	
ls			
23076		LUNA	Sick Individua
ls			
23067		Up In Smoke	KA
ZE			
23061		Beast	PURA
RI			
23062	Losing My Religion - Ton Don Remix	Jones & Bro	
ck			
23063		Black Butterflies	Kill The Bu
zz			
23064		Game Time	Swanky Tun
es			

	playlist_genre	track_popularity	similarity
1	pop	66.0	0.707107
23095	edm	18.0	0.707107
23079	edm	20.0	0.707107

23080	edm	20.0	0.707107
23081	edm	31.0	0.707107
23082	edm	21.0	0.707107
23083	edm	17.0	0.707107
23084	edm	15.0	0.707107
23085	edm	36.0	0.707107
23086	edm	19.0	0.707107
23087	edm	22.0	0.707107
23088	edm	18.0	0.707107
23089	edm	17.0	0.707107
23090	edm	19.0	0.707107
23091	edm	16.0	0.707107
23092	edm	17.0	0.707107
23093	edm	17.0	0.707107
23078	edm	32.0	0.707107
23077	edm	49.0	0.707107
23076	edm	55.0	0.707107
23067	edm	33.0	0.707107
23061	edm	38.0	0.707107
23062	edm	32.0	0.707107
23063	edm	38.0	0.707107
23064	edm	33.0	0.707107

In []:

