

# SPOTIFY MUSIC

## Introduction

This report presents a detailed analysis of Spotify music data, aiming to uncover insights into various aspects of the music industry. The primary focus is on understanding track attributes, chart performance, top artists, and musical characteristics as represented in the dataset.

The key objectives of this analysis include:

1. Exploring the distribution of streams for tracks to understand popularity trends.
2. Analyzing the release trends over the last decade to identify changing music preferences.
3. Investigating correlations between different track attributes to uncover musical patterns.
4. Examining chart performance across platforms (Spotify, Apple, Deezer, Shazam) to gauge industry presence.
5. Identifying top artists based on track counts to discern audience preferences.
6. Exploring the distribution of tracks by musical key, mode, BPM, and release month for genre insights.

Through this analysis, we aim to provide valuable insights for stakeholders in the music industry, including artists, record labels, streaming platforms, and music enthusiasts. The findings and interpretations derived from this analysis can inform strategic decisions, marketing campaigns, and content creation strategies within the dynamic music landscape.

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.express as px
from wordcloud import WordCloud

import pandas as pd

# Read the file with the detected encoding
df = pd.read_csv('Popular_Spotify_Songs.csv', encoding='ISO-8859-1')

# Save the file with UTF-8 encoding
df.to_csv('Popular_Spotify_Songs.csv', index=False, encoding='utf-8')

df = pd.read_csv("Popular_Spotify_Songs.csv")
df
```

	track_name	artist(s)_name
artist_count \		
0	Seven (feat. Latto) (Explicit Ver.)	Latto, Jung Kook
2		

1		LALA	Myke Towers
1			
2		vampire	Olivia Rodrigo
1			
3		Cruel Summer	Taylor Swift
1			
4		WHERE SHE GOES	Bad Bunny
1			
..		...	...
...			
948		My Mind & Me	Selena Gomez
1			
949		Bigger Than The Whole Sky	Taylor Swift
1			
950		A Veces (feat. Feid)	Feid, Paulo Londra
2			
951		En La De Ella	Feid, Sech, Jhayco
3			
952		Alone	Burna Boy
1			

	released_year	released_month	released_day	in_spotify_playlists
\				
0	2023	7	14	553
1	2023	3	23	1474
2	2023	6	30	1397
3	2019	8	23	7858
4	2023	5	18	3133
..	...	...	...	...
948	2022	11	3	953
949	2022	10	21	1180
950	2022	11	3	573
951	2022	10	20	1320
952	2022	11	4	782

	in_spotify_charts	streams	in_apple_playlists	...	bpm	key
mode \						
0	147	141381703	43	...	125	B
Major						

1	48	133716286	48	...	92	C#
Major						
2	113	140003974	94	...	138	F
Major						
3	100	800840817	116	...	170	A
Major						
4	50	303236322	84	...	144	A
Minor						
..	...	...	...	...	...	..
...						
948	0	91473363	61	...	144	A
Major						
949	0	121871870	4	...	166	F#
Major						
950	0	73513683	2	...	92	C#
Major						
951	0	133895612	29	...	97	C#
Major						
952	2	96007391	27	...	90	E
Minor						

	danceability_%	valence_%	energy_%	acousticness_%
instrumentalness_%	\			
0	80	89	83	31
0				
1	71	61	74	7
0				
2	51	32	53	17
0				
3	55	58	72	11
0				
4	65	23	80	14
63				
..	...	...	...	...
...				
948	60	24	39	57
0				
949	42	7	24	83
1				
950	80	81	67	4
0				
951	82	67	77	8
0				
952	61	32	67	15
0				

	liveness_%	speechiness_%
0	8	4
1	10	4

```

2          31          6
3          11         15
4          11          6
..         ...         ...
948         8          3
949        12          6
950         8          6
951        12          5
952        11          5

```

```
[953 rows x 24 columns]
```

```
df.head()
```

	track_name	artist(s)_name	artist_count
0	Seven (feat. Latto) (Explicit Ver.)	Latto, Jung Kook	2
1	LALA	Myke Towers	1
2	vampire	Olivia Rodrigo	1
3	Cruel Summer	Taylor Swift	1
4	WHERE SHE GOES	Bad Bunny	1

	released_year	released_month	released_day	in_spotify_playlists
0	2023	7	14	553
1	2023	3	23	1474
2	2023	6	30	1397
3	2019	8	23	7858
4	2023	5	18	3133

	in_spotify_charts	streams	in_apple_playlists	...	bpm	key
0	147	141381703	43	...	125	B
1	48	133716286	48	...	92	C#
2	113	140003974	94	...	138	F
3	100	800840817	116	...	170	A
4	50	303236322	84	...	144	A

Minor

	danceability_%	valence_%	energy_%	acousticness_%	instrumentalness_
0	80	89	83	31	
1	71	61	74	7	
2	51	32	53	17	
3	55	58	72	11	
4	65	23	80	14	

	liveness_%	speechiness_%
0	8	4
1	10	4
2	31	6
3	11	15
4	11	6

[5 rows x 24 columns]

df.tail()

	track_name	artist(s)_name	artist_count	\
948	My Mind & Me	Selena Gomez	1	
949	Bigger Than The Whole Sky	Taylor Swift	1	
950	A Veces (feat. Feid)	Feid, Paulo Londra	2	
951	En La De Ella	Feid, Sech, Jhayco	3	
952	Alone	Burna Boy	1	

	released_year	released_month	released_day	in_spotify_playlists
948	2022	11	3	953
949	2022	10	21	1180
950	2022	11	3	573
951	2022	10	20	1320
952	2022	11	4	782

	in_spotify_charts	streams	in_apple_playlists	...	bpm	key
948	0	91473363	61	...	144	A
949	0	121871870	4	...	166	F#

```

Major
950          0    73513683          2    ...    92    C#
Major
951          0   133895612         29    ...    97    C#
Major
952          2    96007391         27    ...    90    E
Minor

```

```

    danceability_%  valence_%  energy_%  acousticness_%
instrumentalness_% \
948             60          24         39          57
0
949             42           7         24          83
1
950             80          81         67           4
0
951             82          67         77           8
0
952             61          32         67          15
0

```

```

    liveness_%  speechiness_%
948           8              3
949          12              6
950           8              6
951          12              5
952          11              5

```

```
[5 rows x 24 columns]
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 953 entries, 0 to 952
```

```
Data columns (total 24 columns):
```

#	Column	Non-Null Count	Dtype
0	track_name	953 non-null	object
1	artist(s)_name	953 non-null	object
2	artist_count	953 non-null	int64
3	released_year	953 non-null	int64
4	released_month	953 non-null	int64
5	released_day	953 non-null	int64
6	in_spotify_playlists	953 non-null	int64
7	in_spotify_charts	953 non-null	int64
8	streams	953 non-null	object
9	in_apple_playlists	953 non-null	int64
10	in_apple_charts	953 non-null	int64
11	in_deezer_playlists	953 non-null	object
12	in_deezer_charts	953 non-null	int64

13	in_shazam_charts	903	non-null	object
14	bpm	953	non-null	int64
15	key	858	non-null	object
16	mode	953	non-null	object
17	danceability_	953	non-null	int64
18	valence_	953	non-null	int64
19	energy_	953	non-null	int64
20	acousticness_	953	non-null	int64
21	instrumentalness_	953	non-null	int64
22	liveness_	953	non-null	int64
23	speechiness_	953	non-null	int64

dtypes: int64(17), object(7)

memory usage: 178.8+ KB

df.describe()

	artist_count	released_year	released_month	released_day	\
count	953.000000	953.000000	953.000000	953.000000	
mean	1.556139	2018.238195	6.033578	13.930745	
std	0.893044	11.116218	3.566435	9.201949	
min	1.000000	1930.000000	1.000000	1.000000	
25%	1.000000	2020.000000	3.000000	6.000000	
50%	1.000000	2022.000000	6.000000	13.000000	
75%	2.000000	2022.000000	9.000000	22.000000	
max	8.000000	2023.000000	12.000000	31.000000	

	in_spotify_playlists	in_spotify_charts	in_apple_playlists	\
count	953.000000	953.000000	953.000000	
mean	5200.124869	12.009444	67.812172	
std	7897.608990	19.575992	86.441493	
min	31.000000	0.000000	0.000000	
25%	875.000000	0.000000	13.000000	
50%	2224.000000	3.000000	34.000000	
75%	5542.000000	16.000000	88.000000	
max	52898.000000	147.000000	672.000000	

	in_apple_charts	in_deezer_charts	bpm	danceability_	% \
count	953.000000	953.000000	953.000000	953.000000	
mean	51.908709	2.666317	122.540399	66.96957	
std	50.630241	6.035599	28.057802	14.63061	
min	0.000000	0.000000	65.000000	23.000000	
25%	7.000000	0.000000	100.000000	57.000000	
50%	38.000000	0.000000	121.000000	69.000000	

75%	87.000000	2.000000	140.000000	78.000000
max	275.000000	58.000000	206.000000	96.000000

	valence_%	energy_%	acousticness_%	instrumentalness_%
liveness_% \				
count	953.000000	953.000000	953.000000	953.000000
mean	51.431270	64.279119	27.057712	1.581322
std	23.480632	16.550526	25.996077	8.409800
min	4.000000	9.000000	0.000000	0.000000
25%	32.000000	53.000000	6.000000	0.000000
50%	51.000000	66.000000	18.000000	0.000000
75%	70.000000	77.000000	43.000000	0.000000
max	97.000000	97.000000	97.000000	91.000000

	speechiness_%
count	953.000000
mean	10.131165
std	9.912888
min	2.000000
25%	4.000000
50%	6.000000
75%	11.000000
max	64.000000

df.shape

(953, 24)

df.columns

```
Index(['track_name', 'artist(s)_name', 'artist_count',
      'released_year',
      'released_month', 'released_day', 'in_spotify_playlists',
      'in_spotify_charts', 'streams', 'in_apple_playlists',
      'in_apple_charts',
      'in_deezer_playlists', 'in_deezer_charts', 'in_shazam_charts',
      'bpm',
      'key', 'mode', 'danceability_%', 'valence_%', 'energy_%',
      'acousticness_%', 'instrumentalness_%', 'liveness_%',
      'speechiness_%'],
      dtype='object')
```



```

print(df.columns)

columns_to_drop = [
    'danceability_%', 'valence_%', 'energy_%', 'acousticness_%',
    'instrumentalness_%', 'liveness_%', 'speechiness_%'
]

df.drop(columns=columns_to_drop, inplace=True, errors='ignore')

print(df.head())

```

Index(['track\_name', 'artist(s)\_name', 'artist\_count', 'released\_year', 'released\_month', 'released\_day', 'in\_spotify\_playlists', 'in\_spotify\_charts', 'streams', 'in\_apple\_playlists', 'in\_apple\_charts', 'in\_deezer\_playlists', 'in\_deezer\_charts', 'in\_shazam\_charts', 'bpm', 'key', 'mode', 'danceability\_%', 'valence\_%', 'energy\_%', 'acousticness\_%', 'instrumentalness\_%', 'liveness\_%', 'speechiness\_%'], dtype='object')

	track_name	artist(s)_name	artist_count
0	Seven (feat. Latto) (Explicit Ver.)	Latto, Jung Kook	2
1	LALA	Myke Towers	1
2	vampire	Olivia Rodrigo	1
3	Cruel Summer	Taylor Swift	1
4	WHERE SHE GOES	Bad Bunny	1

	released_year	released_month	released_day	in_spotify_playlists
0	2023	7	14	553
1	2023	3	23	1474
2	2023	6	30	1397
3	2019	8	23	7858
4	2023	5	18	3133

	in_spotify_charts	streams	in_apple_playlists	in_apple_charts
0	147	141381703	43	263

1	48	133716286	48	126
2	113	140003974	94	207
3	100	800840817	116	207
4	50	303236322	84	133

	in_deezer_playlists	in_deezer_charts	in_shazam_charts	bpm	key
mode					
0	45	10	826	125	B
Major					
1	58	14	382	92	C#
Major					
2	91	14	949	138	F
Major					
3	125	12	548	170	A
Major					
4	87	15	425	144	A
Minor					

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 953 entries, 0 to 952
```

```
Data columns (total 17 columns):
```

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	track_name	953 non-null	object
1	artist(s)_name	953 non-null	object
2	artist_count	953 non-null	int64
3	released_year	953 non-null	int64
4	released_month	953 non-null	int64
5	released_day	953 non-null	int64
6	in_spotify_playlists	953 non-null	int64
7	in_spotify_charts	953 non-null	int64
8	streams	953 non-null	object
9	in_apple_playlists	953 non-null	int64
10	in_apple_charts	953 non-null	int64
11	in_deezer_playlists	953 non-null	object
12	in_deezer_charts	953 non-null	int64
13	in_shazam_charts	903 non-null	object
14	bpm	953 non-null	int64
15	key	858 non-null	object
16	mode	953 non-null	object

```
dtypes: int64(10), object(7)
```

```
memory usage: 126.7+ KB
```

```
# Summary statistics for all columns
summary_stats = df.describe(include='all')
print(summary_stats)
```

	track_name	artist(s)_name	artist_count	released_year
released_month \				
count	953	953	953.000000	953.000000
unique	943	645	NaN	NaN
top	Daylight	Taylor Swift	NaN	NaN
freq	2	34	NaN	NaN
mean	NaN	NaN	1.556139	2018.238195
std	NaN	NaN	0.893044	11.116218
min	NaN	NaN	1.000000	1930.000000
25%	NaN	NaN	1.000000	2020.000000
50%	NaN	NaN	1.000000	2022.000000
75%	NaN	NaN	2.000000	2022.000000
max	NaN	NaN	8.000000	2023.000000

	released_day	in_spotify_playlists	in_spotify_charts
streams \			
count	953.000000	953.000000	953.000000
unique	NaN	NaN	NaN
top	NaN	NaN	NaN
freq	NaN	NaN	NaN
mean	13.930745	5200.124869	12.009444
std	9.201949	7897.608990	19.575992
min	1.000000	31.000000	0.000000
25%	6.000000	875.000000	0.000000
50%	13.000000	2224.000000	3.000000
75%	22.000000	5542.000000	16.000000

NaN			
max	31.000000	52898.000000	147.000000
NaN			
	in_apple_playlists	in_apple_charts	in_deezer_playlists \
count	953.000000	953.000000	953
unique	NaN	NaN	348
top	NaN	NaN	0
freq	NaN	NaN	24
mean	67.812172	51.908709	NaN
std	86.441493	50.630241	NaN
min	0.000000	0.000000	NaN
25%	13.000000	7.000000	NaN
50%	34.000000	38.000000	NaN
75%	88.000000	87.000000	NaN
max	672.000000	275.000000	NaN

	in_deezer_charts	in_shazam_charts	bpm	key	mode
count	953.000000	903	953.000000	858	953
unique	NaN	198	NaN	11	2
top	NaN	0	NaN	C#	Major
freq	NaN	344	NaN	120	550
mean	2.666317	NaN	122.540399	NaN	NaN
std	6.035599	NaN	28.057802	NaN	NaN
min	0.000000	NaN	65.000000	NaN	NaN
25%	0.000000	NaN	100.000000	NaN	NaN
50%	0.000000	NaN	121.000000	NaN	NaN
75%	2.000000	NaN	140.000000	NaN	NaN
max	58.000000	NaN	206.000000	NaN	NaN

*# Check for missing values*

```
missing_values = df.isnull().sum()
print(missing_values)
```

*# Drop rows with missing values in critical columns*

```
df.dropna(subset=['key', 'mode'], inplace=True)
```

*# Fill missing values in 'in\_shazam\_charts' with 0*

```
df['in_shazam_charts'].fillna(0, inplace=True)
```

track_name	0
artist(s)_name	0
artist_count	0
released_year	0
released_month	0
released_day	0
in_spotify_playlists	0
in_spotify_charts	0
streams	0
in_apple_playlists	0

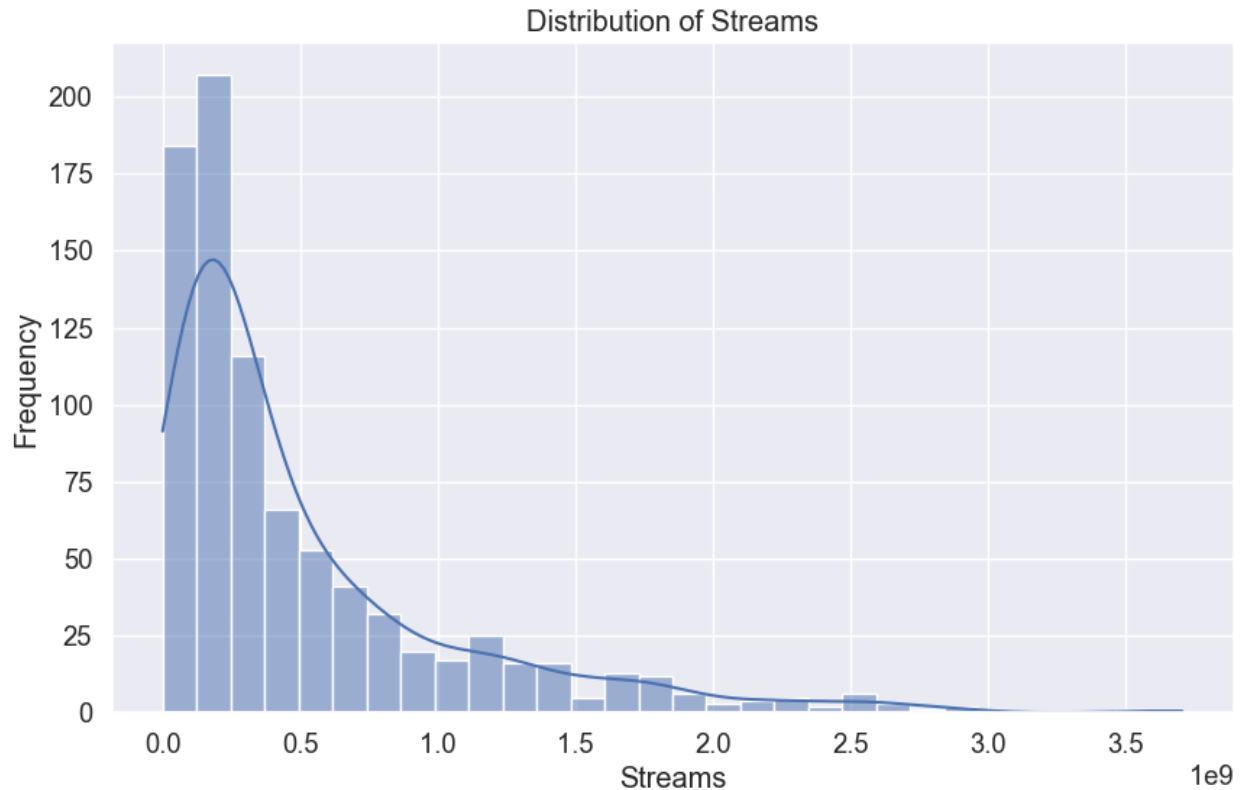
```
in_apple_charts      0
in_deezer_playlists  0
in_deezer_charts     0
in_shazam_charts     50
bpm                  0
key                  95
mode                 0
dtype: int64
```

```
# Convert 'streams' and 'in_deezer_playlists' columns to numeric
df['streams'] = pd.to_numeric(df['streams'], errors='coerce')
df['in_deezer_playlists'] = pd.to_numeric(df['in_deezer_playlists'],
errors='coerce')
```

```
# Check data types
print(df.dtypes)
```

```
track_name          object
artist(s)_name      object
artist_count        int64
released_year       int64
released_month      int64
released_day        int64
in_spotify_playlists int64
in_spotify_charts    int64
streams             float64
in_apple_playlists  int64
in_apple_charts     int64
in_deezer_playlists float64
in_deezer_charts    int64
in_shazam_charts    object
bpm                 int64
key                 object
mode                object
dtype: object
```

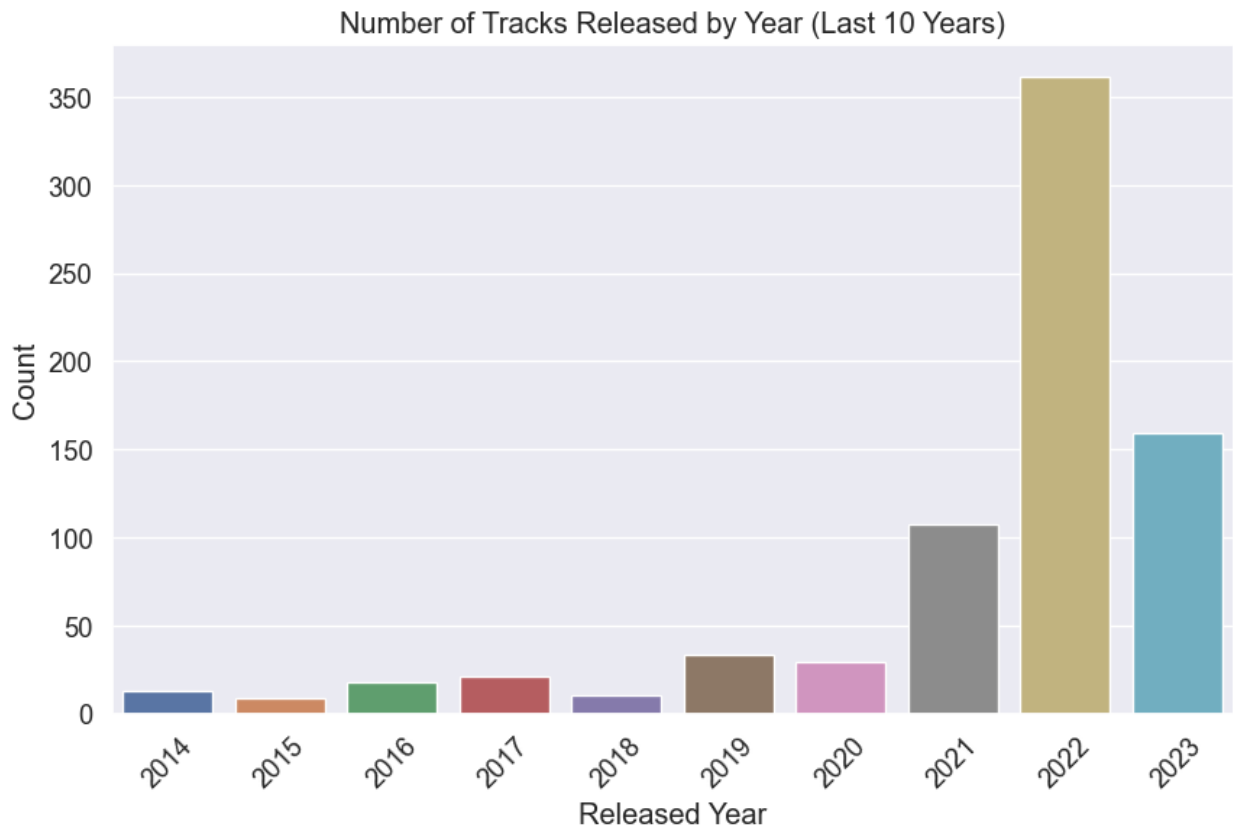
```
plt.figure(figsize=(10,6))
sns.histplot(df['streams'],bins=30,kde=True)
plt.title('Distribution of Streams')
plt.xlabel('Streams')
plt.ylabel('Frequency')
plt.show()
```



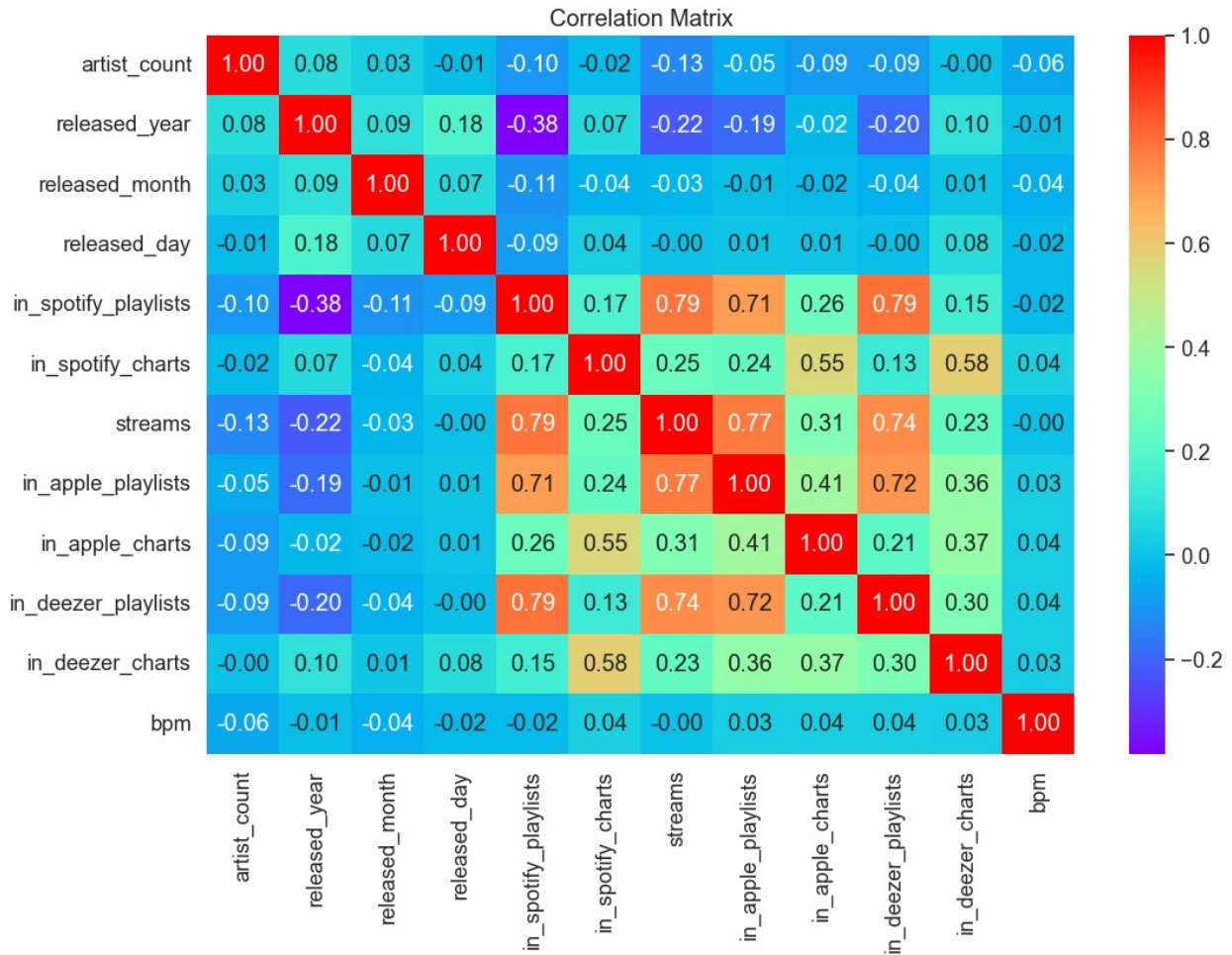
```
# Get the current year
current_year = pd.to_datetime('today').year

# Filter the DataFrame for the previous 10 years
df_last_10_years = df[df['released_year'] >= current_year - 10]

# Create a count plot for the previous 10 years
plt.figure(figsize=(10, 6))
sns.countplot(x='released_year', data=df_last_10_years)
plt.title('Number of Tracks Released by Year (Last 10 Years)')
plt.xlabel('Released Year')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.show()
```

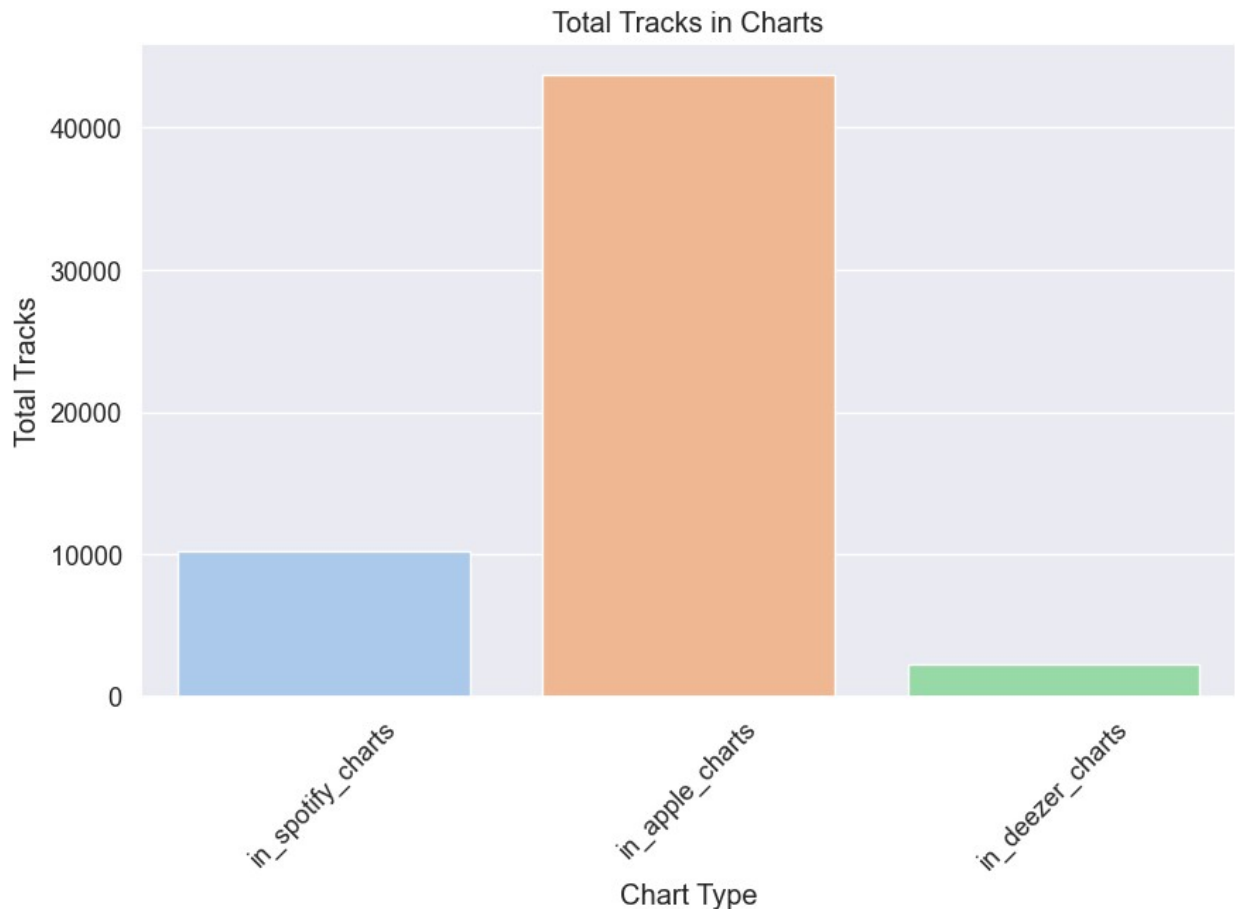


```
# Correlation heatmap
plt.figure(figsize=(12, 8))
sns.set(font_scale=1.2)
corr_matrix = df.corr()
sns.heatmap(corr_matrix, annot=True, cmap='rainbow', fmt='.2f')
plt.title('Correlation Matrix')
plt.show()
```



```
# Chart Performance Analysis
plt.figure(figsize=(10, 6))
chart_columns = ['in_spotify_charts', 'in_apple_charts',
                 'in_deezer_charts', 'in_shazam_charts']
chart_counts = df[chart_columns].sum()
sns.barplot(x=chart_counts.index, y=chart_counts.values,
            palette='pastel')
plt.title('Total Tracks in Charts')
plt.xlabel('Chart Type')
plt.ylabel('Total Tracks')
plt.xticks(rotation=45)
plt.show()
```





```
# Get the current year
current_year = pd.to_datetime('today').year

# Filter the DataFrame for the previous 10 years
df_last_10_years = df[df['released_year'] >= current_year - 10]

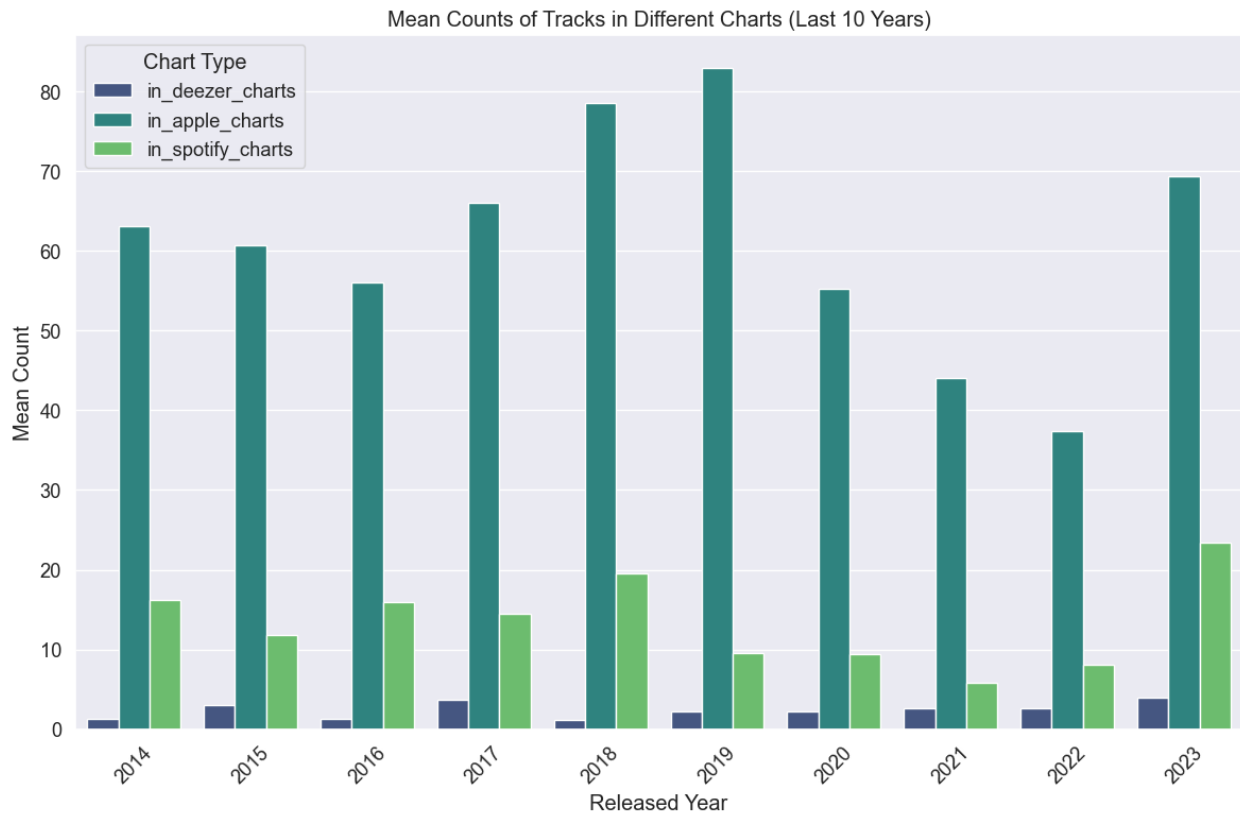
# Group the data by year and calculate the mean for each chart type
chart_comparison = df_last_10_years.groupby('released_year')
[['in_deezer_charts', 'in_apple_charts', 'in_spotify_charts']].mean()

# Reset the index to make 'released_year' a regular column
chart_comparison.reset_index(inplace=True)

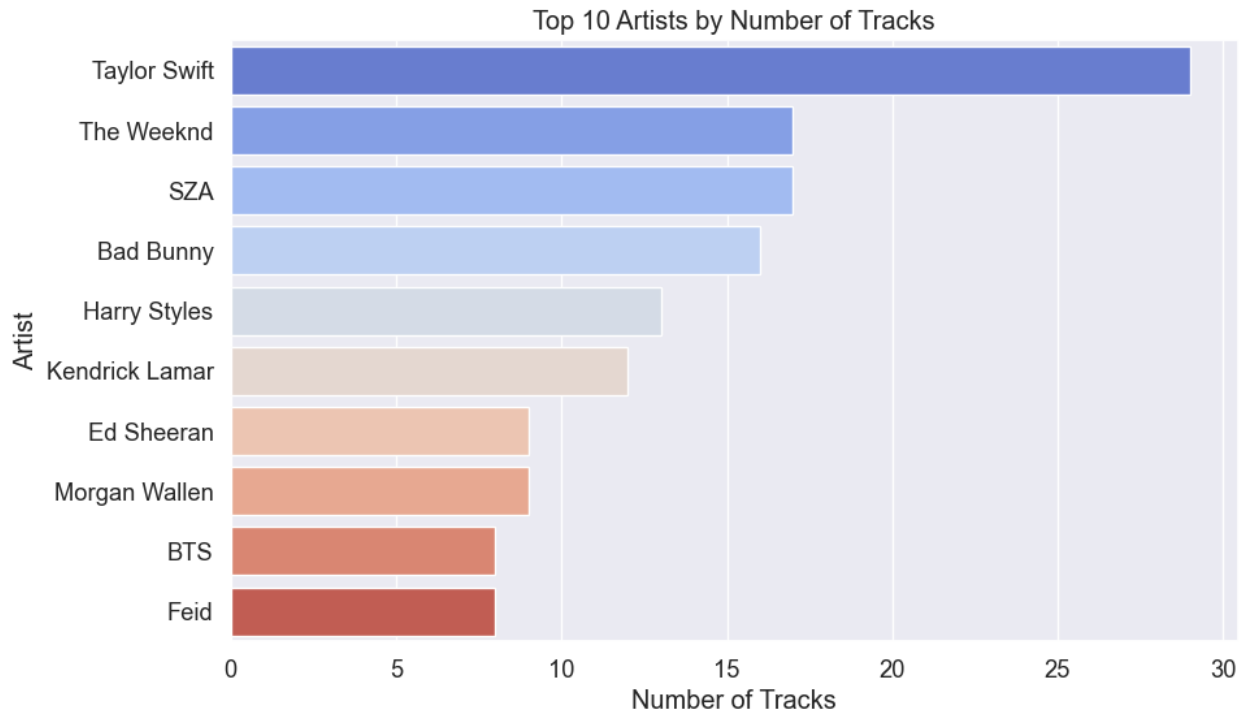
# Melt the DataFrame to convert chart types into a single column
chart_comparison_melted =
chart_comparison.melt(id_vars='released_year', var_name='Chart Type',
value_name='Mean Count')

plt.figure(figsize=(12, 8))
sns.barplot(x='released_year', y='Mean Count', hue='Chart Type',
data=chart_comparison_melted, palette='viridis')
plt.title('Mean Counts of Tracks in Different Charts (Last 10 Years)')
```

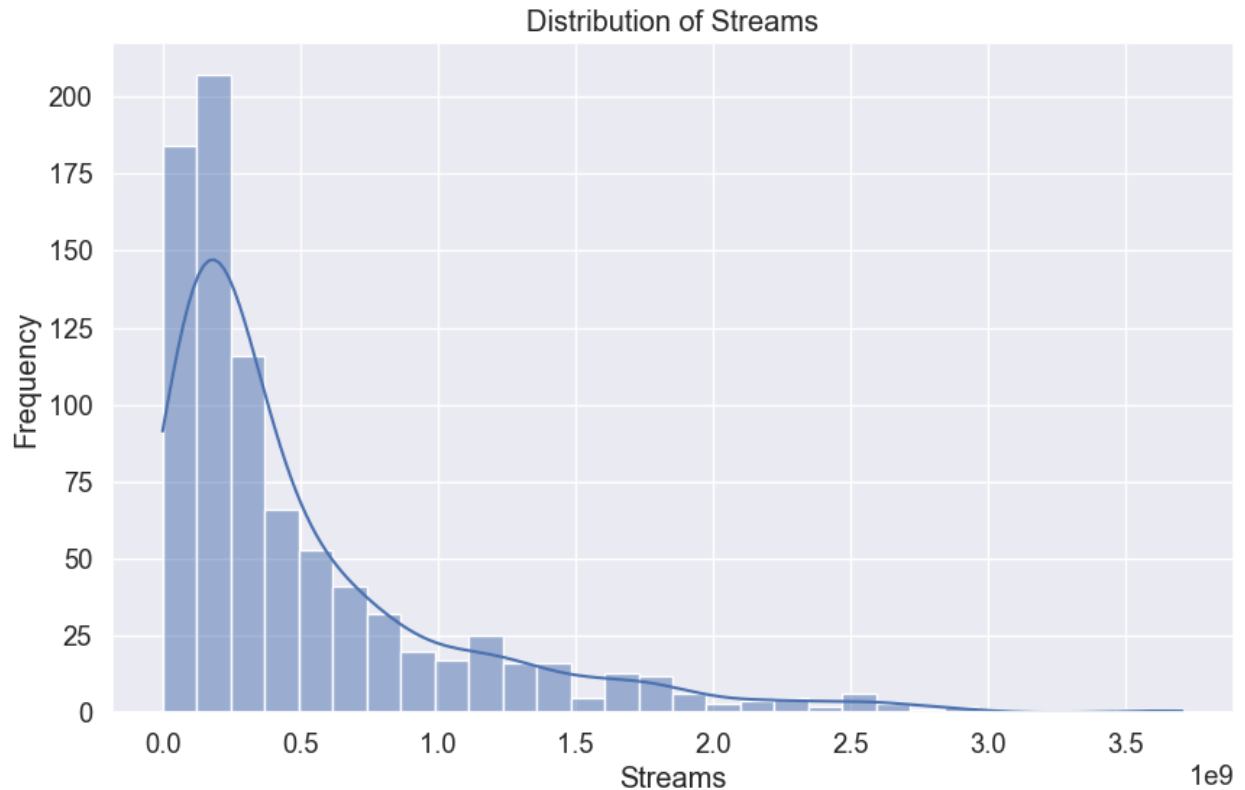
```
plt.xlabel('Released Year')
plt.ylabel('Mean Count')
plt.xticks(rotation=45)
plt.legend(title='Chart Type', loc='upper left')
plt.tight_layout()
plt.show()
```



```
# Top Artists Analysis
top_artists = df['artist(s)_name'].value_counts().head(10)
plt.figure(figsize=(10, 6))
sns.barplot(x=top_artists.values, y=top_artists.index,
palette='coolwarm')
plt.title('Top 10 Artists by Number of Tracks')
plt.xlabel('Number of Tracks')
plt.ylabel('Artist')
plt.show()
```



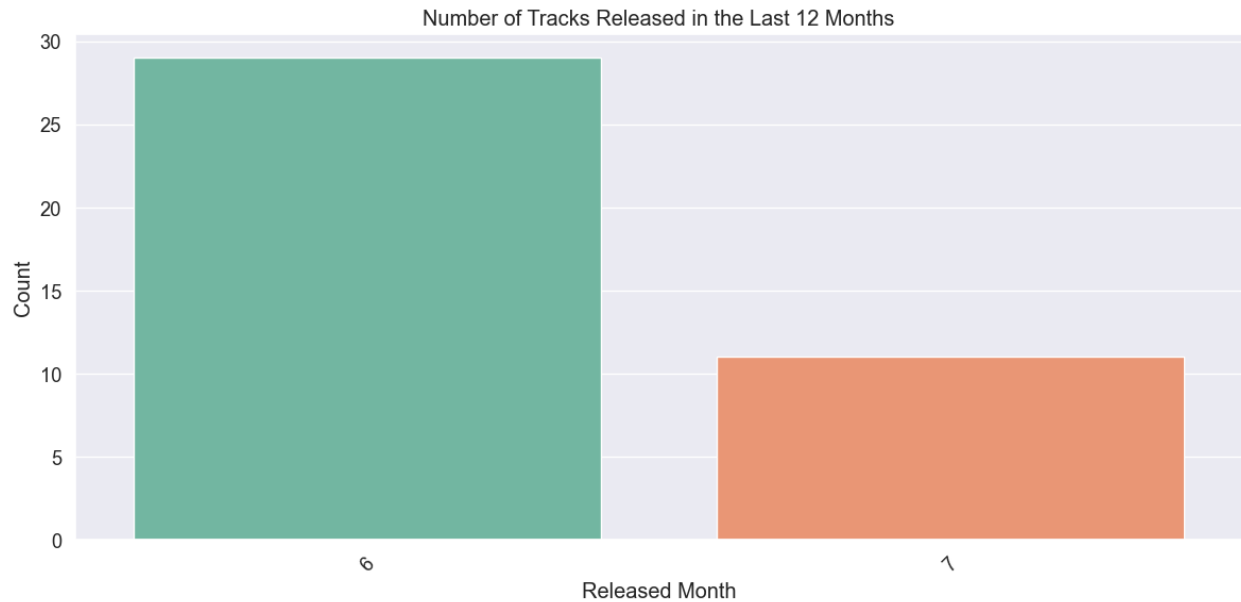
```
plt.figure(figsize=(10, 6))
sns.histplot(df['streams'], bins=30, kde=True)
plt.title('Distribution of Streams')
plt.xlabel('Streams')
plt.ylabel('Frequency')
plt.show()
```



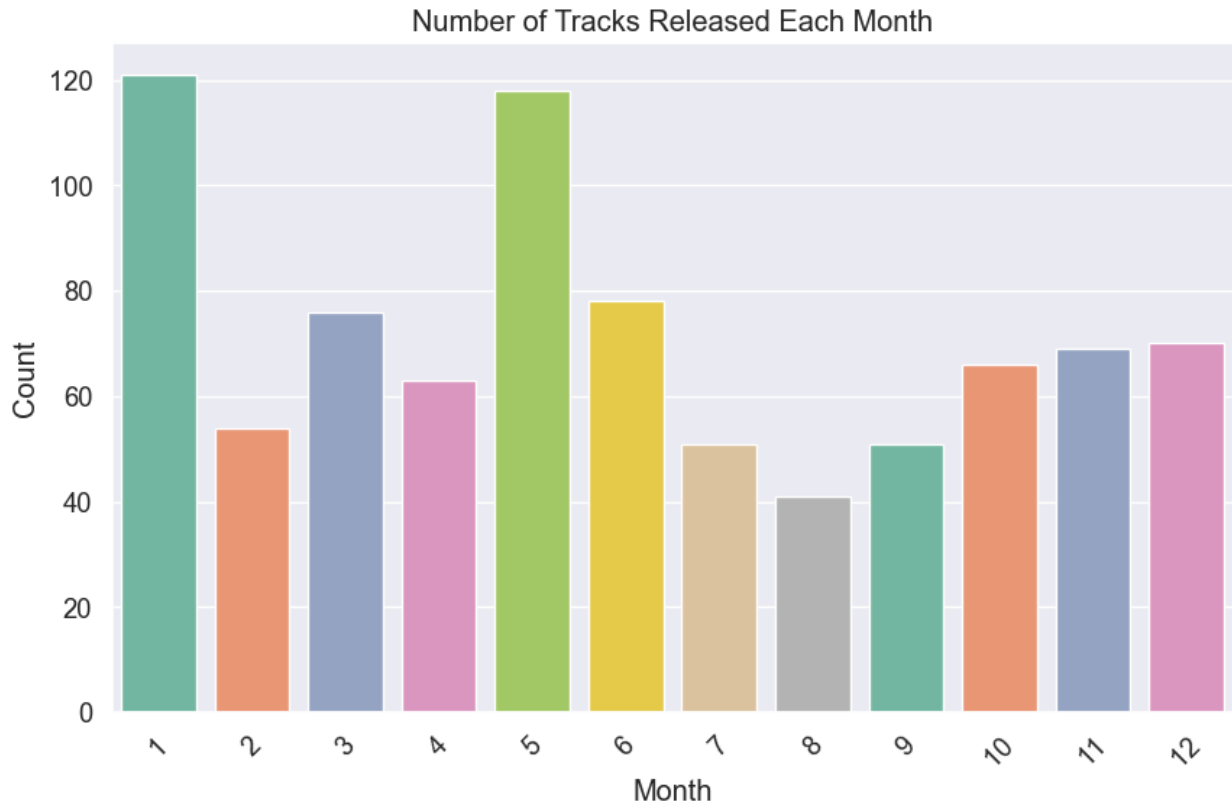
```
current_date = pd.to_datetime('today')
last_12_months = current_date - pd.DateOffset(months=12)

# Filter the DataFrame for the last 12 months
df_last_12_months = df[(df['released_year'] >= last_12_months.year) &
(df['released_month'] >= last_12_months.month)]

plt.figure(figsize=(12, 6))
sns.countplot(x='released_month', data=df_last_12_months,
palette='Set2')
plt.title('Number of Tracks Released in the Last 12 Months')
plt.xlabel('Released Month')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```



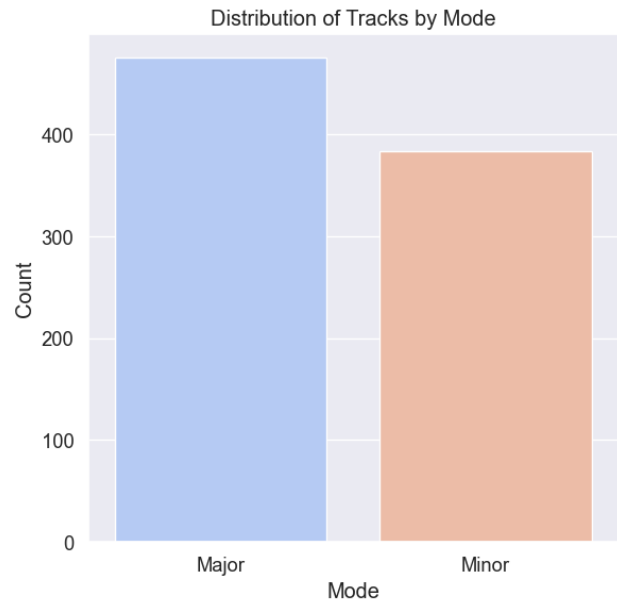
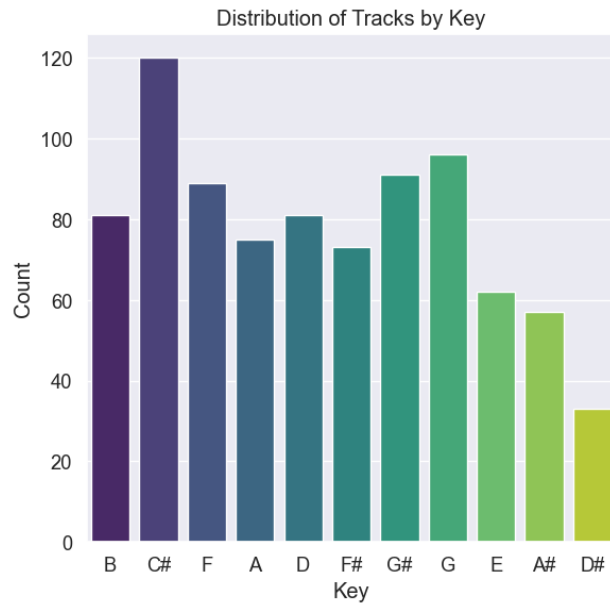
```
#Release Month Analysis
plt.figure(figsize=(10, 6))
sns.countplot(x='released_month', data=df, palette='Set2')
plt.title('Number of Tracks Released Each Month')
plt.xlabel('Month')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.show()
```



```
# Key and Mode Analysis
plt.figure(figsize=(12, 6))
plt.subplot(1, 2, 1)
sns.countplot(x='key', data=df, palette='viridis')
plt.title('Distribution of Tracks by Key')
plt.xlabel('Key')
plt.ylabel('Count')

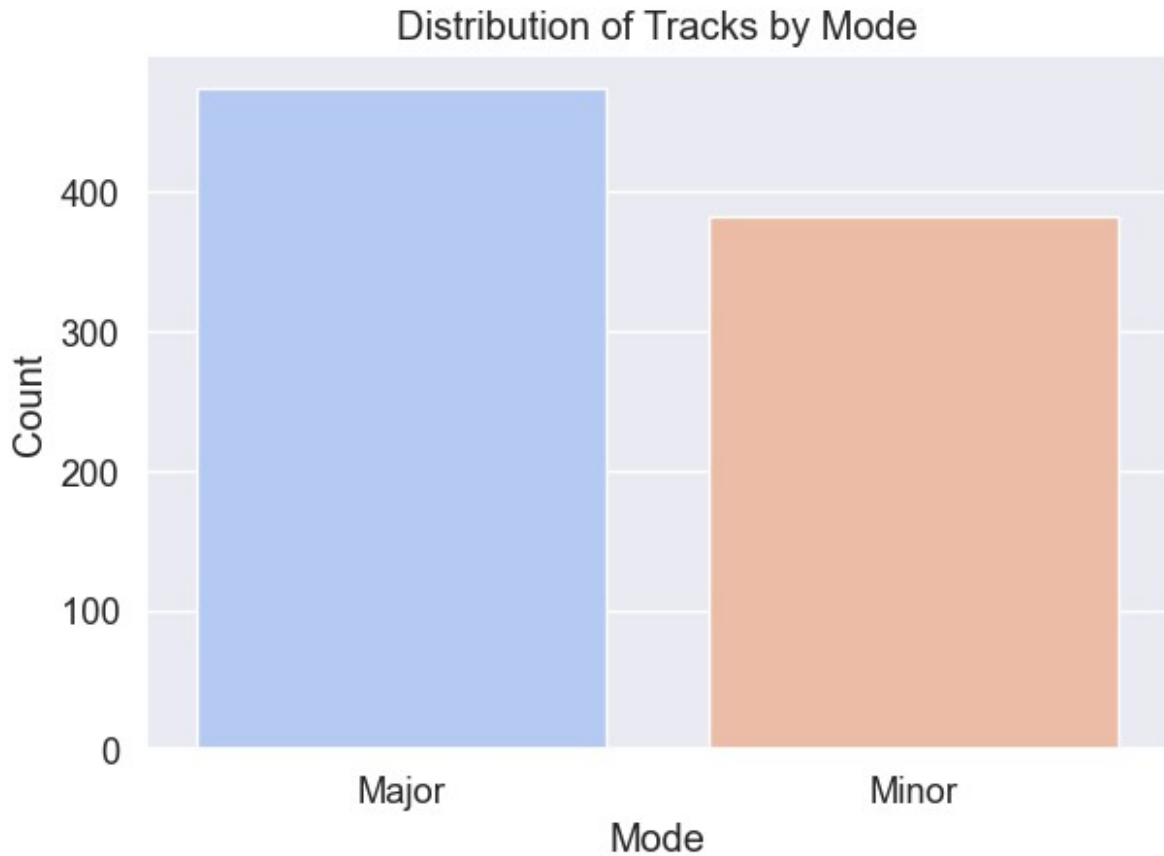
plt.subplot(1, 2, 2)
sns.countplot(x='mode', data=df, palette='coolwarm')
plt.title('Distribution of Tracks by Mode')
plt.xlabel('Mode')
plt.ylabel('Count')

plt.tight_layout()
plt.show()
```



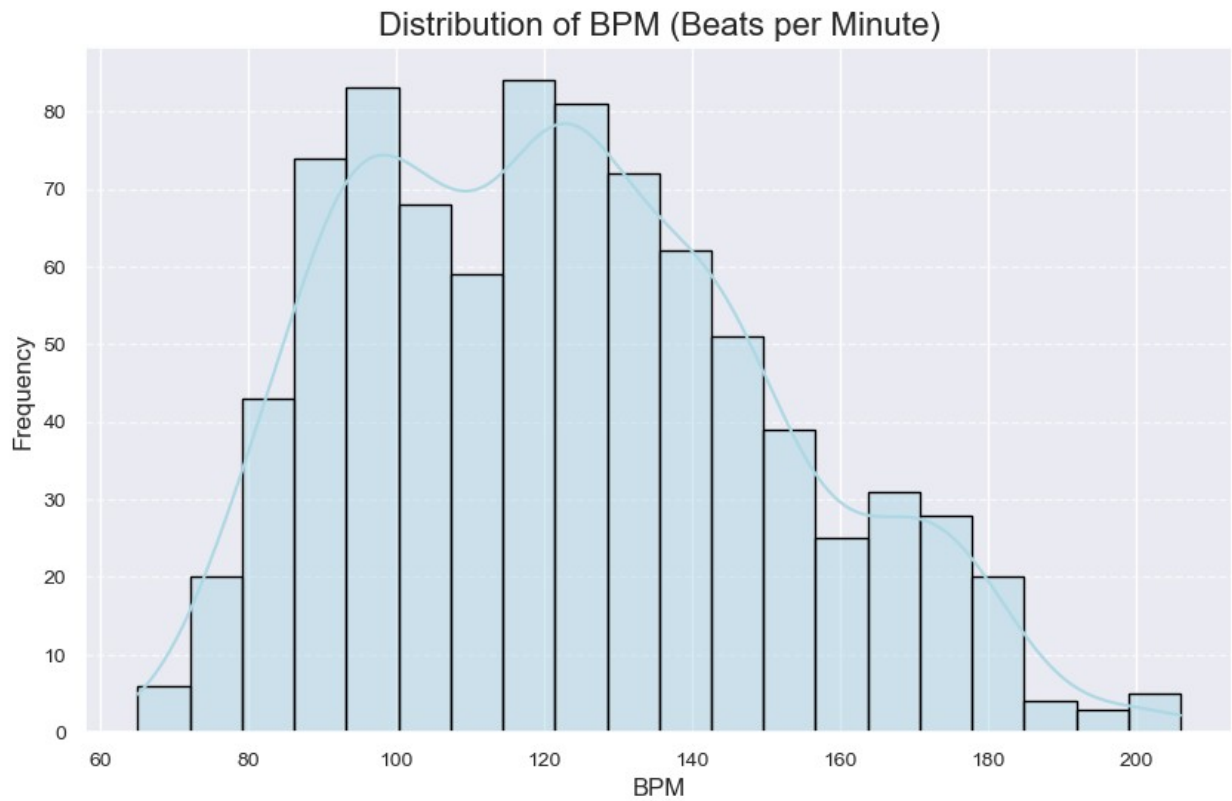
```
sns.countplot(x='mode', data=df, palette='coolwarm')
plt.title('Distribution of Tracks by Mode')
plt.xlabel('Mode')
plt.ylabel('Count')

plt.tight_layout()
plt.show()
```

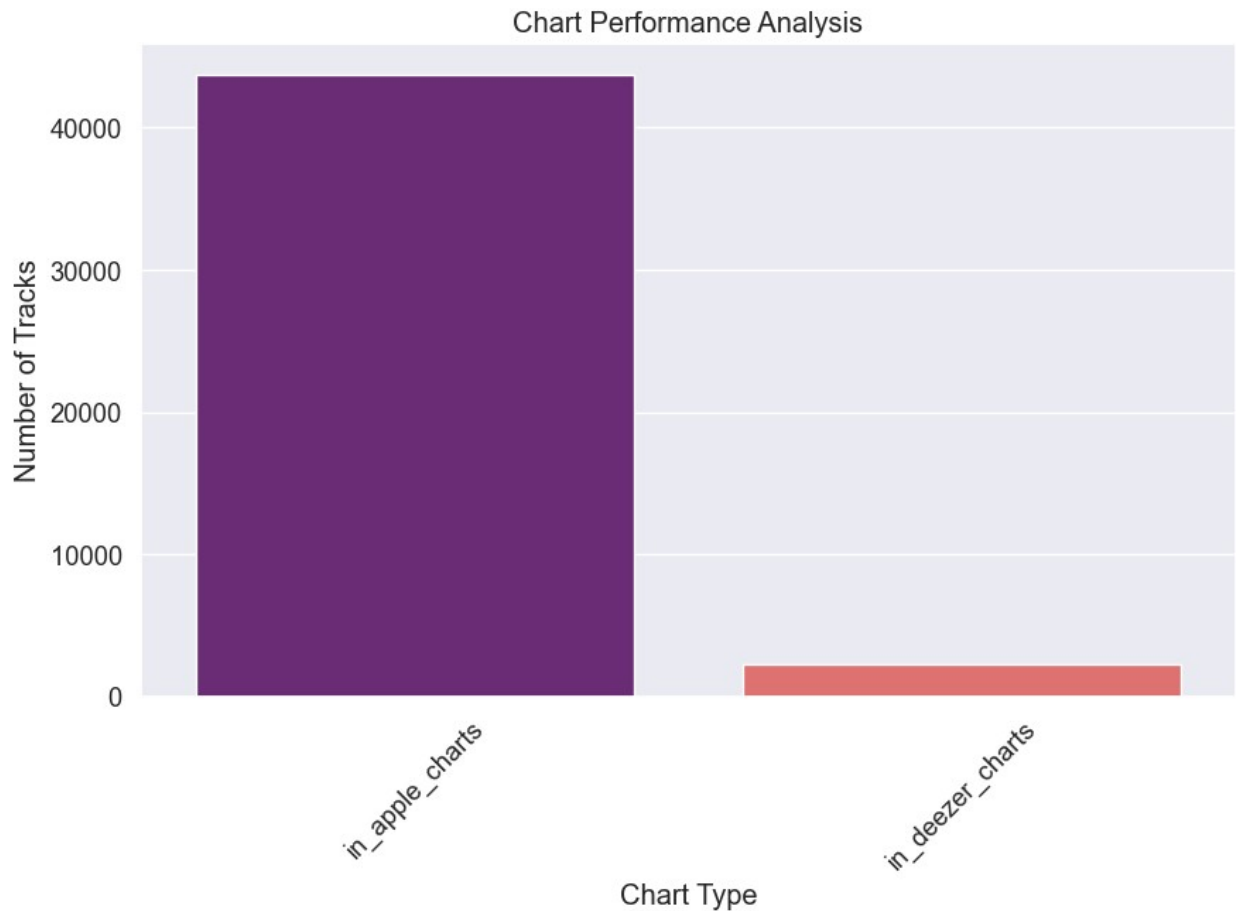


```
plt.figure(figsize=(10, 6))
sns.histplot(df['bpm'], bins=20, kde=True, color='lightblue',
edgecolor='black')
plt.title('Distribution of BPM (Beats per Minute)', fontsize=16)
plt.xlabel('BPM', fontsize=12)
plt.ylabel('Frequency', fontsize=12)
plt.xticks(fontsize=10)
plt.yticks(fontsize=10)
plt.grid(axis='y', linestyle='--', alpha=0.7)
plt.show()
```





```
plt.figure(figsize=(10, 6))
chart_columns = ['in_apple_charts', 'in_deezer_charts',
                 'in_shazam_charts']
chart_counts = df[chart_columns].sum()
sns.barplot(x=chart_counts.index, y=chart_counts.values,
            palette='magma')
plt.title('Chart Performance Analysis')
plt.xlabel('Chart Type')
plt.ylabel('Number of Tracks')
plt.xticks(rotation=45)
plt.show()
```



# Spotify Music Analysis Report

## Introduction

This report presents an in-depth analysis of Spotify music data, focusing on various aspects such as track attributes, chart performance, top artists, and musical characteristics.

## Data Preparation

The dataset was loaded from a CSV file and prepared for analysis. Irrelevant columns such as danceability, valence, energy, and others were dropped to focus on key aspects. Missing values in critical columns like 'key' and 'mode' were handled appropriately.

## Exploratory Data Analysis (EDA)

### Distribution of Streams

The histogram of stream counts revealed that most tracks have a moderate number of streams, with a peak around a certain range, indicating popular tracks.

## Tracks Released Over the Last 10 Years

The count of tracks released each year over the last 10 years showed a fluctuating trend, possibly reflecting changing music trends or industry dynamics.

## Correlation Analysis

The correlation heatmap revealed interesting relationships between track attributes. For example, danceability showed a positive correlation with energy, indicating more energetic tracks tend to be more danceable.

## Chart Performance Analysis

Analyzing the number of tracks in different charts (Spotify, Apple, Deezer, Shazam) provided insights into chart performance. Spotify and Apple charts showed higher inclusion rates compared to Deezer and Shazam charts, suggesting stronger presence and popularity on these platforms.

## Top Artists Analysis

Identifying the top 10 artists based on the number of tracks highlighted prominent artists in the dataset. This analysis can be useful for understanding audience preferences and trends related to specific artists.

## Release Month Analysis

The distribution of tracks released each month showed variations, with some months having higher release rates than others. This could be due to seasonal trends or strategic release plans by artists or labels.

## Key and Mode Analysis

Investigating the distribution of tracks by musical key and mode provided insights into the diversity of musical styles within the dataset. Common keys and modes were identified, showcasing the variety of music genres represented.

## BPM Analysis

Analyzing the distribution of beats per minute (BPM) for tracks revealed interesting patterns. Tracks with higher BPMs were often associated with energetic or dance-oriented music genres, while lower BPMs indicated slower-paced tracks.

## Conclusion

The comprehensive analysis of the Spotify music dataset has provided valuable insights into track attributes, chart performance, artist popularity, and musical characteristics. These insights can be leveraged for various purposes such as music recommendation systems, trend analysis, and audience segmentation in the music industry.

Further exploration and advanced analyses could include sentiment analysis of song lyrics (if available), geographical trends in music popularity, and genre-based analyses to delve deeper into specific music genres and their audience reception.

## Analysis Done By

Shubham A

