```
In [1]: M import pandas as pd
import numpy as np
import warnings
warnings.filterwarnings("ignore")
import seaborn as sns
```

In [2]:

df=pd.read_csv('spotify_songs.csv')
df

Out[2]:

	track_id	track_name	track_artist	track_popularity	track_album_id	track_albı
0	6f807x0ima9a1j3VPbc7VN	I Don't Care (with Justin Bieber) - Loud Luxur	Ed Sheeran	66	2oCs0DGTsRO98Gh5ZSl2Cx	I Don't Justin Bie
1	0r7CVbZTWZgbTCYdfa2P31	Memories - Dillon Francis Remix	Maroon 5	67	63rPSO264uRjW1X5E6cWv6	Memor Franc
2	1z1Hg7Vb0AhHDiEmnDE79I	All the Time - Don Diablo Remix	Zara Larsson	70	1HoSmj2eLcsrR0vE9gThr4	All the [·] Diak
3	75FpbthrwQmzHlBJLuGdC7	Call You Mine - Keanu Silva Remix	The Chainsmokers	60	1nqYsOef1yKKuGOVchbsk6	Call You I
4	1e8PAfcKUYoKkxPhrHqw4x	Someone You Loved - Future Humans Remix	Lewis Capaldi	69	7m7vv9wlQ4i0LFuJiE2zsQ	Son Lov Huma
32828	7bxnKAamR3snQ1VGLuVfC1	City Of Lights - Official Radio Edit	Lush & Simon	42	2azRoBBWEEEYhqV6sb7JrT	City Of Lig
32829	5Aevni09Em4575077nkWHz	Closer - Sultan & Ned Shepard Remix	Tegan and Sara	20	6kD6KLxj7s8eCE3ABvAyf5	Close
32830	7ImMqPP3Q1yfUHvsdn7wEo	Sweet Surrender - Radio Edit	Starkillers	14	0ltWNSY9JgxolZO4VzuCa6	Sweet (F
32831	2m69mhnfQ1Oq6lGtXuYhgX	Only For You - Maor Levi Remix	Mat Zo	15	1fGrOkHnHJcStl14zNx8Jy	On
32832	29zWqhca3zt5NsckZqDf6c	Typhoon - Original Mix	Julian Calor	27	0X3mUOm6MhxR7PzxG95rAo	Typh
32833	rows × 23 columns					
4						•

```
    df.isnull().sum()

In [3]:
   Out[3]: track id
                                       5
            track name
            track artist
                                       5
                                       0
            track_popularity
                                       0
            track_album_id
            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
                                       0
            key
            loudness
                                       0
            mode
                                       0
            speechiness
                                       0
            acousticness
                                       0
            instrumentalness
                                       0
                                       0
            liveness
            valence
                                       a
            tempo
                                       0
            duration_ms
            dtype: int64

    df.info()
In [4]:
            <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
                track_album_release_date 32833 non-null object
            7
                                          32833 non-null object
                playlist name
            8
                playlist id
                                          32833 non-null object
                playlist_genre
                                          32833 non-null object
            10 playlist_subgenre
                                          32833 non-null object
                                          32833 non-null float64
            11 danceability
            12 energy
                                          32833 non-null float64
                                          32833 non-null int64
            13 key
                                          32833 non-null float64
            14
                loudness
                                          32833 non-null int64
            15
                mode
                                          32833 non-null float64
            16 speechiness
                                          32833 non-null float64
            17
                acousticness
                                          32833 non-null float64
            18
                instrumentalness
            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 [5]: M df.drop(['track_id','track_name','track_artist','track_album_id','track_album_name',
                                     'playlist_id','playlist_name','playlist_id','playlist_subgenre'],inpla
```

```
In [6]: ► df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32833 entries, 0 to 32832
Data columns (total 15 columns):
# Column Non-Nul
```

#	Column	Non-Null Count	Dtype	
0	track_popularity	32833 non-null	int64	
1	<pre>track_album_release_date</pre>	32833 non-null	object	
2	playlist_genre	32833 non-null	object	
3	danceability	32833 non-null	float64	
4	energy	32833 non-null	float64	
5	key	32833 non-null	int64	
6	loudness	32833 non-null	float64	
7	mode	32833 non-null	int64	
8	speechiness	32833 non-null	float64	
9	acousticness	32833 non-null	float64	
10	instrumentalness	32833 non-null	float64	
11	liveness	32833 non-null	float64	
12	valence	32833 non-null	float64	
13	tempo	32833 non-null	float64	
14	duration_ms	32833 non-null	int64	
d+vn	oc: $flor+64(0)$ in+64(4)	object(2)		

dtypes: float64(9), int64(4), object(2)

memory usage: 3.8+ MB

Out[7]:

	track_popularity	track_album_release_date	danceability	energy	key	loudness	mode	speecl
playlist_genre								
edm	210499	2013-09-162015-10- 232018-08-242016-12- 262019-0	3958.4120	4849.362000	32343	-32798.051	3143	52
latin	242422	2017-11-242019-03- 082017-07-212019-10- 182019-0	3676.9958	3651.350775	28270	-32293.264	2897	52
рор	262931	2019-06-142019-12- 132019-07-052019-07- 192019-0	3520.6345	3860.560340	29291	-34778.512	3239	40
r&b	223885	2020-01-102019-11- 132019-12-272018-06- 032019-1	3639.7440	3209.364400	29330	-42713.989	2832	63
rap	248316	2020-01-102019-12- 062020-01-172019-12- 122020-0	4127.6550	3738.970500	31436	-40464.880	2996	113
rock	206597	1987-01-012018-05- 041987-01-011987-01- 011987-0	2577.2330	3628.158500	25790	-37572.619	3467	28
								•

In []: ▶

```
In [8]:
           df.info()
              <class 'pandas.core.frame.DataFrame'>
              RangeIndex: 32833 entries, 0 to 32832
              Data columns (total 15 columns):
                   Column
               #
                                                Non-Null Count Dtype
              - - -
               0
                   track_popularity
                                                32833 non-null int64
                    track_album_release_date 32833 non-null object
               1
               2
                    playlist_genre
                                                32833 non-null
                                                                 object
               3
                    danceability
                                                32833 non-null
                                                                 float64
               4
                    energy
                                                32833 non-null float64
               5
                    key
                                                32833 non-null int64
               6
                    loudness
                                                32833 non-null
                                                                 float64
               7
                    mode
                                                32833 non-null
                                                                  int64
               8
                    speechiness
                                                32833 non-null
                                                                 float64
               9
                    acousticness
                                                32833 non-null
                                                                 float64
               10
                   instrumentalness
                                                32833 non-null
                                                                 float64
               11
                   liveness
                                                32833 non-null
                                                                 float64
               12
                   valence
                                                32833 non-null
                                                                 float64
                   tempo
               13
                                                32833 non-null float64
                                                32833 non-null int64
               14 duration_ms
              dtypes: float64(9), int64(4), object(2)
              memory usage: 3.8+ MB

    df.groupby(['track_album_release_date']).sum()

 In [9]:
     Out[9]:
                                      track_popularity
                                                                                    playlist_genre danceability energy ke
               track_album_release_date
                           1957-01-01
                                                 59
                                                                                                              0.962
                                                                                             r&b
                                                                                                       0.565
                                                                                                       0.474
                              1957-03
                                                  1
                                                                                                              0.598
                                                                                            rock
                            1958-03-21
                                                  73
                                                                                                       0.647
                                                                                                              0.582
                                                                                                              1.859
                                 1960
                                                 64
                                                                                     r&br&br&br&b
                                                                                                       1.545
                           1961-10-26
                                                 47
                                                                                                       0.585
                                                                                                              0.036
                                                                                             r&b
                           2020-01-15
                                                918
                                                     poppoppoppoppoppoppoppopraprapraprapraprapr...
                                                                                                      18.724
                                                                                                             19.891
                                                                                                                   17
                           2020-01-16
                                                1132
                                                         poppoppoppopraprapraprapraprapraplatinlatinlat...
                                                                                                      15.298
                                                                                                             13.928
                           2020-01-17
                                                4736
                                                     poppoppoppoppoppoppoppopraprapraprapraprara...
                                                                                                      86.589
                                                                                                             93.460
                                                                                                                    70
                           2020-01-20
                                                 82
                                                                                                              1.324
                                                                                         popedm
                                                                                                       1 4 1 6
                           2020-01-29
                                                 29
                                                                                                              0.880
                                                                                                       0.695
                                                                                             pop
              4530 rows × 14 columns
In [10]:

    def extract year(date str):

                  try:
                       return pd.to_datetime(date_str, format='%Y-%m-%d', errors='coerce').year
                  except:
                       return pd.to_numeric(date_str, errors='coerce')
              df['track_album_release_date'] = df['track_album_release_date'].apply(extract_year)
```

```
In [11]: ▶ df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32833 entries, 0 to 32832
Data columns (total 15 columns):

#	Column	Non-Null Count	Dtype
0	track_popularity	32833 non-null	int64
1	<pre>track_album_release_date</pre>	30947 non-null	float64
2	playlist_genre	32833 non-null	object
3	danceability	32833 non-null	float64
4	energy	32833 non-null	float64
5	key	32833 non-null	int64
6	loudness	32833 non-null	float64
7	mode	32833 non-null	int64
8	speechiness	32833 non-null	float64
9	acousticness	32833 non-null	float64
10	instrumentalness	32833 non-null	float64
11	liveness	32833 non-null	float64
12	valence	32833 non-null	float64
13	tempo	32833 non-null	float64
14	duration_ms	32833 non-null	int64
J.L	C1+C4/40\+C4/4\	-1-44/4	

dtypes: float64(10), int64(4), object(1)

memory usage: 3.8+ MB

In [12]: ► df.head(5)

Out[12]:

	track_popularity	track_album_release_date	playlist_genre	danceability	energy	key	loudness	mode	speechiness
0	66	2019.0	рор	0.748	0.916	6	-2.634	1	0.0583
1	67	2019.0	рор	0.726	0.815	11	-4.969	1	0.0373
2	70	2019.0	рор	0.675	0.931	1	-3.432	0	0.0742
3	60	2019.0	рор	0.718	0.930	7	-3.778	1	0.1020
4	69	2019.0	рор	0.650	0.833	1	-4.672	1	0.0359
4									•

In [13]: ▶ df.tail(5)

Out[13]:

	track_popularity	track_album_release_date	playlist_genre	danceability	energy	key	loudness	mode	speechi
32828	42	2014.0	edm	0.428	0.922	2	-1.814	1	0.
32829	20	2013.0	edm	0.522	0.786	0	-4.462	1	0.
32830	14	2014.0	edm	0.529	0.821	6	-4.899	0	0.
32831	15	2014.0	edm	0.626	0.888	2	-3.361	1	0.
32832	27	2014.0	edm	0.603	0.884	5	-4.571	0	0.
4									•

```
In [14]:

    df.info()

             <class 'pandas.core.frame.DataFrame'>
             RangeIndex: 32833 entries, 0 to 32832
             Data columns (total 15 columns):
              #
                Column
                                            Non-Null Count Dtype
             ---
                                            -----
              0
                  track_popularity
                                            32833 non-null int64
              1
                  track_album_release_date 30947 non-null float64
              2
                  playlist_genre
                                            32833 non-null object
                                            32833 non-null float64
              3
                  danceability
                                            32833 non-null float64
              4
                  energy
                                            32833 non-null int64
              5
                  key
                                            32833 non-null float64
              6
                  loudness
              7
                  mode
                                            32833 non-null int64
              8
                  speechiness
                                            32833 non-null float64
              9
                  acousticness
                                            32833 non-null float64
              10
                  instrumentalness
                                            32833 non-null float64
                                            32833 non-null float64
              11
                  liveness
              12 valence
                                            32833 non-null float64
              13 tempo
                                            32833 non-null float64
              14 duration_ms
                                            32833 non-null int64
             dtypes: float64(10), int64(4), object(1)
             memory usage: 3.8+ MB
In [15]:

    df.isnull().sum()

   Out[15]: track popularity
                                            0
             track album release date
                                         1886
             playlist_genre
                                            0
             danceability
                                            0
                                            0
             energy
                                            0
             key
                                            0
             loudness
                                            0
             mode
                                            0
             speechiness
                                            0
             acousticness
                                            0
             instrumentalness
                                            0
             liveness
             valence
                                            0
             tempo
                                            0
             duration ms
                                            0
             dtype: int64
In [16]:

    df.min()

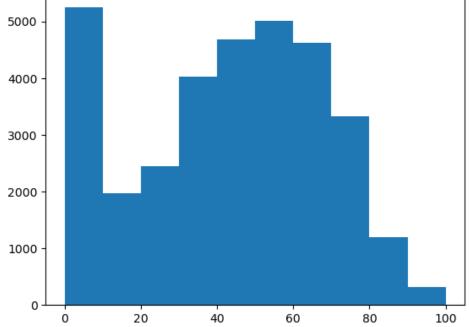
   Out[16]: track_popularity
                                                0
                                           1957.0
             track_album_release_date
             playlist_genre
                                               edm
             danceability
                                               0.0
             energy
                                         0.000175
                                                0
             key
             loudness
                                          -46.448
             mode
                                                0
             speechiness
                                              0.0
             acousticness
                                              0.0
             instrumentalness
                                              0.0
             liveness
                                              0.0
             valence
                                              0.0
             tempo
                                              0.0
                                             4000
             duration_ms
             dtype: object
```

```
▶ df.max()
In [17]:
    Out[17]: track_popularity
                                                 100
              track_album_release_date
                                             2020.0
              playlist_genre
                                                rock
              danceability
                                               0.983
              energy
                                                 1.0
              key
                                                  11
              loudness
                                               1.275
              mode
                                                   1
                                               0.918
              speechiness
              acousticness
                                               0.994
              instrumentalness
                                               0.994
              liveness
                                               0.996
              valence
                                               0.991
              tempo
                                             239.44
              duration ms
                                             517810
              dtype: object
 In [ ]:
           M df['track_album_release_date'] = df['track_album_release_date'].fillna(2012)
In [18]:
In [19]:

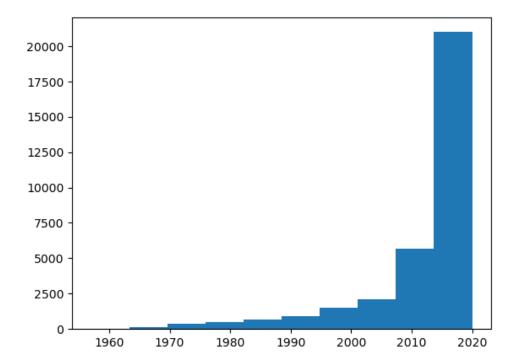
    df.isnull().sum()

    Out[19]: track_popularity
                                             0
              track_album_release_date
                                             0
              playlist_genre
                                             0
                                             0
              danceability
                                             0
              energy
                                             0
              key
              loudness
                                             0
              mode
                                             0
              speechiness
                                             0
              acousticness
                                             0
              instrumentalness
                                             0
              liveness
                                             0
              valence
                                             0
              tempo
                                             0
              duration_ms
                                             0
              dtype: int64
In [20]:
           df.head(5)
    Out[20]:
                  track_popularity track_album_release_date playlist_genre danceability energy
                                                                                             loudness mode speechiness
                                                                                         key
               0
                             66
                                                 2019.0
                                                                           0.748
                                                                                   0.916
                                                                                           6
                                                                                                -2.634
                                                                                                                   0.0583
                                                                 pop
               1
                             67
                                                 2019.0
                                                                           0.726
                                                                                   0.815
                                                                                          11
                                                                                                -4.969
                                                                                                           1
                                                                                                                   0.0373
                                                                 pop
               2
                             70
                                                 2019.0
                                                                           0.675
                                                                                   0.931
                                                                                                           0
                                                                                                                   0.0742
                                                                                           1
                                                                                                -3.432
                                                                 pop
               3
                                                                                           7
                                                 2019.0
                                                                           0.718
                                                                                   0.930
                                                                                                                   0.1020
                             60
                                                                 pop
                                                                                                -3.778
                                                                                                           1
                             69
                                                 2019.0
                                                                            0.650
                                                                                   0.833
                                                                                                -4.672
                                                                                                                   0.0359
                                                                 pop
                                                                                           1
                                                                                                           1
```

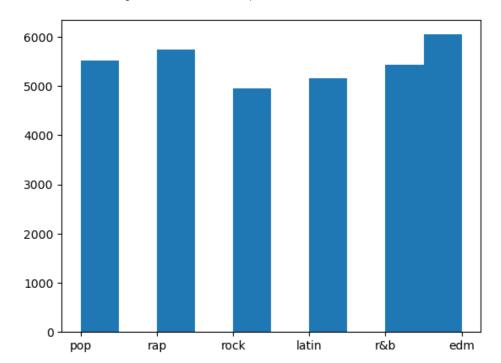
```
▶ list(df)
In [21]:
   Out[21]: ['track_popularity',
             'track_album_release_date',
             'playlist_genre',
             'danceability',
             'energy',
             'key',
             'loudness',
             'mode',
             'speechiness',
             'acousticness',
             'instrumentalness',
             'liveness',
             'valence',
             'tempo',
             'duration ms']
In [22]:
         plt.hist(df['track_popularity'])
   Out[22]: (array([5243., 1977., 2443., 4025., 4685., 5004., 4622., 3322., 1201.,
             array([ 0., 10., 20., 30., 40., 50., 60., 70., 80., 90., 100.]),
             <BarContainer object of 10 artists>)
             5000
```



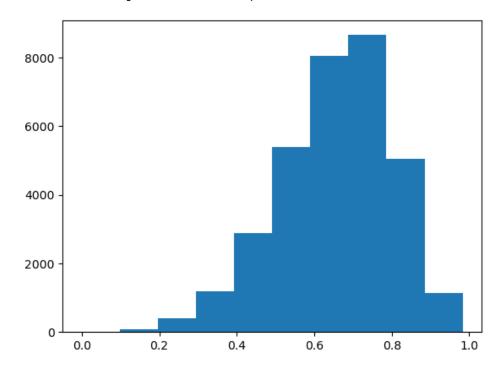
```
In [23]: | import matplotlib.pyplot as plt
plt.hist(df['track_album_release_date'])
```



```
In [24]: M import matplotlib.pyplot as plt
plt.hist(df['playlist_genre'])
```

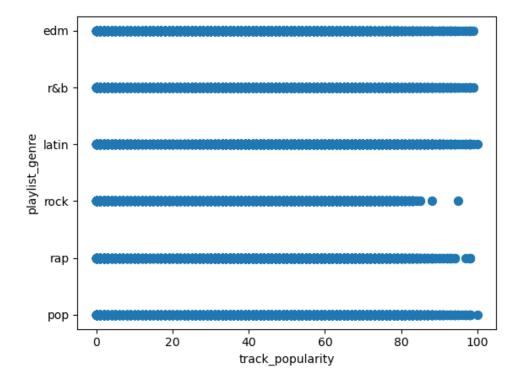


```
In [25]: M import matplotlib.pyplot as plt
plt.hist(df['danceability'])
```



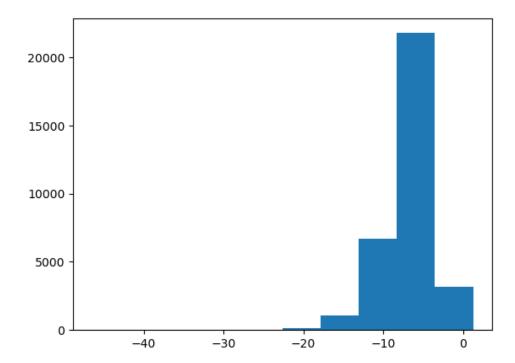
In [26]: | import matplotlib.pyplot as plt
df.plot.scatter(x='track_popularity',y='playlist_genre',s=50)

Out[26]: <Axes: xlabel='track_popularity', ylabel='playlist_genre'>



```
In [27]: | import matplotlib.pyplot as plt
plt.hist(df['loudness'])
```

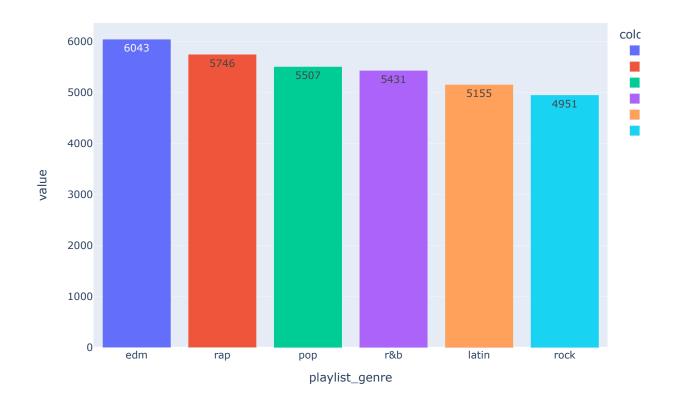
```
Out[27]: (array([1.000e+00, 0.000e+00, 5.000e+00, 2.000e+00, 2.100e+01, 1.170e+02, 1.087e+03, 6.677e+03, 2.178e+04, 3.143e+03]), array([-46.448, -41.6757, -36.9034, -32.1311, -27.3588, -22.5865, -17.8142, -13.0419, -8.2696, -3.4973, 1.275]), <BarContainer object of 10 artists>)
```



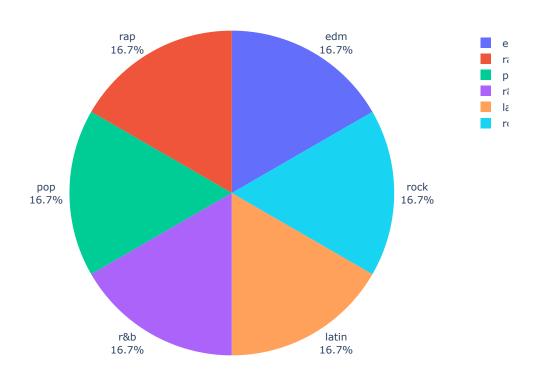
```
Out[28]: playlist_genre
edm 6043
rap 5746
pop 5507
r&b 5431
latin 5155
rock 4951
```

Name: count, dtype: int64

```
In [29]: M import plotly.express as px
fig=px.bar(df.playlist_genre.value_counts(),color=['red','blue','green','orange','teal','black'
fig.update_layout(width = 800)
fig.show()
```



```
import plotly.express as px
fig=px.pie(df.playlist_genre.value_counts().values,df.playlist_genre.value_counts().index,color
fig.update_layout(width = 800)
fig.update_traces(textposition ='outside', textinfo = 'percent+label')
fig.show()
```





 df.groupby('playlist_genre').sum() In [32]: Out[32]: track_popularity track_album_release_date danceability loudness mode speeci energy key playlist_genre edm 210499 12187047.0 3958.4120 4849.362000 32343 -32798.051 3143 52 latin 242422 10387623.0 3651.350775 28270 2897 52 3676.9958 -32293.264 pop 262931 11095162.0 3520.6345 3860.560340 29291 -34778.512 3239 40 r&b 223885 10918823.0 3639.7440 3209.364400 29330 -42713.989 2832 63 248316 11568673.0 4127.6550 3738.970500 31436 -40464.880 2996 113 rap rock 206597 9908874.0 2577.2330 3628.158500 25790 -37572.619 3467 28 In [33]: df.groupby('key').sum()

Out[33]:

	track_popularity	track_album_release_date	playlist_genre	danceability	
key					
0	148823	6948817.0	poppoppoppoppoppoppoppoppoppoppoppoppop	2240.3720	23
1	172406	8072418.0	poppoppoppoppoppoppoppoppoppoppoppoppop	2733.3045	28
2	117551	5685082.0	poppoppoppoppoppoppoppoppoppoppoppoppop	1804.2540	19
3	37798	1837775.0	poppoppoppoppoppoppoppoppoppoppoppoppop	565.7511	6
4	92504	4426188.0	poppoppoppoppoppoppoppoppoppoppoppoppop	1389.4380	15
5	113896	5394182.0	poppoppoppoppoppoppoppoppoppoppoppoppop	1744.2180	18
6	113693	5374194.0	poppoppoppoppoppoppoppoppoppoppoppoppop	1767.2700	18
7	137112	6743752.0	poppoppoppoppoppoppoppoppoppoppoppoppop	2188.8507	23
8	108323	4892192.0	poppoppoppoppoppoppoppoppoppoppoppoppop	1606.2380	17
9	127843	6086585.0	poppoppoppoppoppoppoppoppoppoppoppoppop	1942.0910	20
10	97643	4574948.0	poppoppoppoppoppoppoppoppoppoppoppoppop	1523.3030	15
11	127058	6030069.0	poppoppoppoppoppoppoppoppoppoppoppoppop	1995.5840	21
4					•

```
In [34]:

    df.groupby('track_album_release_date').sum()

   Out[34]:
                                    track_popularity
                                                                                  playlist_genre danceability
                                                                                                             en
              track_album_release_date
                                               59
                                                                                          r&b
                                                                                                   0.5650
                                                                                                            0.96
                             1957.0
                             1958.0
                                               73
                                                                                                   0.6470
                                                                                                            0.58
                                                                                          rock
                             1961.0
                                               47
                                                                                          r&b
                                                                                                   0.5850
                                                                                                            0.03
                             1963.0
                                                                                                   2.0460
                                                                                                            2.05
                                              145
                                                                                 rockrockrock
                                              342
                                                                                                   4.5610
                             1964.0
                                                                    rockrockrockrockrockrock
                                                                                                            6.34
                                                                                                1374.9870 1471.47
                             2016.0
                                            81346
                                                  2017.0
                                           101379
                                                  1608.8787
                                                                                                         1656.22
                             2018.0
                                           151950
                                                                                                2248.5375
                                                                                                        2234.90
                                                  2019.0
                                           466737
                                                  6150.8681
                                                                                                        6325.26
                             2020.0
                                            36472
                                                  527.5090
                                                                                                          526.21
             61 rows × 14 columns

    df['mode']=df['mode'].map({1:'high',0:'low'})

In [36]:
          df=pd.get_dummies(df,dtype=int)
             df
   Out[36]:
                    track_popularity track_album_release_date danceability
                                                                              loudness
                                                                                      speechiness acousticness in
                                                                  energy
                                                                         key
                  0
                               66
                                                 2019.0
                                                             0.748
                                                                    0.916
                                                                            6
                                                                                 -2.634
                                                                                            0.0583
                                                                                                      0.102000
                  1
                                                 2019.0
                                                                           11
                                                                                 -4.969
                                                                                            0.0373
                                                                                                      0.072400
                               67
                                                             0.726
                                                                    0.815
                  2
                               70
                                                 2019.0
                                                             0.675
                                                                    0.931
                                                                            1
                                                                                 -3.432
                                                                                            0.0742
                                                                                                      0.079400
                                                                            7
                  3
                               60
                                                 2019.0
                                                             0.718
                                                                    0.930
                                                                                 -3.778
                                                                                            0.1020
                                                                                                      0.028700
                  4
                               69
                                                  2019.0
                                                             0.650
                                                                    0.833
                                                                            1
                                                                                 -4.672
                                                                                            0.0359
                                                                                                      0.080300
                                                     ...
                                                                ...
              32828
                               42
                                                 2014.0
                                                             0.428
                                                                    0.922
                                                                            2
                                                                                 -1.814
                                                                                            0.0936
                                                                                                      0.076600
              32829
                                                 2013.0
                                                                            0
                                                                                            0.0420
                                                                                                      0.001710
                               20
                                                             0.522
                                                                    0.786
                                                                                 -4.462
              32830
                               14
                                                 2014.0
                                                             0.529
                                                                    0.821
                                                                            6
                                                                                 -4.899
                                                                                            0.0481
                                                                                                      0.108000
              32831
                               15
                                                 2014.0
                                                             0.626
                                                                    0.888
                                                                            2
                                                                                 -3.361
                                                                                            0.1090
                                                                                                      0.007920
              32832
                                                 2014.0
                                                             0.603
                                                                    0.884
                                                                                 -4.571
                                                                                            0.0385
                                                                                                      0.000133
                               27
             32833 rows × 21 columns
```

```
In [37]:
          df.info()
             <class 'pandas.core.frame.DataFrame'>
             RangeIndex: 32833 entries, 0 to 32832
             Data columns (total 21 columns):
             #
                 Column
                                           Non-Null Count Dtype
             - - -
                                           -----
             0
                 track_popularity
                                           32833 non-null int64
                  track_album_release_date 32833 non-null float64
             1
                                           32833 non-null float64
             2
                  danceability
             3
                  energy
                                           32833 non-null float64
                                           32833 non-null int64
             4
                  key
             5
                  loudness
                                           32833 non-null float64
             6
                  speechiness
                                           32833 non-null float64
             7
                  acousticness
                                           32833 non-null float64
             8
                  instrumentalness
                                           32833 non-null float64
             9
                  liveness
                                           32833 non-null
                                                          float64
                                           32833 non-null float64
             10
                 valence
                                           32833 non-null float64
             11
                 tempo
                 duration_ms
             12
                                           32833 non-null int64
             13
                                           32833 non-null int32
                 playlist_genre_edm
             14
                                           32833 non-null int32
                 playlist_genre_latin
             15 playlist_genre_pop
                                           32833 non-null int32
             16 playlist_genre_r&b
                                           32833 non-null int32
             17 playlist_genre_rap
                                           32833 non-null int32
                                           32833 non-null int32
             18 playlist_genre_rock
             19 mode_high
                                           32833 non-null int32
             20 mode low
                                           32833 non-null int32
             dtypes: float64(10), int32(8), int64(3)
             memory usage: 4.3 MB
          In [38]:
             X=df.drop('playlist genre edm',axis=1)#removed for model
In [39]:
          X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.32,random_state=42)
 In [ ]:
In [40]:
          X test.head(5)
   Out[40]:
                   track_popularity track_album_release_date danceability
                                                                          loudness speechiness acousticness in
                                                               energy key
             30056
                                                                             -7.717
                             45
                                               2019.0
                                                           0.520
                                                                 0.789
                                                                        0
                                                                                       0.0432
                                                                                                 0.004910
             11827
                             17
                                               1978.0
                                                           0.651
                                                                 0.661
                                                                        9
                                                                            -11.405
                                                                                       0.0511
                                                                                                 0.265000
             23571
                             30
                                               2015.0
                                                           0.640
                                                                 0.758
                                                                       10
                                                                             -5.204
                                                                                       0.1600
                                                                                                 0.665000
                                               2020.0
             14741
                             35
                                                           0.398
                                                                 0.966
                                                                        4
                                                                             -2.352
                                                                                       0.0453
                                                                                                 0.000006
             25570
                                               2015.0
                                                           0.447
                                                                 0.625
                                                                             -8.212
                                                                                       0.3230
                                                                                                 0.035100
                             62
                                                                       10
In [41]:

► X_train.shape

   Out[41]: (22326, 20)
```

```
In [ ]:
In [42]:
         Out[42]: (22326,)
In [43]:
         Ŋ y_train
   Out[43]: 28688
                   1
           28567
                   1
           28837
                   1
           6624
                   0
           7072
                   0
           16850
                   0
           6265
                   0
           11284
                   0
           860
                   0
           15795
                   0
           Name: playlist_genre_edm, Length: 22326, dtype: int32
In [44]:
         ▶ from sklearn.linear_model import LogisticRegression
           reg=LogisticRegression()
           reg.fit(X_train,y_train)
   Out[44]:
            ▼ LogisticRegression
            LogisticRegression()
In [45]:
         y_pred
   Out[45]: array([0, 0, 0, ..., 0, 0, 0])
         In [46]:
           confusion_matrix(y_test,y_pred)
   Out[46]: array([[8559,
                          1]], dtype=int64)
         ▶ | from sklearn.metrics import accuracy score
In [47]:
           accuracy_score(y_test,y_pred)
   Out[47]: 0.8146949652612544
In [48]:
         X=df.drop('playlist_genre_latin',axis=1)#removed for model
In [49]:
         from sklearn.model selection import train test split
           X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.32,random_state=42)
           from sklearn.linear_model import LogisticRegression
           reg=LogisticRegression()
           reg.fit(X_train,y_train)
           y_pred=reg.predict(X_test)
           y_pred
   Out[49]: array([0, 0, 0, ..., 0, 0, 0])
```

```
In [50]:
            confusion matrix(y test,y pred)
   Out[50]: array([[8889,
                            0],
                            0]], dtype=int64)
                   [1618,
In [51]:
         ▶ from sklearn.metrics import accuracy_score
            accuracy_score(y_test,y_pred)
   Out[51]: 0.846007423622347
In [52]:

y = df['playlist_genre_pop']#copied to check

            X=df.drop('playlist_genre_pop',axis=1)#removed for model
            from sklearn.model selection import train test split
            X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.32,random_state=42)
            from sklearn.linear_model import LogisticRegression
            reg=LogisticRegression()
            reg.fit(X_train,y_train)
            y_pred=reg.predict(X_test)
            y_pred
   Out[52]: array([0, 0, 0, ..., 0, 0, 0])

    ★ from sklearn.metrics import confusion_matrix

In [53]:
            confusion_matrix(y_test,y_pred)
   Out[53]: array([[8743,
                            0]], dtype=int64)
                   [1764]
         | from sklearn.metrics import accuracy score
In [54]:
            accuracy score(y test,y pred)
   Out[54]: 0.832111925383078
In [55]:  ▶| y=df['playlist_genre_r&b']#copied to check
            X=df.drop('playlist_genre_r&b',axis=1)#removed for model
            from sklearn.model_selection import train_test_split
            X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.32,random_state=42)
            from sklearn.linear_model import LogisticRegression
            reg=LogisticRegression()
            reg.fit(X_train,y_train)
            y_pred=reg.predict(X_test)
            y_pred
   Out[55]: array([0, 0, 0, ..., 0, 0, 0])
confusion_matrix(y_test,y_pred)
   Out[56]: array([[8808,
                            0],
                            0]], dtype=int64)
                   [1699]
         ▶ from sklearn.metrics import accuracy_score
In [57]:
            accuracy_score(y_test,y_pred)
   Out[57]: 0.8382982773389169
```

```
In [58]:

    | y=df['playlist_genre_rap']#copied to check

            X=df.drop('playlist genre rap',axis=1)#removed for model
            from sklearn.model selection import train test split
            X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.32,random_state=42)
            from sklearn.linear model import LogisticRegression
            reg=LogisticRegression()
            reg.fit(X train,y train)
            y_pred=reg.predict(X_test)
            y_pred
   Out[58]: array([0, 0, 0, ..., 0, 0, 0])
In [59]:
         ▶ from sklearn.metrics import confusion_matrix
            confusion matrix(y test,y pred)
   Out[59]: array([[8658,
                            0],
                   [1849,
                            0]], dtype=int64)
In [60]:

    ★ from sklearn.metrics import accuracy_score

            accuracy_score(y_test,y_pred)
   Out[60]: 0.82402208051775
In [61]:

y=df['playlist_genre_rock']#copied to check

            X=df.drop('playlist_genre_rock',axis=1)#removed for model
            from sklearn.model_selection import train_test_split
            X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.32,random_state=42)
            from sklearn.linear_model import LogisticRegression
            reg=LogisticRegression()
            reg.fit(X_train,y_train)
            y_pred=reg.predict(X_test)
            y pred
   Out[61]: array([0, 0, 0, ..., 0, 0, 0])
In [62]:
         confusion_matrix(y_test,y_pred)
   Out[62]: array([[8866,
                           12],
                           14]], dtype=int64)
                   [1615]
In [63]:
         accuracy_score(y_test,y_pred)
   Out[63]: 0.845150851813077
 In [ ]:
```