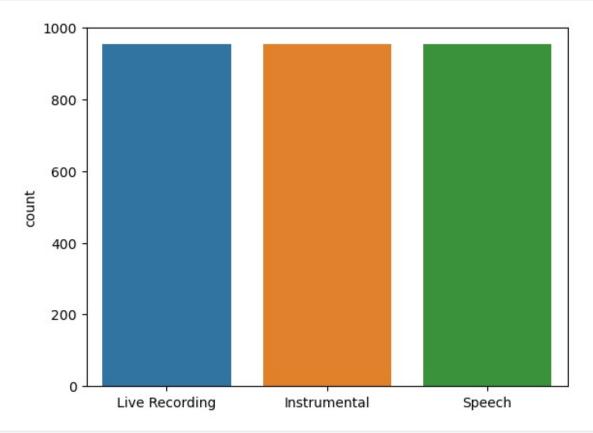
```
threshold = 50

# Create a new column 'live' based on the condition
Data['Live Recording'] = np.where(Data['liveness_%'] >= threshold, 1, 0)
Data['Instrumental'] = np.where(Data['instrumentalness_%'] >= threshold, 1, 0)
Data['Speech'] = np.where(Data['speechiness_%'] >= threshold, 1, 0)
No_Music = Data[['Live Recording', 'Instrumental', 'Speech']]
NM_Cat = ['Live Recording', 'Instrumental', 'Speech']
sns.countplot(data=No_Music)

<a href="mailto:Axes: ylabel='count'">Axes: ylabel='count'>
```

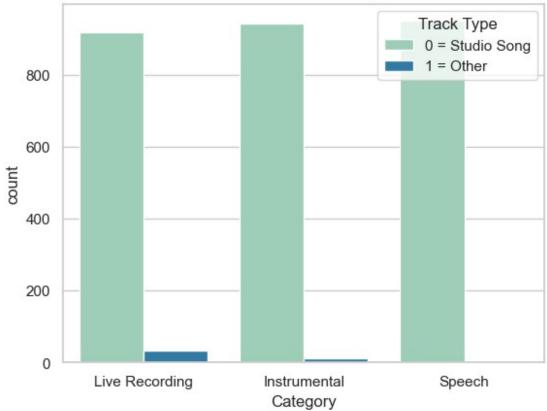


```
# Assuming No_Music is your DataFrame and NM_Cat is the list of column
names
sns.set(style="whitegrid")

# Melt the DataFrame to make it suitable for countplot
melted_data = No_Music.melt(value_vars=NM_Cat, var_name='Category',
value_name='Count')
```

```
# Create the count plot
sns.countplot(x='Category', hue='Count', data=melted_data,
palette='YlGnBu')
plt.title('Tracks: Studio Recorded Songs vs Other')
plt.legend(title='Track Type', labels=['0 = Studio Song', '1 =
Other'])
# Show the plot
plt.show()
```





```
# Get unique months in the dataset
unique_months = Data['month'].unique()

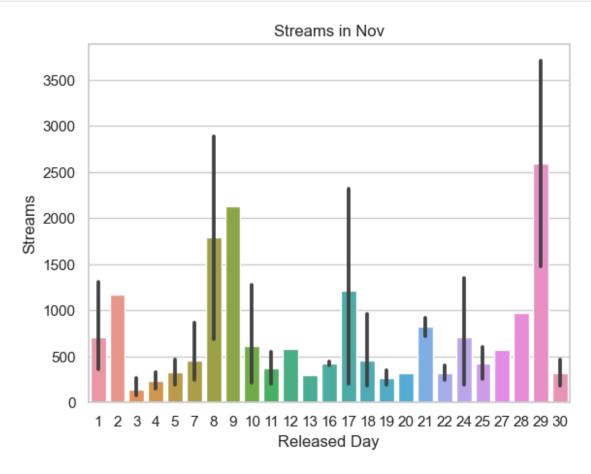
# Create a bar plot for each month
for month in unique_months:
    # Filter data for the current month
    monthly_data = Data[Data['month'] == month]

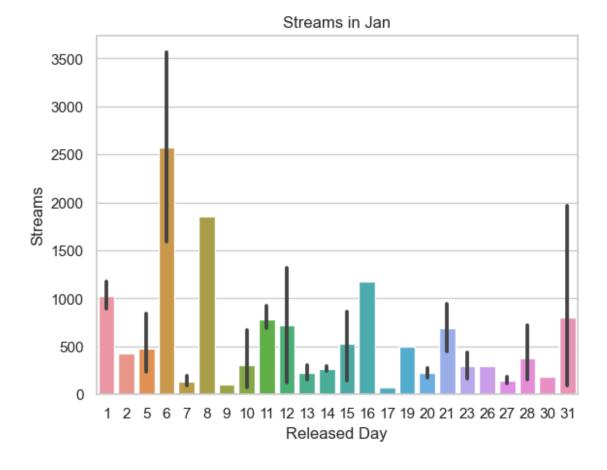
# Create a bar plot
ax = sns.barplot(x="released_day", y="streams", data=monthly_data)

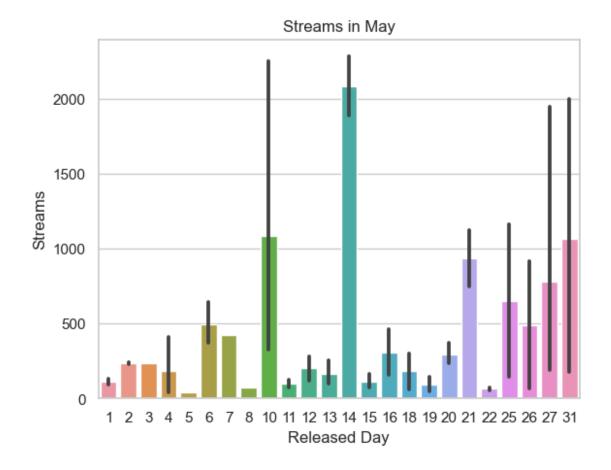
# Set labels and title
```

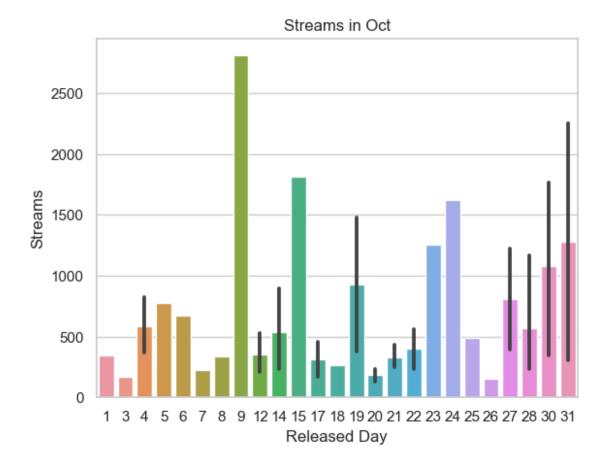
```
ax.set(xlabel="Released Day", ylabel="Streams", title=f"Streams in
{month}")

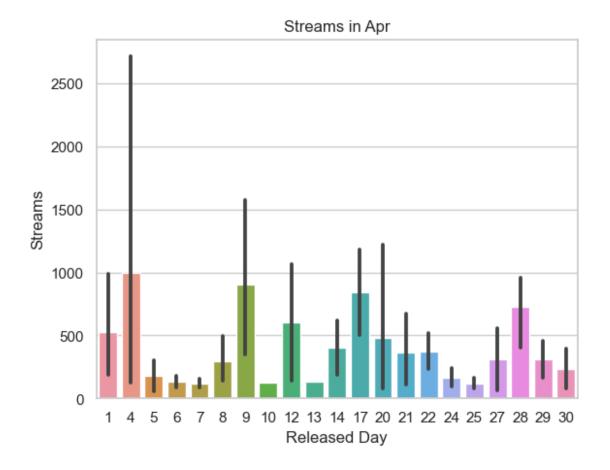
# Show the plot
plt.show()
```

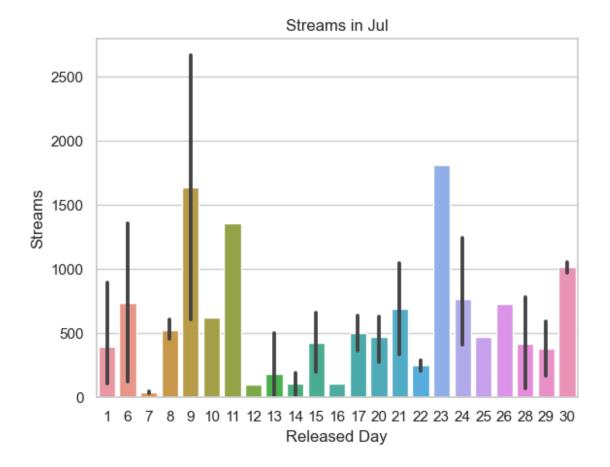


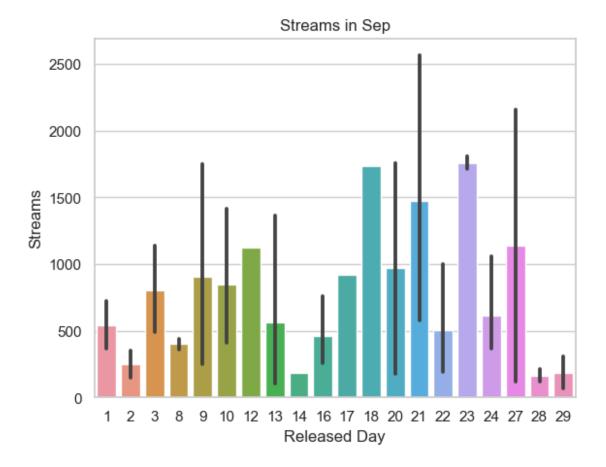


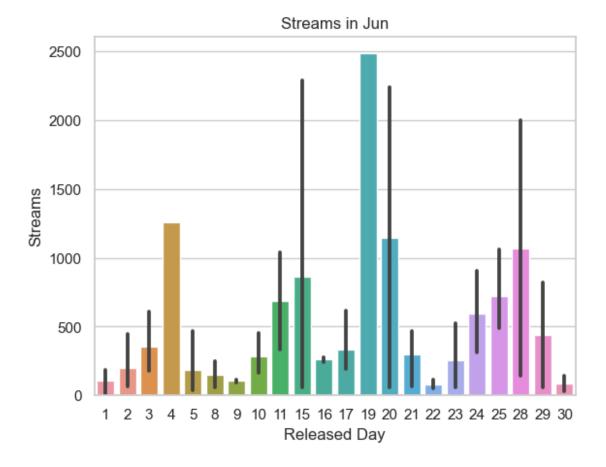


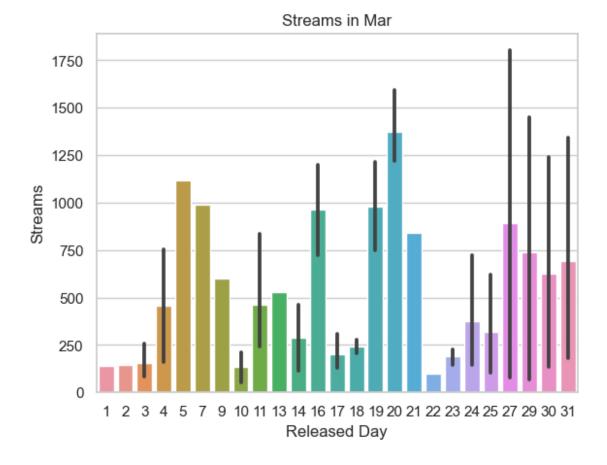


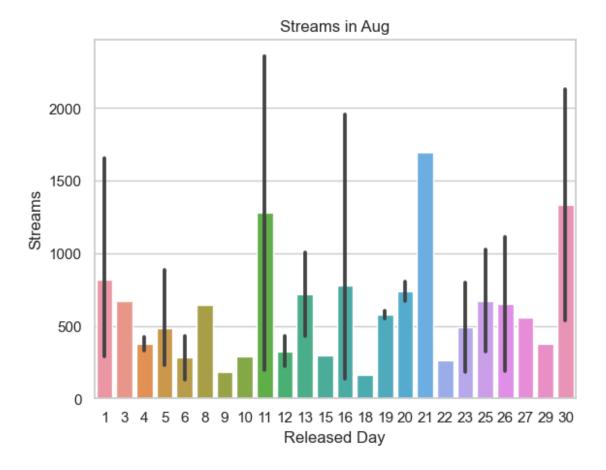


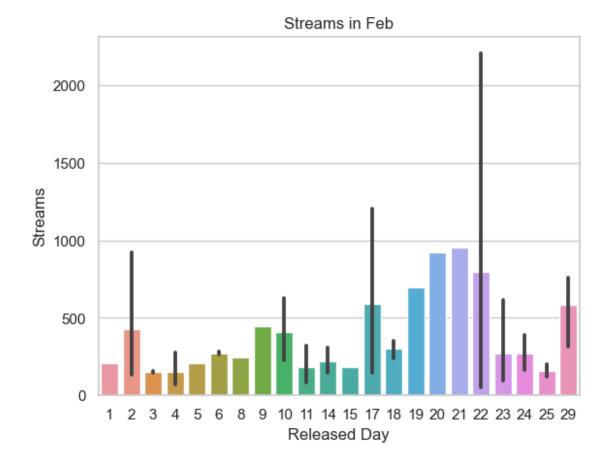


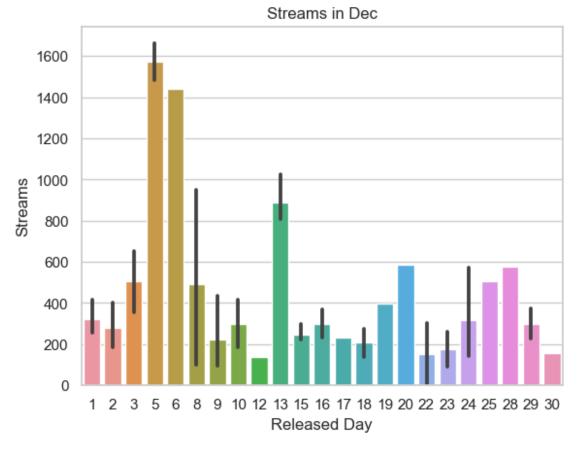












```
# Create scatter plots with different markers
sns.scatterplot(x='danceability_%', y='streams', data=Data,
markers=['o'])
sns.scatterplot(x='energy_%', y='streams', data=Data, markers=['^'])
sns.scatterplot(x='valence_%', y='streams', data=Data, markers=['*'])
sns.scatterplot(x='bpm', y='streams', data=Data, markers=['+'])

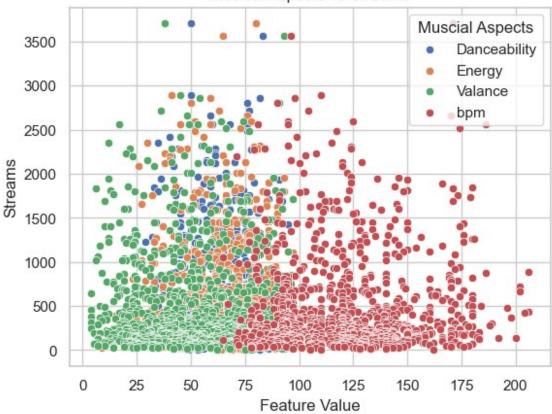
# Set plot labels
plt.xlabel('Feature Value')
plt.ylabel('Streams')

plt.title('Musical Aspects vs Streams')

plt.legend(title='Muscial Aspects', labels=['Danceability', 'Energy', 'Valance', 'bpm'])

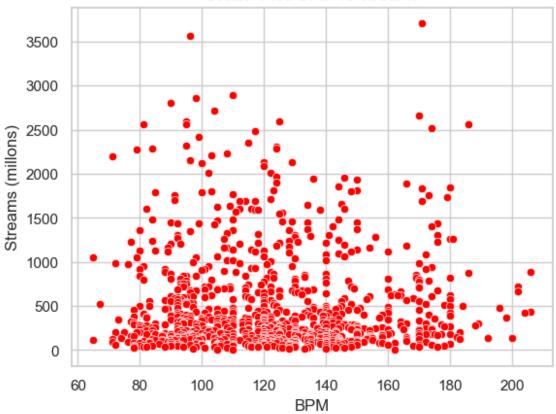
# Show the plot
plt.show()
```





```
sns.scatterplot(data=Data, x="bpm", y="streams", color='red')
plt.title('Scatter Plot: BPM vs Streams')
plt.xlabel('BPM')
plt.ylabel('Streams (millons)')
Text(0, 0.5, 'Streams (millons)')
```

## Scatter Plot: BPM vs Streams

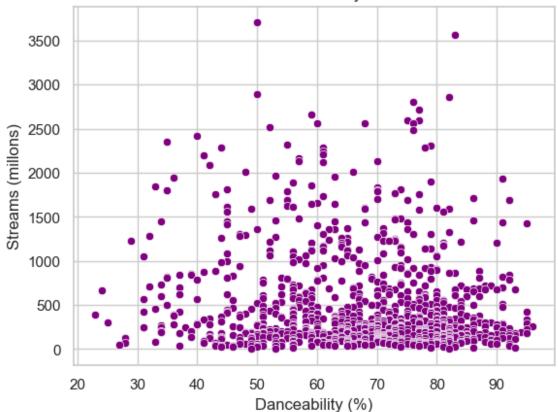


```
sns.scatterplot(data=Data, x="danceability_%", y="streams",
color='purple')

plt.title('Scatter Plot: Danceability vs Streams')
plt.xlabel('Danceability (%)')
plt.ylabel('Streams (millons)')

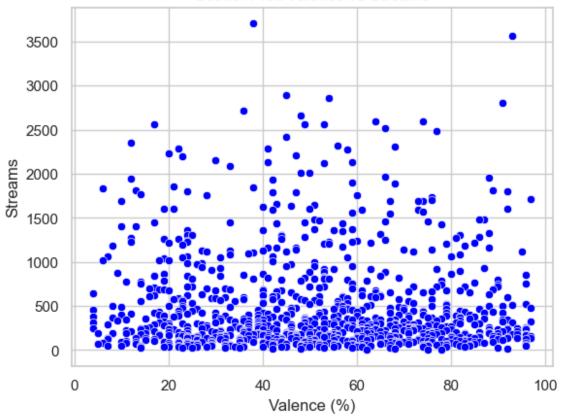
Text(0, 0.5, 'Streams (millons)')
```





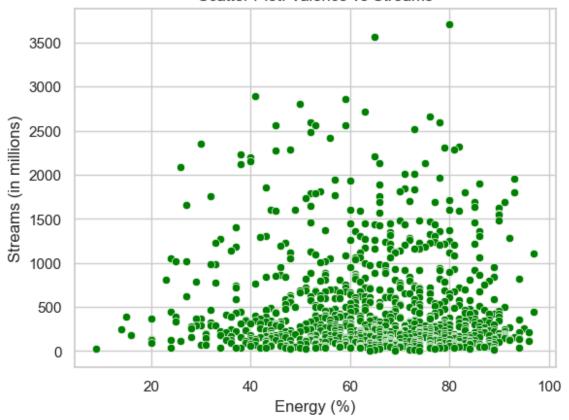
```
sns.scatterplot(data=Data, x="valence_%", y="streams", color='blue')
plt.title('Scatter Plot: Valence vs Streams')
plt.xlabel('Valence (%)')
plt.ylabel('Streams')
Text(0, 0.5, 'Streams')
```





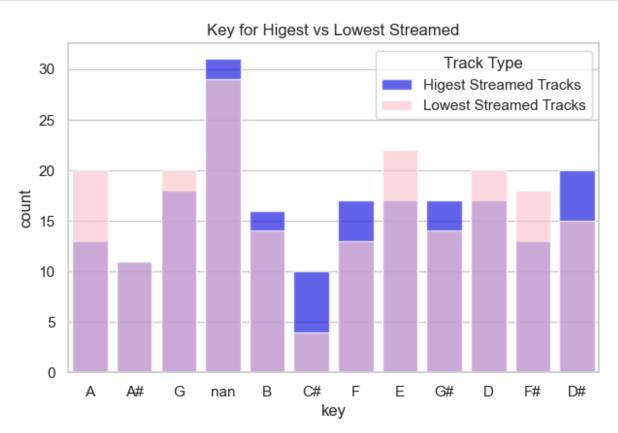
```
sns.scatterplot(data=Data, x="energy_%", y="streams", color='green')
plt.title('Scatter Plot: Valence vs Streams')
plt.xlabel('Energy (%)')
plt.ylabel('Streams (in millions)')
Text(0, 0.5, 'Streams (in millions)')
```

## Scatter Plot: Valence vs Streams



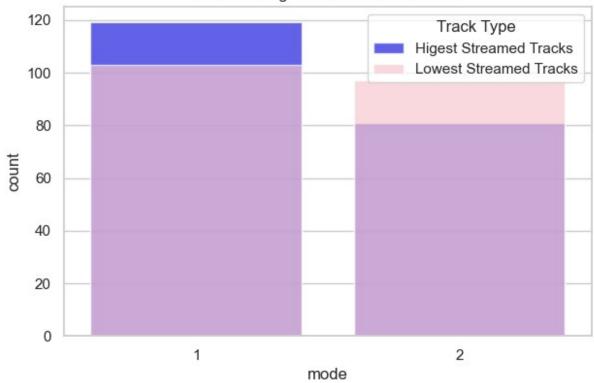
```
import seaborn as sns
import matplotlib.pyplot as plt
# Assuming 'Data' is your DataFrame
tops = [200]
tops_str = ['top200']
low = [200]
low str = ['low200']
# Create a 2x2 subplot for top and low counts
fig, axes = plt.subplots(1, figsize=(12, 8))
for i, (top col, low col) in enumerate(zip(tops_str, low_str), 1):
    # Plot top counts
    plt.subplot(2, 2, i)
    sns.countplot(x="key", data=Data.head(tops[i-1]),
color='blue',alpha=0.7)
    plt.title(top_col, fontsize=10)
    # Plot low counts on the same subplot
    plt.subplot(2, 2, i)
```

```
sns.countplot(x="key", data=Data.tail(low[i-1]), color='pink',
alpha=0.7)
    plt.title(low col, fontsize=10)
keys_encoded = [0, 1, 9, 11, 2, 3, 7, 6, 10, 4, 8, 5]
keys_original = ['A', 'A#', 'G', 'nan', 'B', 'C#', 'F', 'E', 'G#',
'D', 'F#', 'D#']
plt.legend(title='Track Type', labels=['Higest Streamed Tracks',
'Lowest Streamed Tracks', 'Overlap'])
plt.title('Key for Higest vs Lowest Streamed')
plt.xticks(range(len(keys encoded)), keys original)
# Adjust layout for better spacing
plt.tight layout()
plt.show()
/var/folders/j2/gt7wq0h5751027qf23w96tqw0000gn/T/
ipykernel 87432/1126530090.py:17: MatplotlibDeprecationWarning: Auto-
removal of overlapping axes is deprecated since 3.6 and will be
removed two minor releases later; explicitly call ax.remove() as
needed.
  plt.subplot(2, 2, i)
```



```
import seaborn as sns
import matplotlib.pyplot as plt
# Assuming 'Data' is your DataFrame
tops = [200]
tops str = ['top200']
low = [200]
low str = ['low200']
# Create a 2x2 subplot for top and low counts
fig, axes = plt.subplots(1, figsize=(12, 8))
for i, (top col, low col) in enumerate(zip(tops_str, low_str), 1):
   # Plot top counts
   plt.subplot(2, 2, i)
    sns.countplot(x="mode", data=Data.head(tops[i-1]),
color='blue',alpha=0.7)
   plt.title(top col, fontsize=10)
   # Plot low counts on the same subplot
   plt.subplot(2, 2, i)
    sns.countplot(x="mode", data=Data.tail(low[i-1]), color='pink',
alpha=0.7)
   plt.title(low col, fontsize=10)
'D', 'F#', 'D#']
plt.legend(title='Track Type', labels=['Higest Streamed Tracks',
'Lowest Streamed Tracks', 'Overlap'])
plt.title('Mode for Higest vs Lowest Streamed')
#plt.xticks(range(len(keys encoded)), keys original)
# Adjust layout for better spacing
plt.tight_layout()
plt.show()
/var/folders/j2/gt7wg0h5751027gf23w96tgw0000gn/T/
ipykernel 87432/700858123.py:17: MatplotlibDeprecationWarning: Auto-
removal of overlapping axes is deprecated since 3.6 and will be
removed two minor releases later; explicitly call ax.remove() as
needed.
  plt.subplot(2, 2, i)
```

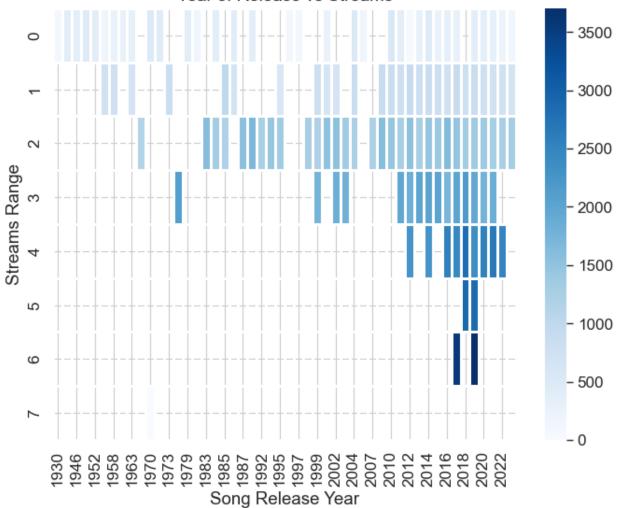
## Mode for Higest vs Lowest Streamed



```
plt.figure(figsize=(8,6))
sns.set_context('paper', font_scale=1.4)

flights = Data.pivot_table(columns='released_year',
index='streams_range', values='streams')
#Data.sort_values('month_integer')
# You can separate data with lines
sns.heatmap(flights, cmap='Blues', linecolor='white', linewidth=1)
plt.title("Year of Release vs Streams")
plt.ylabel('Streams Range')
plt.xlabel('Song Release Year')
Text(0.5, 33.91666666666665, 'Song Release Year')
```

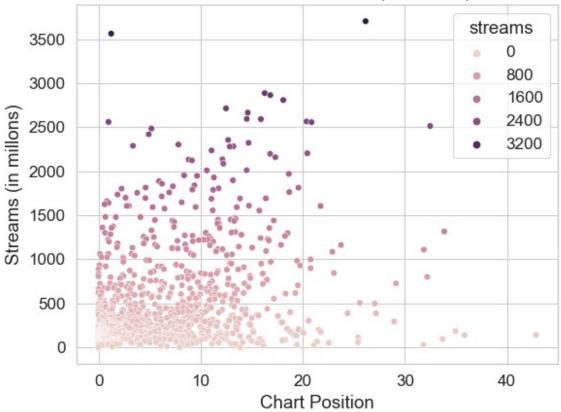




```
rel = sns.scatterplot(data=Data, x='chart', y='streams',
hue='streams')
plt.title('Chart Position vs Streams (in millions)')
plt.ylabel('Streams (in millons)')
plt.xlabel('Chart Position')
#rel.invert_xaxis()

Text(0.5, 0, 'Chart Position')
```

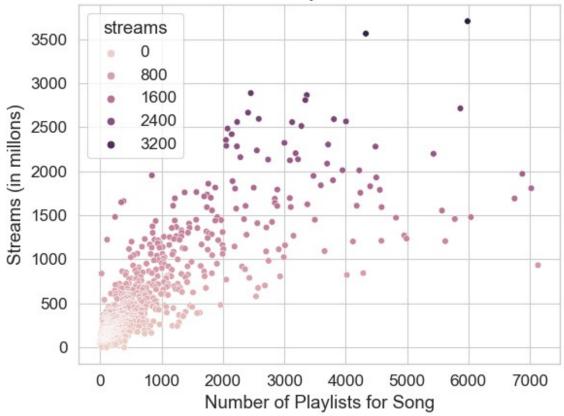




```
playlist =
Data[['in apple playlists','in spotify playlists','in deezer playlists
','playlist','streams']]
charts = Data [['in_spotify_charts', 'in_apple_charts',
'in deezer charts', 'in shazam charts', 'chart', 'streams']]
# Assuming 'Data' is your DataFrame
weights playlists = {
    'in apple playlists':
Data['in apple playlists'].corr(Data['streams']),
    'in spotify playlists':
Data['in_spotify_playlists'].corr(Data['streams']),
    'in deezer playlists':
Data['in deezer playlists'].corr(Data['streams']),
# Normalize weights to sum to 1
total weight playlists = sum(weights playlists.values())
weights playlists = {key: value / total weight playlists for key,
value in weights playlists.items()}
# Calculate the weighted sum for 'playlist'
Data['playlist'] = ((
```

```
weights playlists['in apple playlists'] *
Data['in apple playlists'] +
    weights playlists['in spotify playlists'] *
Data['in spotify playlists'] +
    weights playlists['in deezer playlists'] *
Data['in deezer playlists']
# Weights for 'chart' columns
weights charts = {
    'in spotify charts':
Data['in spotify charts'].corr(Data['streams']),
    'in apple charts': Data['in apple charts'].corr(Data['streams']),
    'in deezer charts':
Data['in deezer charts'].corr(Data['streams']),
    'in shazam charts':
Data['in shazam charts'].corr(Data['streams']),
}
# Normalize weights to sum to 1
total weight charts = sum(weights charts.values())
weights charts = {key: value / total weight charts for key, value in
weights charts.items()}
# Calculate the weighted sum for 'chart'
Data['chart'] = ((
    weights charts['in spotify charts'] * Data['in spotify charts'] +
    weights charts['in apple charts'] * Data['in apple charts'] +
    weights_charts['in_deezer_charts'] * Data['in_deezer_charts'] +
    weights charts['in shazam charts'] * Data['in shazam charts']
)/4)
rel2 = sns.scatterplot(data=Data, x='playlist', y='streams',
hue='streams', cmap="magma")
plt.title('Number of Playlists vs Streams')
plt.ylabel('Streams (in millons)')
plt.xlabel('Number of Playlists for Song')
/Users/leah/anaconda3/lib/python3.11/site-packages/seaborn/
relational.py:573: UserWarning: No data for colormapping provided via
'c'. Parameters 'cmap' will be ignored
  points = ax.scatter(x=x, y=y, **kws)
Text(0.5, 0, 'Number of Playlists for Song')
```





```
NameError
                                           Traceback (most recent call
last)
Cell In[3], line 1
----> 1 df artists =
Data.assign(artists all=Data['artists all'].str.split(',')).explode('a
rtists all')
      4 df artists[['track name', 'artists all', 'artist count',
'released year'
                released month', 'released_day',
'in spotify_playlists',
               'in_spotify_charts', 'streams', 'in_apple_playlists',
'in apple charts',
   (\ldots)
               'instru_per', 'live_per', 'speech_per', 'track_id',
     15
'art combined id'
               'date'll.ffill()
     18 df_artists['artists_id'] =
df artists['artists all'].apply(lambda x: le.fit transform([x])[0])
```

NameError: name 'Data' is not defined

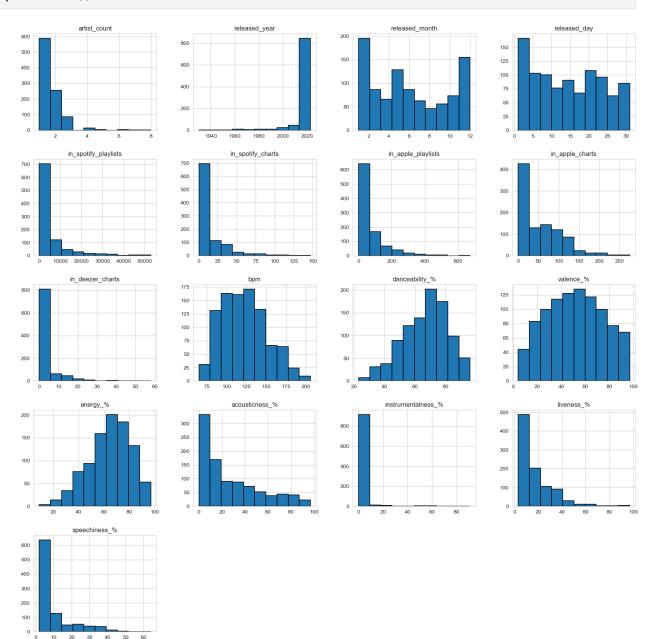
```
import pandas as pd
import numpy as np
from sklearn.model selection import train test split
from sklearn.linear model import LogisticRegression
from sklearn.metrics import confusion matrix, accuracy score
from sklearn.preprocessing import StandardScaler, OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.metrics import confusion matrix, accuracy score
from sklearn.preprocessing import LabelEncoder
from sklearn.model selection import train test split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import confusion matrix, accuracy_score
from sklearn.model selection import cross val score
from sklearn.tree import DecisionTreeClassifier
from sklearn.model selection import train test split
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy score
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy score
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels.api as sm
from sklearn.preprocessing import LabelEncoder
#Libraries for data processing
from sklearn.preprocessing import LabelEncoder
from sklearn.model selection import train test split
#SVM
from sklearn import svm
#Evaluation metrics
from sklearn.metrics import classification report, confusion matrix
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn import svm
from sklearn.metrics import confusion matrix, classification report
from sklearn.model selection import train test split
from sklearn.datasets import make classification
# import Dataset and Data cleaning
Data = pd.read csv("spotify1.csv", encoding='ISO-8859-1')
Data.head(5)
```

		track_r	name artist	(s)_name	artist_d	count		
\ 0 Seven (	feat. Latto	) (Explicit Ve	er.) Latto, J	ung Kook		2		
1		L	.ALA Myk	e Towers		1		
2		vamp	oire Olivia	Rodrigo		1		
3		Cruel Sun	nmer Tayl	or Swift		1		
4		WHERE SHE (	GOES B	ad Bunny		1		
<pre>release in_spotify</pre>		eased_month r \	released_day					
0	2023	` 7	14		ŗ	553		
1	2023	3	23		14	174		
2	2023	6	30		13	397		
3	2019	8	23		78	358		
4	2023	5	18		3.	133		
in_spot mode \	ify_charts	streams ir	n_apple_playli	sts	bpm key			
0 Major	147	141381703		43	125 B			
1 Major	48	133716286		48	92 C#			
2	113	140003974		94	138 F			
Major 3	100	800840817		116	170 A			
Major 4	50	303236322		84	144 A			
Minor								
<pre>danceability_% valence_% energy_% acousticness_% instrumentalness_ % \</pre>								
0 0	80	89 8	33	31				
1	71	61 7	4	7				
0 2	51	32 5	53	17				
0 3	55	58 7	<b>'</b> 2	11				
3 0 4	65		80	14				
63	03	25	, 0	± 1				

liveness_%	speechiness_%
_8	4
10	4
31	6
11	15
11	6
	_8

## [5 rows x 24 columns]

Data.hist(edgecolor='black', linewidth=1.2, figsize=(20, 20))
plt.show()



Data.describe()									
count mean std min 25% 50% 75% max	953 1 0 1 1 1 2	t_coun .00000 .55613 .89304 .00000 .00000 .00000	0 953 9 2018 4 13 0 1930 0 2020 0 2022	sed_year 3.000000 3.238195 1.116218 0.000000 0.000000 2.000000 2.000000 3.000000	9	sed_mo 53.000 6.033 3.566 1.000 3.000 6.000 9.000	0000 0578 0435 0000 0000 0000	released_d 953.0000 13.9307 9.2019 1.0000 6.0000 13.0000 22.0000	00 45 49 00 00 00
count mean std min 25% 50% 75% max	in_sp	52 78 8 22 55	playlists 53.000000 00.124869 97.608990 31.000000 75.000000 24.000000 42.000000	- · · · · · · · · · · · · · · · · · · ·	19.5 0.0 0.0 3.0	00000 09444 75992 00000 00000 00000	in_ap	953.00 953.00 67.81 86.44 0.00 13.00 34.00 88.00 672.00	0000 2172 1493 0000 0000 0000 0000
% \	in_ap	ple_ch	arts in	_deezer_d	charts		bpm	danceabi	lity_
count		953.00	0000	953.0	900000	953.0	00000	953	.00000
mean		51.90	8709	2.0	666317	122.5	40399	66	.96957
std		50.63	0241	6.0	935599	28.0	57802	14	.63061
min		0.00	0000	0.0	900000	65.0	000000	23	.00000
25%		7.00	0000	0.0	900000	100.0	00000	57	.00000
50%		38.00	0000	0.0	900000	121.0	00000	69	.00000
75%		87.00	0000	2.0	900000	140.0	000000	78	.00000
max		275.00	0000	58.0	900000	206.0	00000	96	.00000
livene		nce_%	energy	/_% acou	usticne	SS_%	instru	ımentalnes	S_%
livene:	$9\overline{5}3.0$		953.000	900	953.00	0000		953.000	000
953.000 mean	51.4	31270	64.279	119	27.05	7712		1.581	322
18.2130 std	23.4	80632	16.550	526	25.99	6077		8.409	800
13.7112 min 3.0000	4.0	00000	9.000	900	0.00	0000		0.000	000

```
25%
        32.000000
                     53.000000
                                       6.000000
                                                            0.000000
10.000000
50%
        51.000000
                     66.000000
                                      18.000000
                                                            0.000000
12.000000
75%
        70,000000
                     77,000000
                                      43.000000
                                                            0.000000
24.000000
        97.000000
                     97.000000
                                      97.000000
                                                           91.000000
max
97.000000
       speechiness %
          953.000000
count
           10.131165
mean
std
            9.912888
            2,000000
min
25%
            4.000000
50%
            6.000000
75%
           11.000000
           64.000000
max
Data.shape
(953, 24)
Data.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
                            953 non-null
     artist count
                                              int64
 3
     released year
                            953 non-null
                                              int64
     released_month
 4
                            953 non-null
                                              int64
 5
     released day
                            953 non-null
                                              int64
     in_spotify_playlists
                            953 non-null
 6
                                             int64
 7
     in spotify charts
                            953 non-null
                                              int64
 8
     streams
                            953 non-null
                                              object
                            953 non-null
 9
     in_apple_playlists
                                              int64
 10
     in apple charts
                            953 non-null
                                              int64
     in deezer playlists
 11
                            953 non-null
                                             object
 12
     in deezer charts
                            953 non-null
                                              int64
     in shazam charts
 13
                            903 non-null
                                             object
 14
     bpm
                            953 non-null
                                             int64
                            858 non-null
 15
     key
                                             object
 16
     mode
                            953 non-null
                                             object
 17
     danceability %
                            953 non-null
                                              int64
     valence_%
                            953 non-null
 18
                                              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
def convert mode(mode):
    if mode == 'Major':
        return 1
    elif mode == 'Minor':
        return 2
    else:
        return None
Data['mode'] = Data['mode'].apply(convert mode)
# Singers in total
artist names = Data['artist(s) name'].unique().tolist()
for artist in artist names:
    print(artist)
total_artists = Data['artist(s)_name'].nunique()
print(f'Total number of artists: {total artists}')
Latto, Jung Kook
Myke Towers
Olivia Rodrigo
Taylor Swift
Bad Bunny
Dave, Central Cee
Eslabon Armado, Peso Pluma
Quevedo
Gunna
Peso Pluma, Yng Lvcas
Bad Bunny, Grupo Frontera
NewJeans
Miley Cyrus
David Kushner
Harry Styles
SZA
Fifty Fifty
Billie Eilish
Feid, Young Miko
Jimin
Gabito Ballesteros, Junior H, Peso Pluma
Arctic Monkeys
Bizarrap, Peso Pluma
The Weeknd, Madonna, Playboi Carti
Fuerza Regida
```

```
R��ma, Selena G
Tainy, Bad Bunny
Morgan Wallen
Dua Lipa
Troye Sivan
Peso Pluma, Grupo Frontera
The Weeknd, 21 Savage, Metro Boomin
Karol G, Shakira
Big One, Duki, Lit Killah, Maria Becerra, FMK, Rusherking, Emilia,
Tiago pzk
Yahritza Y Su Esencia, Grupo Frontera
Junior H, Peso Pluma
Post Malone, Swae Lee
Bebe Rexha, David Guetta
Tyler, The Creator, Kali Uchis
Nicki Minaj, Aqua, Ice Spice
OneRepublic
The Weeknd
The Weeknd, Daft Punk
Ariana Grande, The Weeknd
Feid, Myke Towers, Sky Rompiendo
David Guetta, Anne-Marie, Coi Leray
Karol G
Peggy Gou
Tom Odell
Kali Uchis
Manuel Turizo
dennis, MC Kevin o Chris
PinkPantheress, Ice Spice
Charlie Puth, BTS, Jung Kook
Rauw Alejandro, ROSALi;
Ozuna, Feid
Chris Molitor
Libianca
Post Malone
Rauw Alejandro, Bizarrap
Glass Animals
JVKE
The Neighbourhood
Bizarrap, Quevedo
Coldplay
d4vd
Sam Smith, Kim Petras
Yandel, Feid
Maria Becerra
Vance Joy
Em Beihold, Stephen Sanchez
Mc Livinho, DJ Matt D
Justin Bieber, The Kid Laroi
```

```
Marshmello, Manuel Turizo
Lewis Capaldi
Chencho Corleone, Bad Bunny
Jain
Ayparia, unxbected
Luke Combs
Doechii
Bad Bunny, Eladio Carrion
Jung Kook
J. Cole, Lil Durk
Lana Del Rey
BTS
Future, Metro Boomin, Don Toliver
Eminem
Carin Leon
Dua Lipa, Elton John, Pnau
Ruth B.
Imagine Dragons
Adele
Kendrick Lamar, Jay Rock
Anggi Marito
Jasiel Nuï; ½ï; ½ez, Peso P
Tears For Fears
J. Cole
Loreen
Taiu, Milo i
Sebastian Yatra, Manuel Turizo, Bezï
Calvin Harris, Ellie Goulding
Sabrina Carpenter
Carin Leon, Grupo Frontera
Peso Pluma
Feid
Ray Dalton, Ryan Lewis, Macklemore
Billie Eilish, Khalid
(G)I-DLE
Shakira, Bizarrap
Swae Lee, A Boogie Wit da Hoodie, Metro Boomin, NAV
Travis Scott, Young Thug, Metro Boomin
Sachin-Jigar, Shadab Faridi, Altamash Faridi, Amitabh Bhattacharya,
Varun Jain
Ed Sheeran
Veigh, Byga Beatz, Supernova Ent, Prod Malax
Mc Pedrinho, DJ 900
Sog, Ryan Castro, Peso Pluma
The Weeknd, Lana Del Rey
Bomba Est��reo, Bad B
Keane
New West
Chino Pacas
```

Natanael Cano Don Omar, Lucenzo Ana Castela, AgroPlay Bruno Mars Big One, FMK, Ke personajes Drake, 21 Savage Justin Bieber Chris Brown Natanael Cano, Peso Pluma Daddy Yankee Drake, WizKid, Kyla Miguel The Police Lord Huron Rihanna, Calvin Harris Alec Benjamin A\$AP Rocky, Metro Boomin, Roisee YOASOBI Rosa Linn One Direction Oscar Maydon, Junior H Radiohead James Arthur Cigarettes After Sex Shubh Frank Ocean Hozier Fuerza Regida, Grupo Frontera Hotel Ugly Creedence Clearwater Revival Plan B Feid, Mora, Saiko, Quevedo Swae Lee, Lil Wayne, Offset, Metro Boomin Kodak Black, NLE Choppa, Muni Long, JVKE, Jimin Shakira Natanael Cano, Gabito Ballesteros, Peso Pluma Arijit Singh, Sachin-Jigar, Amitabha Bhattacharya Stray Kids Becky G, Peso Pluma Metro Boomin, Coi Leray Kaliii, Kaliii Junior H, Eden Mu½ï Abhijay Sharma, Riar Saab Semicenk, Do��u JIS00 Future, Lil Uzi Vert, Metro Boomin RAYE, 070 Shake Arcangel, Bad Bunny

```
Taylor Swift, Ice Spice
Future, Chris Brown, Metro Boomin
BLESSD, Peso Pluma
Fuerza Regida, Natanael Cano
Lana Del Rey, Taylor Swift
Coi Leray
ENHYPEN
Lil Durk, Morgan Wallen
James Blake, Metro Boomin
Israel & Rodolffo, Mari Fernandez
Steve Lacy
Beyonci¿
Halsey, Suga
Nicky Jam, Feid
Sean Paul, Feid
Drake
Kendrick Lamar, Beyonci;
WizKid, Toian, Metro Boomin, Don Toliver, Beam
Simone Mendes
Tini
Lil Uzi Vert
Sia
Stephen Sanchez
MC Caverinha, KayBlack
Offset, JID
Zï;½ï;½ Neto & Crist
Migrantes, LiL CaKe, Nico Valdi
Tory Lanez
Joji
Mambo Kingz, DJ Luian, Anuel Aa
Nile Rodgers, LE SSERAFIM
Agust D
Rich The Kid, Matuï¿
Labrinth
Karol G, Romeo Santos
Lizzy McAlpine
IVE
sped up 8282
Cartel De Santa, La Kelly
SEVENTEEN
NLE Choppa
The Weeknd, Future
Nicki Minaj, Ice Spice
Grupo Marca Registrada, Grupo Frontera
Fuerza Regida, Peso Pluma
TAEYANG, Lisa
Raim Laode
ROSALi;
Meghan Trainor
```

```
sped up nightcore, ARIZONATEARS, Lil Uzi Vert
Grupo Frontera
TV Girl
INTERWORLD
Quevedo, Jhayco
El Chachito, Junior H
The Weeknd, Gesaffelstein
RM, Colde
BLACKPINK
Nengo Flow, Bad Bunny
Nicky Youre, Dazy
Eminem, Dina Rae
IU, Agust D
Arcangel, Bizarrap
Tyler, The Creator
Tii;½i;½sto, Tate M
NF
Yng Lvcas
Kanii, PR1ISVX
Mari; lijlia Mendo
Zach Bryan
Melanie Martinez
Fuerza Regida, Chino Pacas
Mae Stephens
Feid, Mora
Kenia OS
Kordhell
Coolio, L.V.
Karol G, Ovy On The Drums
The Walters
Mac DeMarco
Jack Black
Duki, NICKI NICOLE, Cris Mj, Standly, Stars Music Chile
Brray, Rauw Alejandro, Lyanno
Linkin Park
Gorillaz, Bad Bunny
j-hope, J. Cole
Robin Schulz, Oliver Tree
Don Toliver, Future, Justin Bieber
Niall Horan
Central Cee
Mahalini
Arijit Singh, Sachin-Jigar
ThxSoMch
Myke Towers, Quevedo
Morgan Wallen, Eric Church
Nicki Minaj
Kaifi Khalil
Karol G, Quevedo
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```
Lady Gaga
Ozuna, Shakira
Leo Santana
Treyce
Post Malone, Doja Cat
MC Xenon, Os Gemeos da Putaria
Imagine Dragons, League of Legends, JID, Arcane
XXXTENTACION
Israel & Rodolffo, Ana Castela
J Balvin, Bad Bunny
Yuridia, Angela Aguilar
The Kid Laroi
Lizzo
Fujii Kaze
Wisin & Yandel, ROSAL�
King
Coldplay, BTS
Sam Smith, Calvin Harris, Jessie Reyez
LE SSERAFIM
PinkPantheress
TAEYANG, Jimin
Rihanna
Lil Nas X
Beach Weather
Halsey, BTS
Baby Tate
Rïį¹̇́zï
Arijit Singh, Vishal Dadlani, Sukriti Kakar, Vishal-Shekhar, Shekhar
Ravjiani, Kumaar
Beach House
BTS, Jung Kook, FIFA Sound
Shakira, Rauw Alejandro
Steve Aoki, Tini, La Joaqui
Sam Smith
Conan Gray
Vishal-Shekhar, Shilpa Rao, Caralisa Monteiro, Kumaar, Vishal Dadlani,
Shekhar Ravjiani
Calvin Harris, Dua Lipa
TOMORROW X TOGETHER
Drake, Future, Tems
Mr.Kitty
Gustavo Mioto, Mari Fernandez
Jack Harlow, Lil Nas X
Elley Duhï¿
Seafret
Baby Rasta, Rauw Alejandro
DJ Escobar, MC MENOR SG, MC MENOR HR
Dean Lewis
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Bellakath
Skrillex, Flowdan, Fred again..
Travis Scott, 21 Savage, Metro Boomin
Kate Bush
Aerosmith
Freddie Dredd
Lost Frequencies, Calum Scott
Twisted, Oliver Tree
Kenshi Yonezu
Omar Apollo
Dua Lipa, DaBaby
Doja Cat
James Hype, Miggy Dela Rosa
Pritam, Arijit Singh, Amitabh Bhattacharya
Styrx, utku INC, Thezth
Maroon 5, Wiz Khalifa
Mariah Carey
Wham!
Brenda Lee
Bobby Helms
Michael Bubli;
Ariana Grande
Andy Williams
Dean Martin
Kelly Clarkson
Josi; \fi; Felic
The Ronettes
John Lennon, The Harlem Community Choir, The Plastic Ono Band, Yoko
0no
SZA, Don Toliver
Burl Ives
Nat King Cole
Paul McCartney
Band Aid
SZA, Phoebe Bridgers
Shakin' Stevens
Frank Sinatra, B. Swanson Quartet
SZA, Travis Scott
Bing Crosby, John Scott Trotter & His Orchestra, Ken Darby Singers
Chris Rea
Darlene Love
Ed Sheeran, Elton John
Perry Como, The Fontane Sisters, Mitchell Ayres & His Orchestra
Britney Spears
Mabel Matiz, Mert Demir
RM
Travis Scott, Metro Boomin
Taylor Swift, Lana Del Rey
Dj LK da Esc��cia, Tchakabum, mc jhenny, M
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Bad Bunny, Jhay Cortez Gwen Stefani, Blake Shelton Chencho Corleone, Rauw Alejandro Bad Bunny, Rauw Alejandro Chuck Berry Frank Sinatra John Legend, Metro Boomin Maldy, Karol G Gayle Lauren Spencer Smith, Lauren Spencer Smith, Lauren Spencer Smith Cherish, ACRAZE Adassa, Mauro Castillo, Stephanie Beatriz, Encanto - Cast, Rhenzy Feliz, Diane Guerrero, Carolina Gaitan Farruko Jaymes Young The Weeknd, Tyler, The Creator SZA, Doja Cat The Weeknd, Lil Wayne Ckay Jessica Darrow The Weeknd, Post Malone Mi; 1/5 i; 1/5 ne Aitana, zzoilo Julieta Venegas, Bad Bunny, Tainy Lisa Tii; ½i; ½sto, Ava The Weeknd, Swedish House Mafia Juan Cruz Toledo, Huilen Toledo BoyWithUke Xam��, Gustah, Neo B Lit Killah, Maria Becerra, Tiago pzk, NICKI NICOLE Justin Quiles, Lenny Taviį laiz rez, BL Shawn Mendes Kodak Black Aventura, Bad Bunny Rauw Alejandro Ckay, AX'EL, Dj Yo! Bruno Mars, Anderson .Paak, Silk Sonic THE ANXIETY, Willow, Tyler Cole Sebastian Yatra Justin Bieber, Daniel Caesar, Giveon Bizarrap, Tiago pzk Stromae Chris Brown, Ryssian, Rauw Alejandro NEIKED, Mae Muller, Polo G Oliver Tree Kali Uchis, Amaarae, Moliy

Young Thug, Future, Gunna

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Doja Cat, The Weeknd
Jnr Choi
Trueno, Tiago pzk
Sech, Bad Bunny, Mora
Drake, Project Pat, 21 Savage
Ti��sto, Kar
Edison Lighthouse
The Weeknd, ROSALi;
Olga Merediz, Stephanie Beatriz, Encanto - Cast
Surf Curse
Jason Derulo
21 Savage, Gunna
24kgoldn, Iann Dior
Stephanie Beatriz, Diane Guerrero
Trueno
Kanye West
Playboi Carti
Maluma
Ed Sheeran, Fireboy DML
Prezioso, Gabry Ponte, LUM!X
Tii; 1/5 i;
Mari; 1/5; 1/5 lia Mendoni; 1/5; 1/5a, George Henrique &
TWICE
Giveon
Sean Paul, Dua Lipa
Charli XCX, Jax Jones, Joel Corry, Saweetie
SALES
Sleepy hallow
Masked Wolf
Anuel Aa, Jhay Cortez
NIKI
Tate McRae
Avicii
Drake, Future, Young Thug
Imagine Dragons, League of Legends, Arcane
Tones and I
Juice WRLD
J Balvin, Maria Becerra
John Legend
Nirvana
Duncan Laurence
Drake, Travis Scott
Tini, L-Gante
SiM
Kendrick Lamar, Baby Keem
The Killers
J Balvin, Nio Garcia, Bad Bunny
Lady Gaga, Bradley Cooper
Marshmello, Jonas Brothers
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Riton, Nightcrawlers, Mufasa & Hypeman, Dopamine
Polo G
Eminem, Nate Dogg
Maroon 5
Arcangel, De La Ghetto, Justin Quiles, Lenny Tavi; 1/3; rez, Sech,
Dalex, Dimelo Flow, Rich Music
Gunna, Lil Baby
Katy Perry, Alesso
Mariį lia Mendoni; lia, Maiara &
Tini, Maria Becerra
Anuel Aa, Myke Towers, Jhay Cortez
Karol G, Becky G
Dr. Dre, Snoop Dogg
Dove Cameron
Ed Sheeran, Taylor Swift
Charlie Puth
Kanye West, Alicia Keys, Fivio Foreign
Mahmood, Blanco
50 Cent
Yung Lean
Labrinth, Zendaya
The Rare Occasions
Intense, AP Dhillon, Gurinder Gill
Quevedo, La Pantera, Juseph, Cruz Cafun��, B��jo, Abhir Hathi
Dr. Dre, 2Pac, Roger
Eminem, Dr. Dre
Luude, Colin Hay
Ryan Castro
Kendrick Lamar
Eminem, Dido
GODZZ -, Zakaria
Chase Atlantic
Em Beihold
Mainstreet, Chefin
Alvaro Diaz, Rauw Alejandro
Nicki Minaj, Lil Baby
Future
Z�� Fe
Muni Long
Vundabar
V
Mari; ½i; ½lia Mendoni; ½i; ½a, Hugo & G
Shawn Mendes, Camila Cabello
Daddy Yankee, Bad Bunny
Dua Lipa, Megan Thee Stallion
Camila Cabello, Ed Sheeran
Cris Mi
Anitta
Dave
Tulus
```

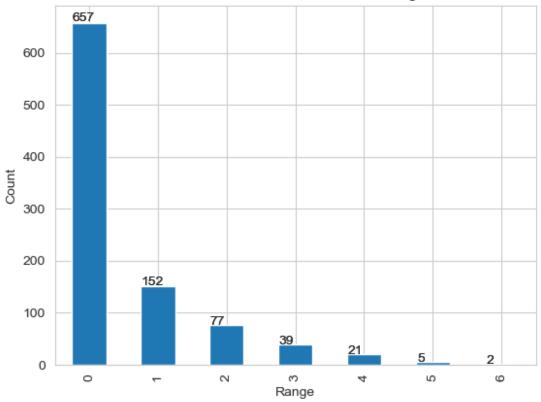
```
Mc Pedrinho, Pedro Sampaio
Zion & Lennox
Residente, Bizarrap
Jack Harlow
Eden Muï; 15 ï
Treasure
Christian Nodal
Kevin Gates, Juicy J
Tyga, Doja Cat
Queen
YEAT
C. Tangana
The Chainsmokers, Coldplay
The Chainsmokers, Halsey
NMIXX
Gotye, Kimbra
AnnenMayKantereit, Giant Rooks
Paulo Londra
Danny Ocean
Lil Tjay
Willow
Feid, Alejo, Robi
Demi Lovato
Camila Cabello, Willow
Matui; ½i; ½, Wiu,
De La Ghetto, Duki, Quevedo
Lil Baby
Ak4:20, Cris Mj, Pailita
Yahritza Y Su Esencia
Lu��sa Sonza, MC Frog, Dj Gabriel do Borel, Davi K
Ed Sheeran, J Balvin
teto
Blackbear, BoyWithUke
Pharrell Williams, Nile Rodgers, Daft Punk
Jordan Fisher, Josh Levi, Finneas O'Connell, 4*TOWN (From Disney and
Pixari; ½i; ½i; Turning Red), Topher Ngo, Grayson Vill
BIGBANG
Kendrick Lamar, Blxst, Amanda Reifer
Kendrick Lamar, Sampha
Post Malone, Roddy Ricch
Kendrick Lamar, Kodak Black
Tony Dize, Bad Bunny
Buscabulla, Bad Bunny
Kendrick Lamar, Taylour Paige
Kendrick Lamar, Sam Dew, Baby Keem
Bad Bunny, The Mari¿¹zi
Kendrick Lamar, Ghostface Killah, Summer Walker
Polima WestCoast, Pailita
Bad Bunny, Tainy
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Panic! At The Disco
Kendrick Lamar, Tanna Leone
PSY, Suga
Benson Boone
KALUSH
Luciano
Bizarrap, Paulo Londra
Shae Gill, Ali Sethi
Kendrick Lamar, Beth Gibbons
Sech, Mora
MNEK, Jax Jones
Ugly Dray, Tesla Jnr
Justin Bieber, Don Toliver
Jessi
Chanel
Leah Kate
Sam Ryder
HA SUNG WOON, Jimin
The Goo Goo Dolls
Kevin Kaarl
Suki Waterhouse
Lauren Spencer Smith
Pharrell Williams, Tyler, The Creator, 21 Savage
Calvin Harris, Dua Lipa, Young Thug
Nengo Flow, Anuel Aa, Chris Jedi, Chencho Corleone
Kanye West, XXXTENTACION
Musical Youth
Bizarrap, Villano Antillano
Halsey
BYOR, Imanbek
Sleepy hallow, 347aidan
Paloma Faith
Marshmello, Khalid
Ozuna, Tiago pzk
Childish Gambino
David Guetta, Ella Henderson, Becky Hill
Duki
Sidhu Moose Wala
Keisya Levronka
Victor Cibrian
Metallica
De La Ghetto, Feid, Polima WestCoast, Paloma Mami, Pailita
Yung Gravy
Lasso
Naveon
Sch��rze, DJ R
i-hope
Guns N' Roses
Burna Boy
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```
Kanye West, Lil Durk, Cardi B
L7nnon, DJ Biel do Furduncinho, Bianca
Anitta, Tini, Becky G
Nessa Barrett
LF System
Mc Vitin Da Igrejinha, MC Tairon, DJ Win
Snoop Dogg, BTS, Benny Blanco
Drake, DJ Khaled, Lil Baby
Luar La L
Ghost
Rex Orange County
Sofia Carson
Grupo Marca Registrada, Junior H
Calvin Harris, Halsey, Pharrell Williams, Justin Timberlake
Peso Pluma, Luis R Conriquez
Luis R Conriguez, La Adictiva
David Guetta, Shakira, Black Eyed Peas
Melody, Ana Castela, Dj Chris No Beat
Elton John, Britney Spears
Interplanetary Criminal, Eliza Rose
Armani White
0xlade
PnB Rock
Ovy On The Drums, Quevedo
Lil Yachty
Marshmello, Juice WRLD
Fran C, Polima WestCoast, Nickoog Clk, Pablito Pesadilla
Southstar
girl in red
Rosa Walton, Hallie Coggins
Luciano, Aitch, Bï; \frac{1}{2}
MC Ryan SP, Love Funk, Mc Paiva ZS
Drake, Travis Scott, 21 Savage
Ludwig Goransson, Foudeqush
Jin
Selena Gomez
Feid, Paulo Londra
Feid, Sech, Jhayco
Total number of artists: 645
Data['streams'] = pd.to numeric(Data['streams'],
errors='coerce').fillna(0).astype(int)
Data['in deezer playlists'] =
pd.to_numeric(Data['in_deezer_playlists'],
errors='coerce').fillna(0).astype(int)
Data['in shazam charts'] = pd.to numeric(Data['in shazam charts'],
errors='coerce').fillna(0).astype(int)
```

```
Data['key'] = pd.to numeric(Data['key'],
errors='coerce').fillna(0).astype(int)
Data['artist(s) name'] = pd.to numeric(Data['artist(s) name'],
errors='coerce').fillna(0).astype(int)
bins = 7
Data['streams range'] = pd.cut(Data['streams'], bins, labels=False)
max value = Data['streams'].max()
bins = 7
bin edges = np.linspace(0, max value, bins + 1)
Data['streams range'] = pd.cut(Data['streams'], bins=bin edges,
labels=False)
new bin edges = pd.cut(Data['streams'], bins=bin edges).cat.categories
for edge in new bin edges:
    print(f"Bin range: {edge.left} to {edge.right}")
Bin range: 0.0 to 529127867.714
Bin range: 529127867.714 to 1058255735.429
Bin range: 1058255735.429 to 1587383603.143
Bin range: 1587383603.143 to 2116511470.857
Bin range: 2116511470.857 to 2645639338.571
Bin range: 2645639338.571 to 3174767206.286
Bin range: 3174767206.286 to 3703895074.0
#std dev = Data['streams'].std()
#bin edges = np.arange(Data['streams'].min(), Data['streams'].max() +
std dev, std dev)
# Create a new column with bin labels
#labels = ['streams_sd_0', 'streams_sd_1', 'streams_sd_2',
'streams_sd_3', 'streams_sd_4', 'streams_sd_5', 'streams_sd_6']
\#labels = list(range(7, -1, -1))
#Data['streams range'] = pd.cut(Data['streams'], bins=bin edges,
labels=False)
# Count the occurrences in each bin
#bin counts = Data['streams range'].value counts().sort index()
Data['streams range'].value counts().plot(kind='bar')
ax =
Data['streams_range'].value_counts().sort_index().plot(kind='bar')
for p in ax.patches:
    ax.annotate(str(p.get height()), (p.get x() * 1.005,
p.get height() * 1.005))
plt.xlabel('Range')
plt.vlabel('Count')
plt.title('Distribution of Streams Across range')
plt.figure(figsize=(20,15))
plt.show()
```





```
<Figure size 2000x1500 with 0 Axes>
Data.info()
Data = Data.dropna(axis=1, how='all')
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 953 entries, 0 to 952
Data columns (total 25 columns):
#
     Column
                            Non-Null Count
                                            Dtype
     -----
0
     track name
                            953 non-null
                                            object
1
     artist(s) name
                            953 non-null
                                            int64
 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
                            953 non-null
     in spotify charts
                                            int64
 8
     streams
                            953 non-null
                                            int64
 9
     in apple playlists
                            953 non-null
                                            int64
    in apple charts
 10
                            953 non-null
                                            int64
     in deezer playlists
                            953 non-null
 11
                                            int64
     in deezer charts
 12
                            953 non-null
                                            int64
 13
     in shazam charts
                            953 non-null
                                            int64
```

```
14 bpm
                              953 non-null
                                                 int64
 15 key
                              953 non-null
                                                 int64
 16 mode
                              953 non-null
                                                 int64
 17 danceability %
                              953 non-null
                                                 int64
 18 valence %
                              953 non-null
                                                int64
 19 energy_%
                              953 non-null
                                                int64
                              953 non-null
953 non-null
953 non-null
953 non-null
953 non-null
 20 acousticness_% 953 non-null 21 instrumentalness_% 953 non-null
                                                int64
                                                int64
                                                int64
 22 liveness %
 23 speechiness %
                                                int64
 24 streams range
                                                int64
dtypes: int64(24), object(1)
memory usage: 186.3+ KB
Data = Data.drop(['track name'], axis=1)
# Backward selection
X = Data.drop(['streams', 'streams_range'], axis=1)
y = Data['streams range']
X = sm.add constant(X)
p_values_dict = {}
while True:
    model = sm.OLS(y, X).fit()
    p values = model.pvalues.drop('const')
    max_p_value = p_values.max()
    if max p value > 0.05:
         feature to remove = p values.idxmax()
         p values dict[feature to remove] = p values[feature to remove]
         X = X.drop(feature to remove, axis=1)
    else:
         break
selected features = X.columns.tolist()
print("Selected Features:", selected features)
print("Columns and their p-values:")
for col in p values dict:
    print(f"{col}: {p values dict[col]}")
Selected Features: ['const', 'artist(s)_name', 'released_year',
'in_spotify_playlists', 'in_spotify_charts', 'in_apple_playlists',
'in_deezer_playlists', 'in_deezer_charts', 'in_shazam_charts', 'key',
'valence_%', 'acousticness_%']
Columns and their p-values:
liveness %: 0.992686023541067
mode: 0.\overline{9}204383445605774
```

```
released month: 0.8552747485618422
bpm: 0.6687624171216515
speechiness %: 0.5860408462133606
energy %: 0.5368390339044422
danceability %: 0.4936508719392937
instrumentalness_%: 0.22882560800703433
released day: 0.\overline{2}4235413468782666
in apple charts: 0.148316368842711
artist count: 0.058450810034350834
#KNN without Backward selection
X = Data.drop(['streams', 'streams range'], axis=1)
y = Data['streams range'].values
scaler = StandardScaler()
X scaled = scaler.fit transform(X)
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y,
test size=0.3, random state=0)
classifier = KNeighborsClassifier(n neighbors=5)
classifier.fit(X_train, y_train)
y pred = classifier.predict(X test)
accuracy = accuracy_score(y_test, y_pred)
conf matrix = confusion matrix(y test, y pred)
print(f'Accuracy: {accuracy * 100}%')
print('Confusion Matrix:\n', conf matrix)
Accuracy: 74.82517482517483%
Confusion Matrix:
 [[196
       4
           0
                 0
                             01
                    0
 [ 35
        9
            4
                0
                        0
                            01
                    0
            5
   9
        4
                1
                    0
                        0
                            01
       3 2 3
                        0
   1
                  1
                            0]
      1 1 2
   1
                  1
                        0
                            01
            0
                0
                    2
   0
        0
                        0
                            01
                0
                    1
        0
                        0
                            011
# KNN with Backward selection
import numpy as np
from sklearn.model selection import train test split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score, confusion_matrix
feature columns = ['artist(s) name', 'artist count', 'released year',
'in_spotify_playlists', 'in_spotify_charts',
                   'in_apple_playlists', 'in_deezer_playlists',
                   'in_deezer_charts', 'in_shazam_charts', 'key',
'valence %', 'acousticness %']
X = Data[feature columns].values
```

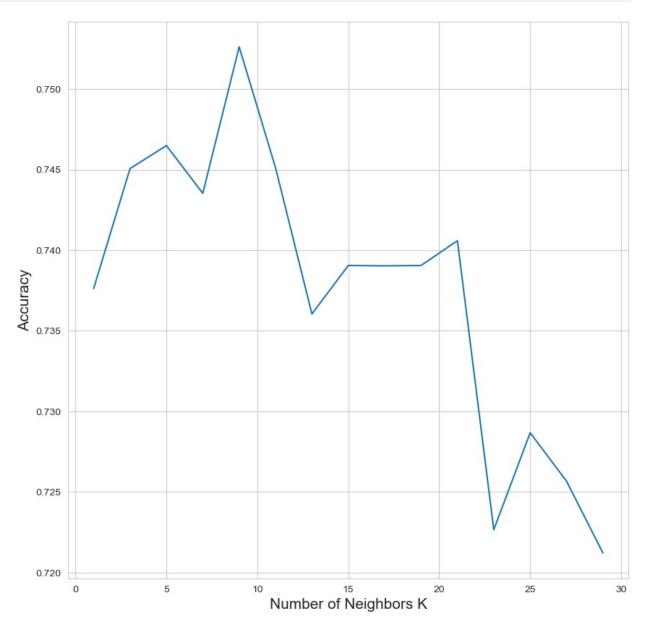
```
v = Data['streams range'].values
scaler = StandardScaler()
X scaled = scaler.fit transform(X)
X train, X test, y train, y test = train test split(X scaled, y,
test size=0.3, random state=0)
classifier = KNeighborsClassifier(n neighbors=5)
classifier.fit(X train, y train)
y pred = classifier.predict(X test)
accuracy = accuracy score(y test, y pred)
conf matrix = confusion matrix(y test, y pred)
print(f'Accuracy: {accuracy * 100}%')
print('Confusion Matrix:\n', conf_matrix)
Accuracy: 76.57342657342657%
Confusion Matrix:
 [[197
        3
           0
                 0
                     0
                         0
                             01
 [ 34
      10
            4
                0
                    0
                        0
                            01
           5
               2
                            01
    4
        8
                    0
                        0
        2
           2
    1
               5
                    0
                        0
                            01
    0
        0
           3 1
                    2
                        0
                            0]
        0
            0
                        0
    0
                1
                    1
                            01
    0
        0
            0
                0
                    1
                        0
                            011
k list = list(range(1,30,2))
cv scores = []
for k in k list:
    knn = KNeighborsClassifier(n neighbors=k)
    scores = cross val score(knn, X train, y train, cv=10,
scoring='accuracy')
    cv scores.append(scores.mean())
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/
sklearn/model selection/ split.py:725: UserWarning: The least
populated class in y has only 1 members, which is less than
n splits=10.
  warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model_selection/_split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
 warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
 warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
 warnings.warn(
/Users/marco9951qg.com/anaconda3/lib/python3.11/site-packages/sklearn/
```

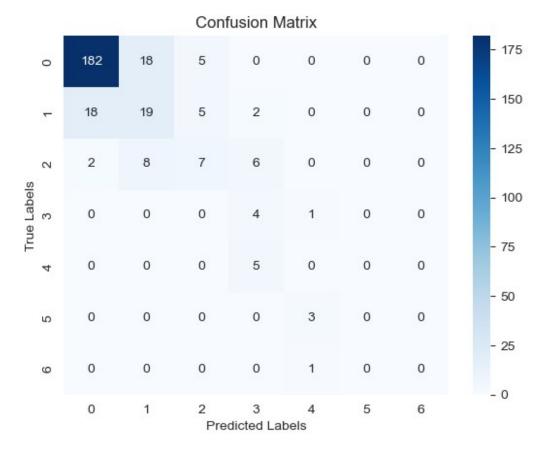
```
model selection/ split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
 warnings.warn(
/Users/marco9951qg.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
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 warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
  warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
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  warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
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/Users/marco9951qg.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
 warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
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/Users/marco9951gq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
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/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
 warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
  warnings.warn(
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
model selection/ split.py:725: UserWarning: The least populated class
in y has only 1 members, which is less than n splits=10.
 warnings.warn(
plt.figure()
plt.figure(figsize=(10,10))
plt.xlabel('Number of Neighbors K', fontsize=15)
plt.ylabel('Accuracy', fontsize=15)
sns.set style("whitegrid")
```

```
plt.plot(k_list, cv_scores)
plt.show()

conf_matrix = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix')
plt.xlabel('Predicted Labels')
plt.ylabel('True Labels')
plt.show()

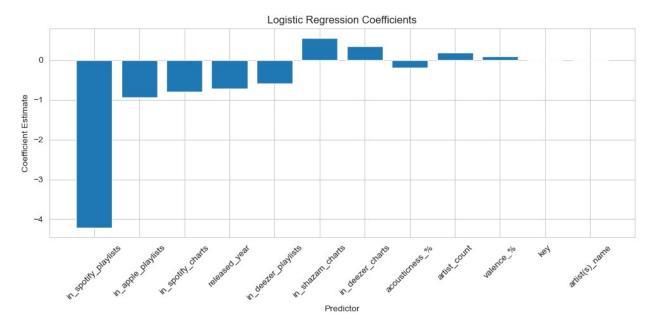
<Figure size 640x480 with 0 Axes>
```





```
#Logistic Regression
from sklearn.metrics import confusion matrix, accuracy score
from sklearn.preprocessing import StandardScaler
from sklearn.linear model import LogisticRegression
#feature columns = ['artist(s) name', 'artist count', 'released year',
'in spotify playlists', 'in spotify charts',
                   #'in apple playlists', 'in deezer playlists',
                   #'in_deezer_charts', 'in_shazam_charts', 'key',
'valence %', 'acousticness %']
X = Data.drop(['streams', 'streams_range'], axis=1)
#X = Data[feature columns].values
y = Data['streams range']
scaler = StandardScaler()
X train scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
logreg = LogisticRegression(random state=16, max iter=1000)
logreg.fit(X train scaled, y train)
y pred = logreg.predict(X test scaled)
conf matrix = confusion matrix(y test, y pred)
```

```
print(conf matrix)
accuracy = accuracy score(y test, y pred)
print(accuracy)
[[197
      8
                   0
                          0]
                      0
[ 24
     14
           6
               0
                      0
                          01
   2 12 7
               2
                 0
                      0
                          0]
   0 0 0 4 1 0 0]
       0
             2
                  2 0
   0
          1
                          01
       0
           0 0
                  3
                      0
  0
                          0]
               0
  0
       0
           0
                 1
                      0
                          0]]
0.7832167832167832
import matplotlib.pyplot as plt
import numpy as np
# Assuming logreg is your trained logistic regression model
coefficients = logreg.coef_[0] # Get the coefficients from the model
features = feature columns # List of feature names
indices = np.argsort(np.abs(coefficients))[::-1] # Sort by magnitude
# Create a bar chart
plt.figure(figsize=(10, 5))
plt.title('Logistic Regression Coefficients')
plt.bar(range(len(indices)), coefficients[indices], align='center')
plt.xticks(range(len(indices)), np.array(features)[indices],
rotation=45)
plt.xlabel('Predictor')
plt.ylabel('Coefficient Estimate')
plt.tight layout()
plt.show()
```



```
cm = confusion_matrix(y_test, y_pred)
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')
plt.title('Confusion Matrix')
plt.xlabel('Predicted Labels')
plt.ylabel('True Labels')
plt.show()
```

Confusion Matrix								
0	196	3	1	0	0	0	0	- 175
-	24	21	2	1	0	0	0	- 150
s 2	4	8	4	3	0	0	0	- 125
True Labels 3	0	1	2	4	3	0	0	- 100
T 4	0	0	1	0	3	2	0	- 75
rs.	0	0	0	0	2	0	0	- 50
9	0	0	0	0	1	0	0	- 25
	0	1	2 Pred	3 dicted Lai	4 bels	5	6	- 0

```
#SVM
clf = svm.SVC()
clf.fit(X_train, y_train)
y_pred = clf.predict(X_test)
print(confusion_matrix(y_test,y_pred))
print(classification_report(y_test,y_pred))
[[194
       6
             0
                 0
                     0
                          0
                              0]
             5
 [ 26
       17
                 0
                     0
                          0
                              0]
    3
       11
             5
                          0
                              0]
                 0
                     0
    1
        2
             3
                 4
                     0
                          0
                              0]
             3
                 0
                     3
    0
        0
                          0
                              0]
             1
                     1
    0
        0
                 0
                          0
                              0]
        0
                 0
                     0
                          0
                              0]]
               precision
                             recall
                                     f1-score
                                                  support
                               0.97
                                                      200
            0
                    0.86
                                          0.91
                               0.35
            1
                    0.47
                                          0.40
                                                       48
            2
                    0.29
                                                       19
                               0.26
                                          0.28
            3
                    1.00
                               0.40
                                          0.57
                                                       10
            4
                    0.75
                               0.50
                                          0.60
                                                        6
            5
                                                        2
                    0.00
                               0.00
                                          0.00
```

```
0.00
                             0.00
                                       0.00
                                                     1
                                                   286
    accuracy
                                       0.78
                             0.36
                                       0.40
   macro avg
                   0.48
                                                   286
weighted avg
                   0.75
                             0.78
                                       0.76
                                                   286
/Users/marco9951gg.com/anaconda3/lib/python3.11/site-packages/
sklearn/metrics/ classification.py:1469: UndefinedMetricWarning:
Precision and F-score are ill-defined and being set to 0.0 in labels
with no predicted samples. Use `zero division` parameter to control
this behavior.
  warn prf(average, modifier, msg_start, len(result))
/Users/marco9951gg.com/anaconda3/lib/python3.11/site-packages/sklearn/
metrics/ classification.py:1469: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero division` parameter to control this
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  warn prf(average, modifier, msg start, len(result))
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
metrics/ classification.py:1469: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero division` parameter to control this
behavior.
  warn prf(average, modifier, msg start, len(result))
y pred = clf.predict(X test)
cm = confusion matrix(y test, y pred)
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', cbar=False)
plt.ylabel('Actual Label')
plt.xlabel('Predicted Label')
plt.title('SVM Confusion Matrix')
accuracy = np.trace(cm) / np.sum(cm).astype('float')
plt.figtext(0.5, 0.01, f'Accuracy: {accuracy:.2f}', ha='center',
fontsize=12)
plt.show()
print(classification report(y_test, y_pred))
```

SVM Confusion Matrix

0	194	6	0	0	0	0	0
-	26	17	5	0	0	0	0
2	3	11	5	0	0	0	0
Actual Label 3	1	2	3	4	0	0	0
4	0	0	3	0	3	0	0
2	0	0	1	0	1	0	0
9	1	0	0	0	0	0	0
	0	1	2	3	4	5	6

Predicted Label Accuracy: 0.78

	precision	recall	f1-score	support
0	0.86	0.97	0.91	200
1	0.47	0.35	0.40	48
2	0.29	0.26	0.28	19
3	1.00	0.40	0.57	10
4	0.75	0.50	0.60	6
5	0.00	0.00	0.00	2
6	0.00	0.00	0.00	1
accuracy			0.78	286
macro avg	0.48	0.36	0.40	286
weighted avg	0.75	0.78	0.76	286

/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/metrics/\_classification.py:1469: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control

```
this behavior.
  warn prf(average, modifier, msg start, len(result))
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
metrics/ classification.py:1469: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this
  warn prf(average, modifier, msg start, len(result))
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
metrics/ classification.py:1469: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero_division` parameter to control this
behavior.
 warn prf(average, modifier, msg start, len(result))
#Random Forest
from sklearn.ensemble import RandomForestClassifier
feature_columns = ['artist(s)_name', 'artist_count', 'released_year',
'in_spotify_playlists', 'in_spotify_charts',
                   'in_apple_playlists', 'in_deezer_playlists',
                   'in_deezer_charts', 'in_shazam_charts', 'key',
'valence %', 'acousticness %']
X = Data[feature columns].values
y = Data['streams range'].values
X_train, X_test, y_train, y_test = train_test_split(X, y,
test size=0.3, random state=42)
rf = RandomForestClassifier(n estimators=100, random state=42)
rf.fit(X train, y train)
rf predictions = rf.predict(X test)
rf_accuracy = accuracy_score(y_test, rf_predictions)
print(f"Random Forest Accuracy: {rf accuracy}")
Random Forest Accuracy: 0.8041958041958042
#Random Forest with Cross validation
from sklearn.model_selection import train_test_split, cross_val_score
cv scores = cross val score(rf, X, y, cv=5)
print(f"Cross-Validation Scores: {cv scores}")
print(f"Average CV Score: {cv scores.mean()}")
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/
sklearn/model selection/ split.py:725: UserWarning: The least
```

```
populated class in y has only 2 members, which is less than
n splits=5.
 warnings.warn(
Cross-Validation Scores: [0.80628272 0.83246073 0.65968586 0.76315789
0.773684211
Average CV Score: 0.7670542849269771
#Feature Importance
import numpy as np
importances = rf.feature importances
indices = np.argsort(importances)
plt.title('Feature Importances')
plt.barh(range(len(indices)), importances[indices], color='b',
align='center')
plt.yticks(range(len(indices)), [Data.columns[i] for i in indices])
plt.xlabel('Relative Importance')
plt.show()
monthly streams = Data.groupby('released month')['streams'].mean()
best month = monthly streams.idxmax()
print(f"The best month for release, based on average streams, is:
{best month}")
The best month for release, based on average streams, is: 9
#Naive Bayes
from sklearn.naive bayes import GaussianNB
model = GaussianNB()
X train, X test, y train, y test = train test split(X, y,
test size=0.3, random state=42)
model.fit(X train, y train);
from sklearn.metrics import (
    accuracy score,
    confusion matrix,
    ConfusionMatrixDisplay,
    fl score,
    classification report,
)
y pred = model.predict(X test)
accuracy = accuracy_score(y_pred, y_test)
f1 = f1 score(y pred, y test, average="weighted")
```

```
print("Accuracy:", accuray)
print("F1 Score:", f1)

Accuracy: 0.7412587412587412
F1 Score: 0.7412719510012666

cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', cbar=False)
plt.ylabel('Actual Label')
plt.xlabel('Predicted Label')
plt.title('Bayes Confusion Matrix')
accuracy = np.trace(cm) / np.sum(cm).astype('float')
plt.figtext(0.5, 0.01, f'Accuracy: {accuracy:.2f}', ha='center', fontsize=12)
plt.show()
print(classification_report(y_test, y_pred))
```

Bayes Confusion Matrix								
0	182	18	5	0	0	0	0	
-	18	19	5	2	0	0	0	
2	2	8	7	6	0	0	0	
Actual Label 3	0	0	0	4	1	0	0	
4	0	0	0	5	0	0	0	
2	0	0	0	0	3	0	0	
9	0	0	0	0	1	0	0	
	0	1	2 F	3 Predicted Label	4	5	6	

Accuracy: 0.74

	precision	recall	f1-score	support
0	0.90	0.89	0.89	205
1	0.42	0.43	0.43	44
2	0.41	0.30	0.35	23
3 4	0.24	0.80	0.36	5
	0.00	0.00	0.00	5
5	0.00	0.00	0.00	3
6	0.00	0.00	0.00	
accuracy	0100	0.00	0.74	286
macro avg	0.28	0.35	0.29	286
weighted avg	0.75	0.74	0.74	286

/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/metrics/\_classification.py:1469: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
metrics/\_classification.py:1469: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero\_division` parameter to control this
behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))
/Users/marco9951qq.com/anaconda3/lib/python3.11/site-packages/sklearn/
metrics/\_classification.py:1469: UndefinedMetricWarning: Precision and
F-score are ill-defined and being set to 0.0 in labels with no
predicted samples. Use `zero\_division` parameter to control this
behavior.

warn prf(average, modifier, msg start, len(result))

```
# labels = [ "streams_sd 1", "streams_sd 2", "streams_sd 3", "streams_sd
4", "streams_sd 5", "streams_sd 6", "streams_sd 7"]
# cm = confusion_matrix(y_test, y_pred)
# disp =
ConfusionMatrixDisplay(confusion_matrix=cm , display_labels=labels)
# disp.plot()
# plt.xticks(rotation=45)
```