

spotify-genre-recommendation

February 7, 2024

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
from sklearn.decomposition import PCA
from sklearn.metrics import pairwise_distances_argmin_min
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report, \
    confusion_matrix
```

```
[2]: spotify_data = pd.read_csv(r'/Users/akurisivanagendrareddy/Downloads/spotify_
    dataset.csv')
spotify_data.head(5)
```

```
[2]:
```

	track_id	track_name	\
0	6f807x0ima9a1j3VPbc7VN	I Don't Care (with Justin Bieber) - Loud Luxur...	
1	0r7CVbZTWZgbTCYdfa2P31	Memories - Dillon Francis Remix	
2	1z1Hg7Vb0AhHDIEmnDE79l	All the Time - Don Diablo Remix	
3	75FpbthrwQmzHlBJLuGdC7	Call You Mine - Keanu Silva Remix	
4	1e8PAfcKUYoKkxPhrHqw4x	Someone You Loved - Future Humans Remix	

	track_artist	track_popularity	track_album_id	\
0	Ed Sheeran	66	2oCsODGTsR098Gh5ZS12Cx	
1	Maroon 5	67	63rPS0264uRjW1X5E6cWv6	
2	Zara Larsson	70	1HoSmj2eLcsrR0vE9gThr4	
3	The Chainsmokers	60	1nqYsOeflyKKuGOVchbsk6	
4	Lewis Capaldi	69	7m7vv9wlQ4i0LFuJiE2zsQ	

	track_album_name	track_album_release_date	\
0	I Don't Care (with Justin Bieber) [Loud Luxury...	2019-06-14	
1	Memories (Dillon Francis Remix)	2019-12-13	
2	All the Time (Don Diablo Remix)	2019-07-05	
3	Call You Mine - The Remixes	2019-07-19	
4	Someone You Loved (Future Humans Remix)	2019-03-05	

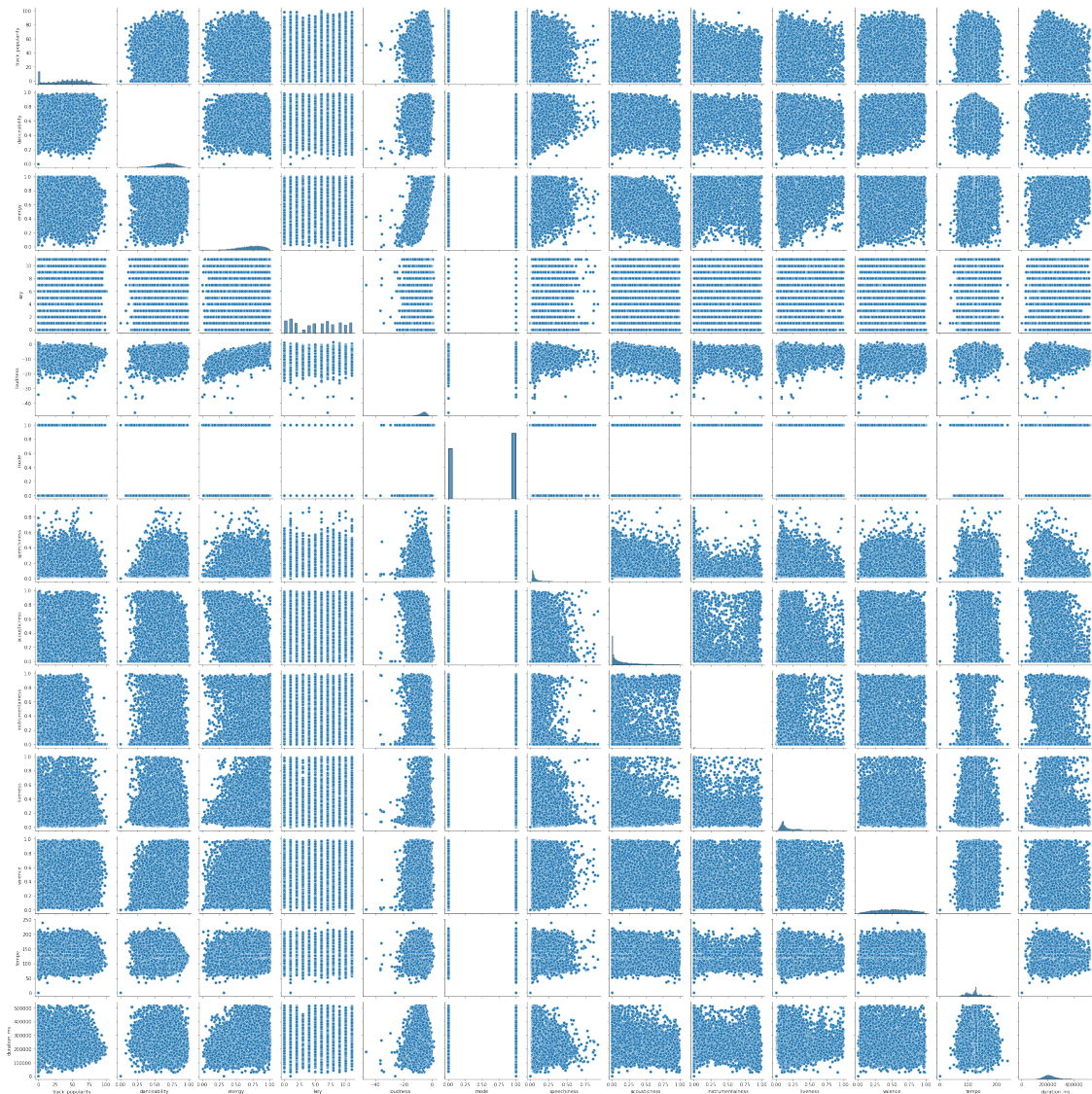
	playlist_name	playlist_id	playlist_genre	...	key	loudness	\
0	Pop Remix	37i9dQZF1DXcZDD7cfEKhW	pop	...	6	-2.634	
1	Pop Remix	37i9dQZF1DXcZDD7cfEKhW	pop	...	11	-4.969	
2	Pop Remix	37i9dQZF1DXcZDD7cfEKhW	pop	...	1	-3.432	
3	Pop Remix	37i9dQZF1DXcZDD7cfEKhW	pop	...	7	-3.778	
4	Pop Remix	37i9dQZF1DXcZDD7cfEKhW	pop	...	1	-4.672	

	mode	speechiness	acousticness	instrumentalness	liveness	valence	\
0	1	0.0583	0.1020	0.000000	0.0653	0.518	
1	1	0.0373	0.0724	0.004210	0.3570	0.693	
2	0	0.0742	0.0794	0.000023	0.1100	0.613	
3	1	0.1020	0.0287	0.000009	0.2040	0.277	
4	1	0.0359	0.0803	0.000000	0.0833	0.725	

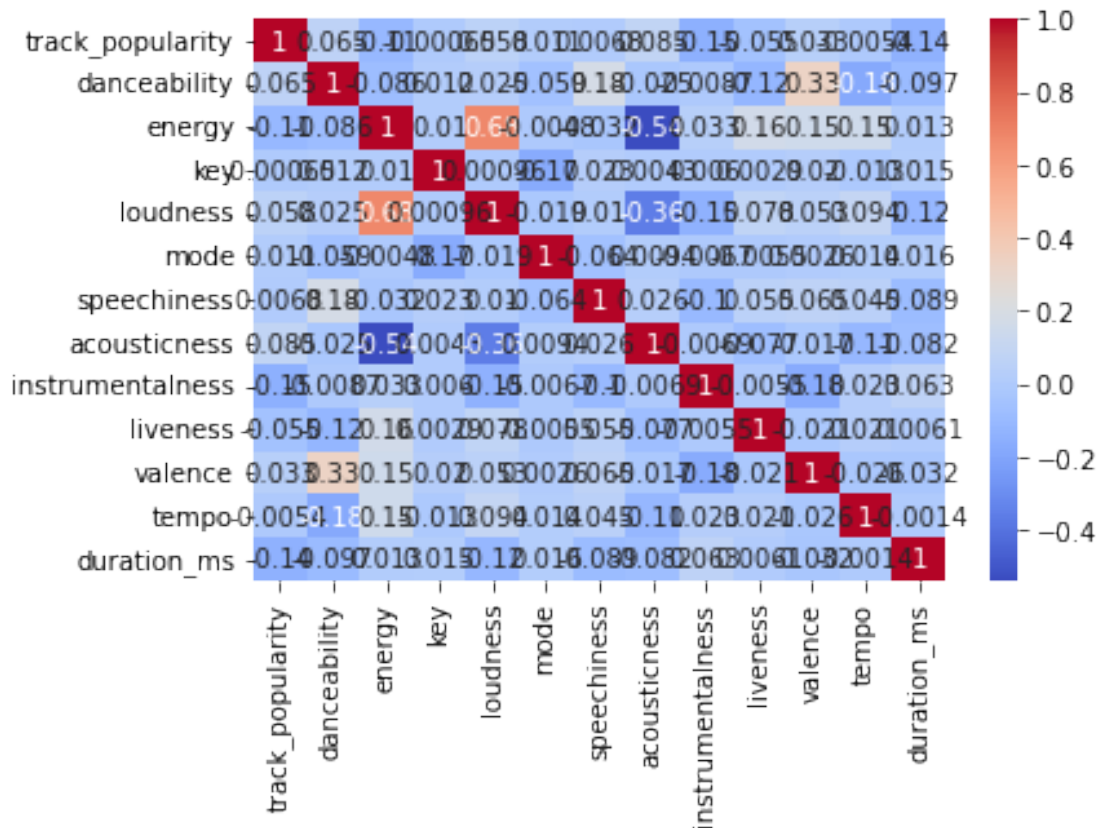
	tempo	duration_ms
0	122.036	194754
1	99.972	162600
2	124.008	176616
3	121.956	169093
4	123.976	189052

[5 rows x 23 columns]

```
[3]: sns.pairplot(spotify_data)
plt.show()
```



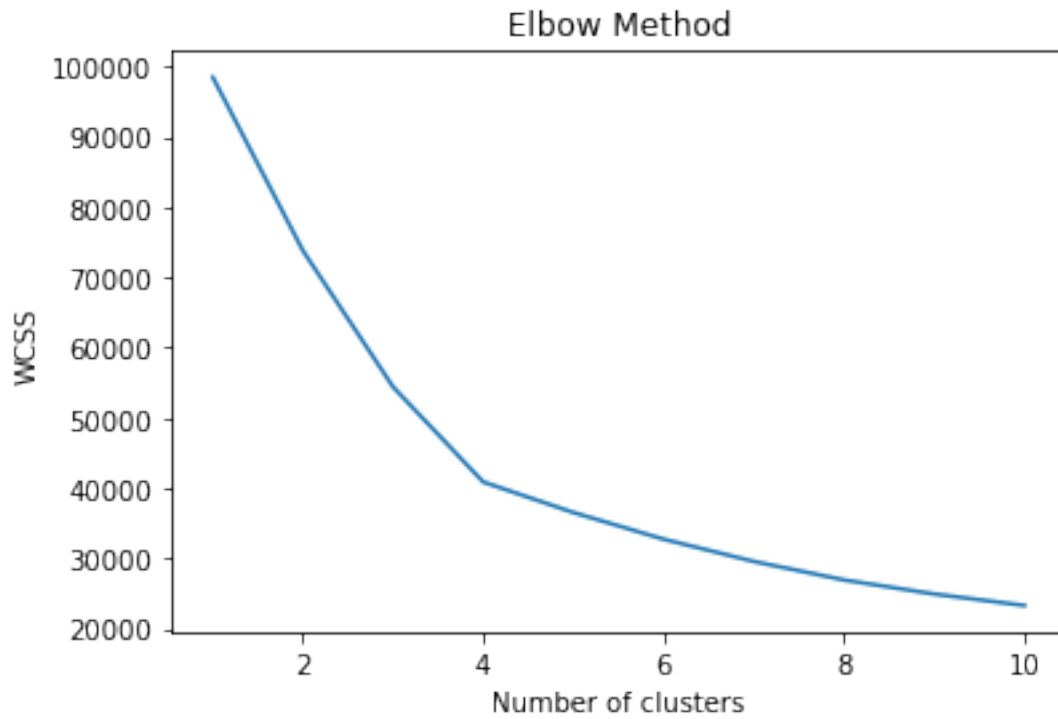
```
[4]: correlation_matrix = spotify_data.corr()
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
plt.show()
```



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[5]: features_for_clustering = spotify_data[['acousticness', 'duration_ms', 'valence']]
      scaler = StandardScaler()
      scaled_features = scaler.fit_transform(features_for_clustering)

[6]: wcss = []
      for i in range(1, 11):
          kmeans = KMeans(n_clusters=i, init='k-means++', max_iter=300, n_init=10, random_state=0)
          kmeans.fit(scaled_features)
          wcss.append(kmeans.inertia_)

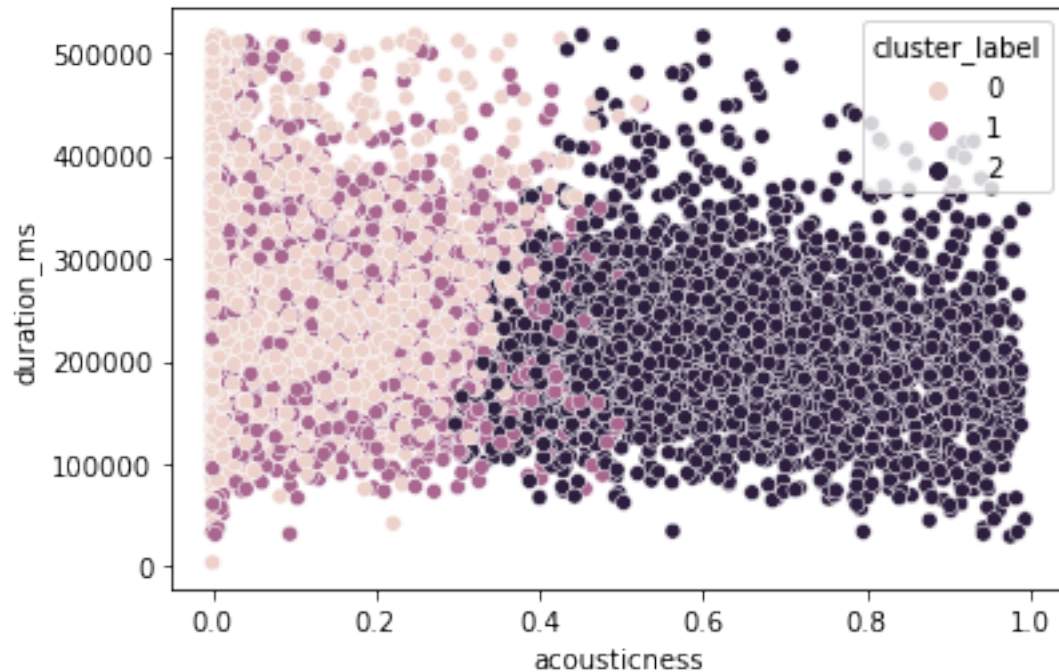
      plt.plot(range(1, 11), wcss)
      plt.title('Elbow Method')
      plt.xlabel('Number of clusters')
      plt.ylabel('WCSS')
      plt.show()
```



```
[7]: kmeans = KMeans(n_clusters=3, init='k-means++', max_iter=300, n_init=10,
    random_state=0)
cluster_labels = kmeans.fit_predict(scaled_features)

spotify_data['cluster_label'] = cluster_labels

sns.scatterplot(x='acousticness', y='duration_ms', hue='cluster_label',
    data=spotify_data)
plt.show()
```



```
[8]: X = spotify_data[['acoustiness', 'duration_ms', 'valence']]
y = spotify_data['playlist_genre']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
↳ random_state=42)

rf_model = RandomForestClassifier(n_estimators=100, random_state=42)
rf_model.fit(X_train, y_train)

y_pred = rf_model.predict(X_test)
print("Accuracy:", accuracy_score(y_test, y_pred))
print("Classification Report:\n", classification_report(y_test, y_pred))
print("Confusion Matrix:\n", confusion_matrix(y_test, y_pred))
```

Accuracy: 0.360590832952642

Classification Report:

	precision	recall	f1-score	support
edm	0.44	0.49	0.47	1218
latin	0.35	0.34	0.34	1033
pop	0.25	0.24	0.25	1081
r&b	0.30	0.33	0.31	1031
rap	0.33	0.30	0.31	1168
rock	0.49	0.45	0.47	1036

accuracy			0.36	6567
macro avg	0.36	0.36	0.36	6567
weighted avg	0.36	0.36	0.36	6567

Confusion Matrix:

```
[[601 102 191  88 138  98]
 [135 348 158 172 152  68]
 [211 168 261 187 136 118]
 [ 99 139 152 340 199 102]
 [191 162 162 209 347  97]
 [124  88 108 151  94 471]]
```

[]: