In [126...
import pandas as pd
import numpy as np
from sklearn.model_selection import KFold
spotify = pd.read_csv('dataset.csv')

In [127... spotify.head(10)

Out[127]:

Unnamed: 0	track_id	artists	album_name	track_name	popularity	duration_ms	
0 5SuOikwiRyPMVolQDJUgSV		Gen Hoshino	Comedy	Comedy	73	230666	
1 1	4qPNDBW1i3p13qLCt0Ki3A	Ben Woodward	Ghost (Acoustic)	Ghost - Acoustic	55	149610	
2 2	1iJBSr7s7jYXzM8EGcbK5b	Ingrid Michaelson;ZAYN	To Begin Again	To Begin Again	57	210826	
3 6lfxq3CG4xtTiEg7opyCyx		Kina Grannis	Crazy Rich Asians (Original Motion Picture Sou	Can't Help Falling In Love	71	201933	
4 4	5vjLSffimiIP26QG5WcN2K	Chord Overstreet	Hold On	Hold On	82	198853	
5 5	01MVOl9KtVTNfFiBU9I7dc	Tyrone Wells	Days I Will Remember	Days I Will Remember	58	214240	
6 6	6Vc5wAMmXdKIAM7WUoEb7N	A Great Big World;Christina Aguilera	Is There Anybody Out There?	Say Something	74	229400	
7 7	1EzrEOXmMH3G43AXT1y7pA	Jason Mraz	We Sing. We Dance. We Steal Things.	I'm Yours	80	242946	
8 8	01ktbUcnAGrvD03AWnz3Q8	Jason Mraz;Colbie Caillat	We Sing. We Dance. We Steal Things.	Lucky	74	189613	
9	7k9GuJYLp2AzqokyEdwEw2	Ross Copperman	Hunger	Hunger	56	205594	

10 rows × 21 columns

In [128... spotify.describe()

Out[128]:

	Unnamed: 0	popularity	duration_ms	danceability	energy	key	loudness
count	114000.000000	114000.000000	1.140000e+05	114000.000000	114000.000000	114000.000000	114000.000000
mean	56999.500000	33.238535	2.280292e+05	0.566800	0.641383	5.309140	-8.258960
std	32909.109681	22.305078	1.072977e+05	0.173542	0.251529	3.559987	5.029337
min	0.000000	0.000000	0.000000e+00	0.000000	0.000000	0.000000	-49.531000
25%	28499.750000	17.000000	1.740660e+05	0.456000	0.472000	2.000000	-10.013000
50%	56999.500000	35.000000	2.129060e+05	0.580000	0.685000	5.000000	-7.004000
75%	85499.250000	50.000000	2.615060e+05	0.695000	0.854000	8.000000	-5.003000
max	113999.000000	100.000000	5.237295e+06	0.985000	1.000000	11.000000	4.532000

```
In [129... spotify.info()
             <class 'pandas.core.frame.DataFrame'>
              RangeIndex: 114000 entries, 0 to 113999
              Data columns (total 21 columns):
               # Column
                                                Non-Null Count Dtype
              O Unnamed: 0 114000 non-null int64
1 track_id 114000 non-null object
2 artists 113999 non-null object
3 album_name 113999 non-null object
4 track_name 113999 non-null object
5 popularity 114000 non-null int64
6 duration_ms 114000 non-null int64
7 explicit 114000 non-null bool
8 danceability 114000 non-null float64
9 energy 114000 non-null float64
10 key 114000 non-null int64
11 loudness 114000 non-null int64
              ---
                                                  -----
              10 key 114000 non-null int64
11 loudness 114000 non-null float64
12 mode 114000 non-null int64
13 speechiness 114000 non-null float64
14 acousticness 114000 non-null float64
               15 instrumentalness 114000 non-null float64

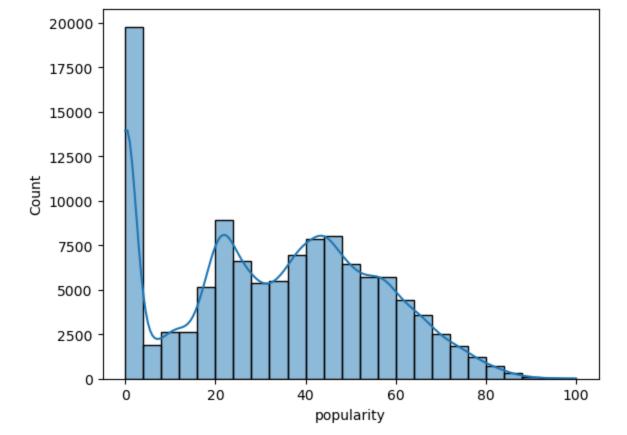
      16 liveness
      114000 non-null float64

      17 valence
      114000 non-null float64

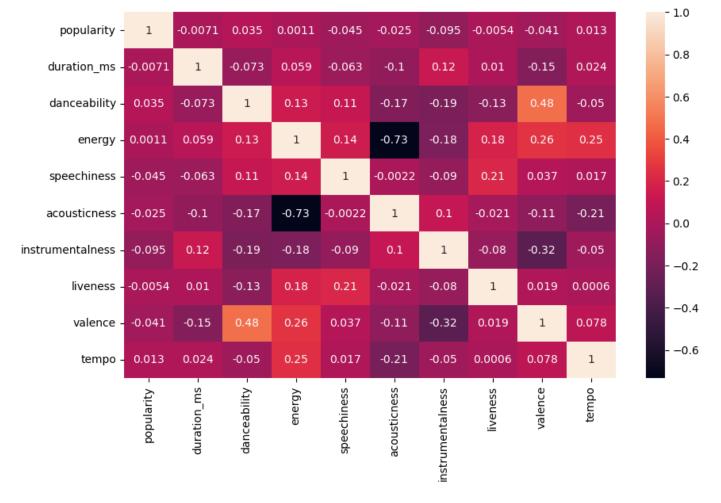
      18 tempo
      114000 non-null float64

               19 time_signature 114000 non-null int64
               20 track genre 114000 non-null object
              dtypes: bool(1), float64(9), int64(6), object(5)
             memory usage: 17.5+ MB
In [130... from scipy.stats import skew
              import matplotlib.pyplot as plt
              import seaborn as sns
              %matplotlib inline
In [131... dataset=spotify.copy()
              print(spotify['popularity'], 'skewness is :', skew(dataset['popularity']))
              sns.histplot(x='popularity', data=dataset, bins=25, kde=True)
              plt.show()
              0 73
              1
                             55
              2
                            57
              3
                            71
                           82
             113995 21
             113996 22
              113997 22
             113998 41
             113999 22
```

Name: popularity, Length: 114000, dtype: int64 skewness is: 0.04640190527012667



```
In [132... combined = ['popularity','duration_ms', 'danceability', 'energy', 'speechiness', 'acoust
In [133... dataset=spotify.copy()
   plt.figure(figsize=(10,6))
   sns.heatmap(spotify[combined].corr(), annot=True)
   plt.show()
```



```
In [134... # Normalization process
numerical_features = ['duration_ms', 'danceability', 'energy', 'speechiness', 'acoustion
target_feature = 'popularity'

spotify[numerical_features] = (spotify[numerical_features] - spotify[numerical_features]
spotify[target_feature] = (spotify[target_feature] - spotify[target_feature].min()) / (s

# Converting the data to 64-bit
spotify[numerical_features] = spotify[numerical_features].astype('float64')
spotify[target_feature] = spotify[target_feature].astype('float64')

# Gradient clipping using 0.5 threshold
spotify[numerical_features] = np.clip(spotify[numerical_features], 0, 0.5)

X = spotify[numerical_features].values
y = spotify[target_feature].values
```

```
In [135... # Ridge Regression function
def ridge_regression(X, y, alpha, learning_rate, num_iterations):
    m, n = X.shape
    weights = np.zeros(n)
    for i in range(num_iterations):
        # Calculate predictions and error
        predictions = np.dot(X, weights)
        error = predictions - y

# Compute gradient and update weights
        gradient = 2/m * (np.dot(X.T, error) + alpha * weights)
        weights -= learning_rate * gradient

return weights
```

```
def k fold cross validation(X, y, alpha, learning rate, num iterations, num folds=5):
             kf = KFold(n splits=num folds)
             mse scores = []
             mae scores = []
             for train index, test index in kf.split(X):
                  X train, X test = X[train index], X[test index]
                  y train, y test = y[train index], y[test index]
                  # Trainining
                 weights = ridge regression(X train, y train, alpha, learning rate, num iteration
                 # mean squared error
                 predictions = np.dot(X test, weights)
                 mse = np.mean((predictions - y test) ** 2)
                 mse scores.append(mse)
                  # Mean absolute error
                 mae = mean absolute error(y test, predictions)
                 mae scores.append(mae)
             avg mse = np.mean(mse scores)
             avg mae = np.mean(mae scores)
             return avg mse, avg mae
         alpha = 10
          learning rate = 0.001
         num iterations = 10000
          # Appllying the algortihm
          avg mse, avg mae = k fold cross validation(X, y, alpha, learning rate, num iterations, n
          # results
         print("Mean squared error:", avg mse)
         print("Mean absolute error:", avg mae)
         Mean squared error: 0.05108363768221352
         Mean absolute error: 0.18996849886361447
In [137... categorical features = ['explicit', 'time signature', 'track genre', 'key']
In [138... spotify[categorical features]
Out[138]:
                 explicit time_signature track_genre key
```

0	False	4	acoustic	1
1	False	4	acoustic	1
2	False	4	acoustic	0
3	False	3	acoustic	0
4	False	4	acoustic	2
•••				
113995	False	5	world-music	5
113996	False	4	world-music	0
113997	False	4	world-music	0

4 world-music

113998

False

113999 False 4 world-music 1

114000 rows × 4 columns

```
In [139... spotify[categorical_features].describe()
```

Out[139]:

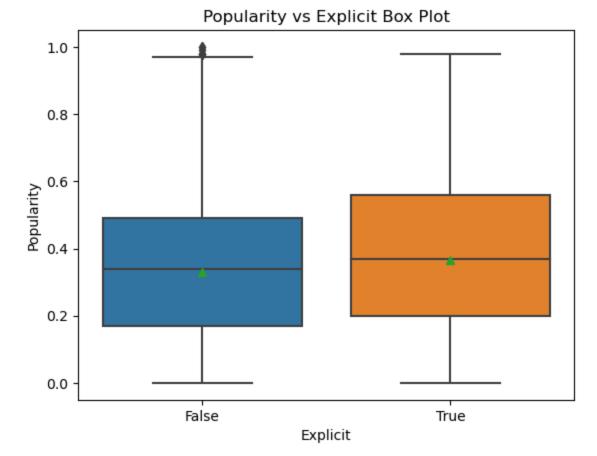
	time_signature	key
count	114000.000000	114000.000000
mean	3.904035	5.309140
std	0.432621	3.559987
min	0.000000	0.000000
25%	4.000000	2.000000
50%	4.000000	5.000000
75%	4.000000	8.000000
max	5.000000	11.000000

```
In [140... import pandas as pd import matplotlib.pyplot as plt
```

```
In [141... import seaborn as sns

sns.boxplot(
    x = 'explicit',
    y = 'popularity',
    showmeans=True,
    data=spotify
)

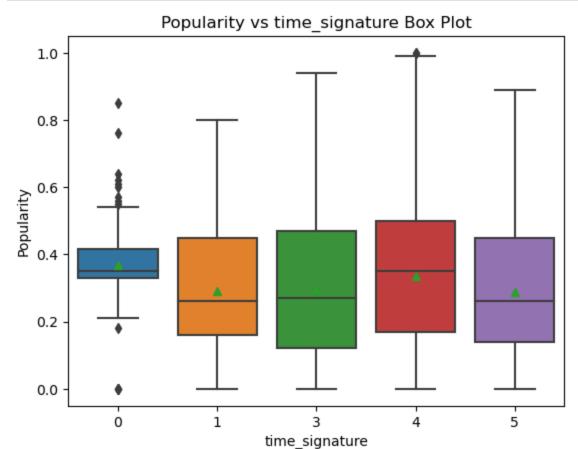
plt.title('Popularity vs Explicit Box Plot')
plt.xlabel('Explicit')
plt.ylabel('Popularity')
```



```
import pandas as pd
In [142...
          mean popularity by explicit = spotify.groupby('explicit')['popularity'].mean()
          print(mean popularity by explicit)
         explicit
                   0.329379
         False
         True
                   0.364542
         Name: popularity, dtype: float64
In [143...
          explicit_to_popularity_mean_mapping = dict(zip(mean_popularity_by_explicit.index, mean_p
          spotify['explicit'] = spotify['explicit'].replace(explicit to popularity mean mapping)
         spotify['explicit']
In [144...
                    0.329379
Out[144]:
                    0.329379
                    0.329379
          3
                    0.329379
                    0.329379
         113995
                    0.329379
          113996
                    0.329379
         113997
                    0.329379
         113998
                    0.329379
          113999
                    0.329379
         Name: explicit, Length: 114000, dtype: float64
         sns.boxplot(
In [145...
              x = 'time signature',
              y = 'popularity',
              showmeans=True,
              data=spotify
```

```
plt.title('Popularity vs time_signature Box Plot')
plt.xlabel('time_signature')
plt.ylabel('Popularity')

plt.show()
```



In [146...

```
print(mean popularity by time signature)
         time signature
               0.369755
         1
              0.290740
         3
               0.296924
               0.336725
               0.287782
         Name: popularity, dtype: float64
         time signature to popularity mean mapping = dict(zip(mean popularity by time signature.i
In [147...
          spotify['time signature'] = spotify['time signature'].replace(time signature to populari
         spotify['time_signature']
In [148...
                    0.336725
Out[148]:
                    0.336725
          2
                    0.336725
          3
                    0.296924
                    0.336725
                      . . .
          113995
                    0.287782
          113996
                    0.336725
         113997
                    0.336725
          113998
                    0.336725
         113999
                    0.336725
         Name: time signature, Length: 114000, dtype: float64
```

mean popularity by time signature = spotify.groupby('time signature')['popularity'].mean

In [149... spotify.head()

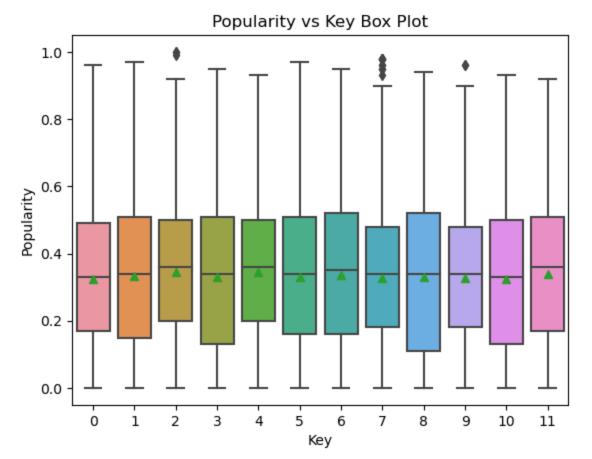
Out[149]:	Unnamed: 0	track_id	artists	album_name	track_name	popularity	duration_ms	ex
	0 0	5SuOikwiRyPMVoIQDJUgSV	Gen Hoshino	Comedy	Comedy	0.73	0.044043	0.32
	1 1	4qPNDBW1i3p13qLCt0Ki3A	Ben Woodward	Ghost (Acoustic)	Ghost - Acoustic	0.55	0.028566	0.32
	2 2	1iJBSr7s7jYXzM8EGcbK5b	Ingrid Michaelson;ZAYN	To Begin Again	To Begin Again	0.57	0.040255	0.32
	3 3	6lfxq3CG4xtTiEg7opyCyx	Kina Grannis	Crazy Rich Asians (Original Motion Picture Sou	Can't Help Falling In Love	0.71	0.038557	0.32
	4 4	5vjLSffimiIP26QG5WcN2K	Chord Overstreet	Hold On	Hold On	0.82	0.037969	0.32

5 rows × 21 columns

```
In [150...
sns.boxplot(
    x = 'key',
    y = 'popularity',
    showmeans=True,
    data=spotify
)

plt.title('Popularity vs Key Box Plot')
plt.xlabel('Key')
plt.ylabel('Popularity')

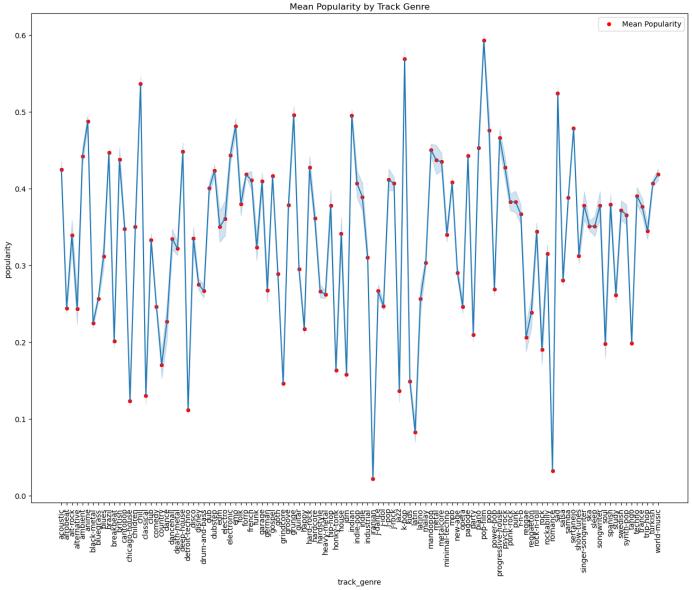
plt.show()
```



```
In [151... import pandas as pd
        mean popularity by key = spotify.groupby('key')['popularity'].mean()
        print(mean popularity by key)
        key
              0.325199
        0
              0.332002
        2
              0.343776
              0.330317
        4
              0.346437
        5
             0.329309
        6
              0.335855
        7
             0.325796
        8
             0.329587
             0.328236
        9
        10
              0.324614
        11
              0.338876
        Name: popularity, dtype: float64
In [152... import pandas as pd
        mean popularity by genre = spotify.groupby('track genre')['popularity'].mean()
        print(mean_popularity_by_genre)
        track genre
                       0.42483
        acoustic
                      0.24399
        afrobeat
        alt-rock
                      0.33943
        alternative 0.24337
                      0.44191
        ambient
                        . . .
        techno
                       0.39042
        trance
                      0.37635
        trip-hop
                      0.34460
                       0.40698
        turkish
        world-music 0.41873
        Name: popularity, Length: 114, dtype: float64
In [153... a = spotify['track genre'].unique()
In [154... print(a)
         ['acoustic' 'afrobeat' 'alt-rock' 'alternative' 'ambient' 'anime'
         'black-metal' 'bluegrass' 'blues' 'brazil' 'breakbeat' 'british'
          'cantopop' 'chicago-house' 'children' 'chill' 'classical' 'club' 'comedy'
         'country' 'dance' 'dancehall' 'death-metal' 'deep-house' 'detroit-techno'
         'disco' 'disney' 'drum-and-bass' 'dub' 'dubstep' 'edm' 'electro'
         'electronic' 'emo' 'folk' 'forro' 'french' 'funk' 'garage' 'german'
         'gospel' 'goth' 'grindcore' 'groove' 'grunge' 'guitar' 'happy'
         'hard-rock' 'hardcore' 'hardstyle' 'heavy-metal' 'hip-hop' 'honky-tonk'
          'house' 'idm' 'indian' 'indie-pop' 'indie' 'industrial' 'iranian'
          'j-dance' 'j-idol' 'j-pop' 'j-rock' 'jazz' 'k-pop' 'kids' 'latin'
         'latino' 'malay' 'mandopop' 'metal' 'metalcore' 'minimal-techno' 'mpb'
         'new-age' 'opera' 'pagode' 'party' 'piano' 'pop-film' 'pop' 'power-pop'
         'progressive-house' 'psych-rock' 'punk-rock' 'punk' 'r-n-b' 'reggae'
          'reggaeton' 'rock-n-roll' 'rock' 'rockabilly' 'romance' 'sad' 'salsa'
         'samba' 'sertanejo' 'show-tunes' 'singer-songwriter' 'ska' 'sleep'
          'songwriter' 'soul' 'spanish' 'study' 'swedish' 'synth-pop' 'tango'
          'techno' 'trance' 'trip-hop' 'turkish' 'world-music']
In [161... | mean popularity by track genre = spotify.groupby('track genre')['popularity'].mean()
```

```
mean_popularity_df = pd.DataFrame({'track_genre': mean_popularity_by_track_genre.index,

plt.figure(figsize=(16,12))
sns.lineplot(x='track_genre', y='popularity', data=spotify)
sns.scatterplot(x='track_genre', y='popularity', data=mean_popularity_df, label='Mean Popularity's (rotation=90)
plt.xticks(rotation=90)
plt.title('Mean Popularity by Track Genre')
plt.legend()
plt.show()
```



```
mean_popularity_by_genre.describe()
In [162...
                    114.000000
          count
Out[162]:
          mean
                      0.332385
                      0.112948
          std
                      0.022100
          min
          25%
                      0.257840
          50%
                      0.348855
          75%
                      0.411240
                      0.592830
          max
          Name: popularity, dtype: float64
In [163...
          genre_to_popularity_mean_mapping = dict(zip(mean_popularity_by_genre.index, mean_popular
          spotify['track_genre'] = spotify['track_genre'].replace(genre_to_popularity_mean_mapping)
In [164...
          spotify['track genre']
In [165...
```

```
0.42483
           0
Out[165]:
                      0.42483
                      0.42483
           3
                      0.42483
                      0.42483
           113995
                      0.41873
           113996
                      0.41873
           113997
                      0.41873
           113998
                      0.41873
           113999
                      0.41873
           Name: track genre, Length: 114000, dtype: float64
           spotify.head()
In [166...
Out[166]:
              Unnamed:
                                       track_id
                                                        artists album_name track_name popularity duration_ms
                                                                                                              ex
           0
                     0 5SuOikwiRyPMVoIQDJUgSV
                                                                                            0.73
                                                                                                             0.32
                                                   Gen Hoshino
                                                                   Comedy
                                                                              Comedy
                                                                                                    0.044043
                                                                     Ghost
                                                                               Ghost -
           1
                        4qPNDBW1i3p13qLCt0Ki3A
                                                 Ben Woodward
                                                                                            0.55
                                                                                                    0.028566
                                                                                                            0.32
                                                                  (Acoustic)
                                                                              Acoustic
                                                        Ingrid
                                                                   To Begin
                                                                              To Begin
           2
                     2
                                                                                                    0.040255 0.32
                          1iJBSr7s7jYXzM8EGcbK5b
                                                                                            0.57
                                               Michaelson;ZAYN
                                                                     Again
                                                                                Again
                                                                 Crazy Rich
                                                                    Asians
                                                                             Can't Help
           3
                     3
                          6lfxq3CG4xtTiEg7opyCyx
                                                   Kina Grannis
                                                                   (Original
                                                                              Falling In
                                                                                            0.71
                                                                                                    0.038557 0.32
                                                                    Motion
                                                                                 Love
                                                                Picture Sou...
           4
                                                                                                    0.037969 0.32
                          5vjLSffimiIP26QG5WcN2K Chord Overstreet
                                                                   Hold On
                                                                              Hold On
                                                                                            0.82
          5 rows × 21 columns
           target feature
In [167...
           'popularity'
Out[167]:
           import pandas as pd
In [168...
           import numpy as np
           #Normalizationv
           categorical features = ['track genre', 'explicit', 'time signature'] # Replace these wi
           # Converting into 64 bit
           spotify[categorical_features] = spotify[categorical features].astype('float64')
           # Gradient clipping (thershold = 0.5)
           spotify[categorical features] = np.clip(spotify[categorical features], 0, 0.5)
           combined = ['track genre', 'explicit', 'time signature','duration ms', 'danceability',
           combined for corr = ['popularity','track genre', 'explicit', 'time signature','duration
           combined
In [169...
           ['track genre',
Out[169]:
            'explicit',
            'time signature',
            'duration ms',
```

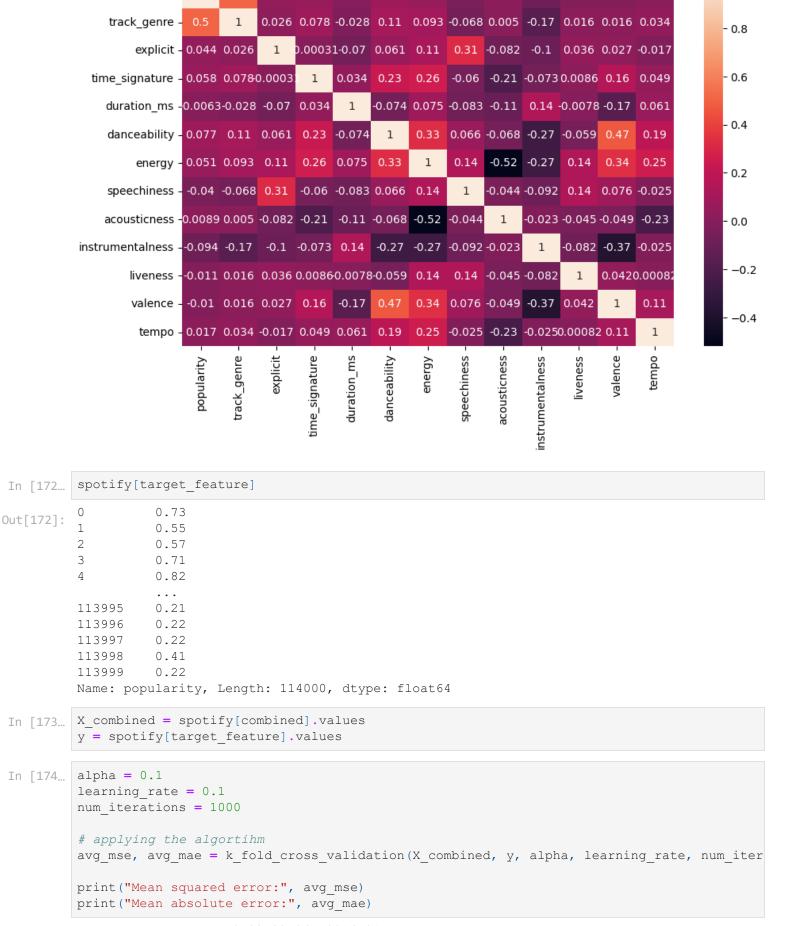
```
'danceability',
'energy',
'speechiness',
'acousticness',
'instrumentalness',
'liveness',
'valence',
'tempo']
```

In [170... spotify[combined]

Out[170]:		track_genre	explicit	time_signature	duration_ms	danceability	energy	speechiness	acousticness	inst
	0	0.42483	0.329379	0.336725	0.044043	0.500000	0.4610	0.148187	0.032329	
	1	0.42483	0.329379	0.336725	0.028566	0.426396	0.1660	0.079067	0.500000	
	2	0.42483	0.329379	0.336725	0.040255	0.444670	0.3590	0.057720	0.210843	
	3	0.42483	0.329379	0.296924	0.038557	0.270051	0.0596	0.037617	0.500000	
	4	0.42483	0.329379	0.336725	0.037969	0.500000	0.4430	0.054508	0.470884	
	•••									
	113995	0.41873	0.329379	0.287782	0.073511	0.174619	0.2350	0.043731	0.500000	
	113996	0.41873	0.329379	0.336725	0.073511	0.176650	0.1170	0.041554	0.500000	
	113997	0.41873	0.329379	0.336725	0.051833	0.500000	0.3290	0.043523	0.500000	
	113998	0.41873	0.329379	0.336725	0.054206	0.500000	0.5000	0.030777	0.382530	
	113999	0.41873	0.329379	0.336725	0.046174	0.500000	0.4870	0.075130	0.500000	

114000 rows × 12 columns

```
In [171... dataset=spotify.copy()
    plt.figure(figsize=(10,6))
    sns.heatmap(spotify[combined_for_corr].corr(), annot=True)
    plt.show()
```



0.044 0.058-0.0063 0.077 0.051 -0.04 0.0089-0.094 -0.011 -0.01 0.017

- 1.0

Mean squared error: 0.03738796510216581
Mean absolute error: 0.14498254711290487

popularity - 1