```
In [15]: import spotipy
from spotipy.oauth2 import SpotifyClientCredentials
import pandas as pd
import matplotlib as plt
```

## In [16]: !pip install spotipy

Requirement already satisfied: spotipy in ./anaconda3/lib/python3.11/site-packages (2.23.0)

Requirement already satisfied: redis>=3.5.3 in ./anaconda3/lib/python 3.11/site-packages (from spotipy) (5.0.2)

Requirement already satisfied: requests>=2.25.0 in ./anaconda3/lib/py thon3.11/site-packages (from spotipy) (2.31.0)

Requirement already satisfied: six>=1.15.0 in ./anaconda3/lib/python 3.11/site-packages (from spotipy) (1.16.0)

Requirement already satisfied: urllib3>=1.26.0 in ./anaconda3/lib/pyt hon3.11/site-packages (from spotipy) (1.26.16)

Requirement already satisfied: async-timeout>=4.0.3 in ./anaconda3/lib/python3.11/site-packages (from redis>=3.5.3->spotipy) (4.0.3)

Requirement already satisfied: charset-normalizer<4,>=2 in ./anaconda 3/lib/python3.11/site-packages (from requests>=2.25.0->spotipy) (2.0.4)

Requirement already satisfied: idna<4,>=2.5 in ./anaconda3/lib/python 3.11/site-packages (from requests>=2.25.0->spotipy) (3.4)

Requirement already satisfied: certifi>=2017.4.17 in ./anaconda3/lib/python3.11/site-packages (from requests>=2.25.0->spotipy) (2023.7.22)

## In [17]: import spotipy from spotipy.oauth2 import SpotifyClientCredentials import pandas as pd

```
In [18]: # Replace these with your Spotify API credentials
    client_id = os.getenv('SPOTIFY_CLIENT_ID')
    client_secret = os.getenv('SPOTIFY_CLIENT_SECRET')

auth_manager = SpotifyClientCredentials(client_id=client_id, client_se
    sp = spotipy.Spotify(auth_manager=auth_manager)
```

```
In [19]: def search_playlists(sp, query, limit=10):
             playlists = sp.search(q=f'playlist:"{query}"', type='playlist', li
             return [playlist['id'] for playlist in playlists['playlists']['ite
         def get tracks from playlists(sp, playlist ids):
             track ids = []
             for playlist_id in playlist_ids:
                 results = sp.playlist_tracks(playlist_id)
                 track ids.extend([item['track']['id'] for item in results['ite
             return track_ids
In [20]: def fetch_audio_features(sp, track_ids):
             features list = []
             for track_id in track_ids:
                 features = sp.audio_features(track_id)[0]
                 if features:
                     selected features = {
                          'track id': track id,
                          'danceability': features['danceability'],
                          'energy': features['energy'],
                          'valence': features['valence'],
                          'tempo': features['tempo'],
                          'acousticness': features['acousticness']
                     features list.append(selected features)
             return features list
In [21]: # Example: Adjust the limit as needed, considering API rate limits and
         happy_playlist_ids = search_playlists(sp, "Happy", limit=5)
         sad playlist ids = search_playlists(sp, "Sad", limit=5)
         # Fetch track IDs
         happy_track_ids = get_tracks_from_playlists(sp, happy_playlist_ids)
         sad_track_ids = get_tracks_from_playlists(sp, sad_playlist_ids)
         # Fetch audio features and label them
         happy_features = fetch_audio_features(sp, happy_track_ids)
         for feature in happy features:
             feature['mood'] = 'Happy'
         sad_features = fetch_audio_features(sp, sad_track_ids)
         for feature in sad features:
             feature['mood'] = 'Sad'
         # Combine happy and sad features
         all features = happy features + sad features
         # Create DataFrame
         df = pd.DataFrame(all features)
         # Check the combined DataFrame
         print(df.head()) # For the first few rows
```

```
print(df['mood'].value counts()) # To check the distribution of moods
# Save to CSV
df.to_csv('spotify_mood_audio_features.csv', index=False)
                  track_id
                            danceability
                                           energy
                                                   valence
                                                               tempo
                                                                      \
0
   3EslM7q3nwkGRvY7HNbGBx
                                   0.754
                                            0.543
                                                     0.458
                                                             104.678
1
   2okho7vU7Nsq1UZD0kqIMi
                                   0.598
                                            0.714
                                                     0.467
                                                              99.979
2
   53Iv1sdaJ8TYFfhtTBGyv0
                                   0.728
                                            0.491
                                                     0.700
                                                             129,988
  77sMIMlNaSURUAXq5coCxE
3
                                   0.569
                                            0.741
                                                     0.430
                                                             100.118
   00a9mH0nH14GZ4toWwkf4c
                                   0.572
                                                     0.540
                                            0.778
                                                             168.073
   acousticness
                  mood
0
        0.09400
                 Happy
1
        0.03640
                 Happy
2
        0.84100
                 Happy
3
        0.01220
                 Happy
4
        0.00638
                 Happy
                              danceability
                                                     valence
                    track_id
                                             energy
                                                                 tempo
/
656
     0V5cvmTKsYmF5FmGGEAfmS
                                     0.303
                                             0.1870
                                                      0.2130
                                                               132.731
     3IYU2BjHdPyTmuk81fUZXZ
                                             0.2070
657
                                     0.499
                                                      0.1890
                                                                75.036
658
     2ZosoKioRi0SEF0DWFKfsJ
                                     0.461
                                             0.0605
                                                      0.1950
                                                               126.556
659
     3eTqq18rKBD30Hef1qv0wz
                                     0.312
                                             0.0766
                                                      0.0493
                                                               136.866
660
     4R2kfaDFhslZEMJqAFNpdd
                                     0.613
                                             0.5810
                                                      0.5510
                                                               130.033
     acousticness mood
656
            0.989
                    Sad
657
            0.912
                    Sad
            0.980
                    Sad
658
659
            0.967
                    Sad
660
            0.537
                    Sad
mood
         335
Happy
         326
Sad
Name: count, dtype: int64
from IPython.display import FileLink
```

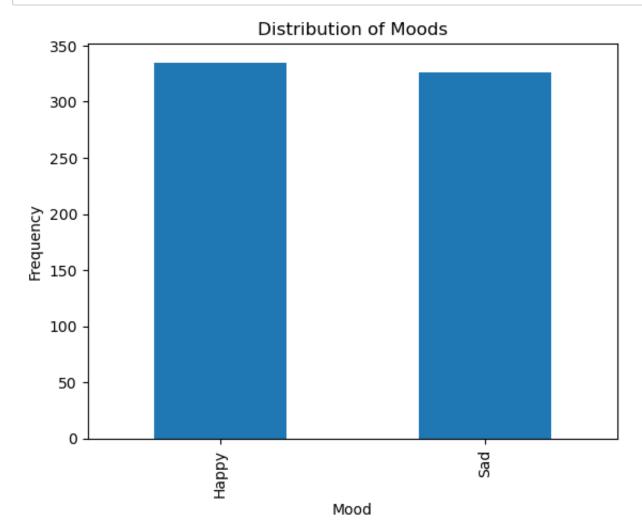
print(df.tail()) # For the last few rows to verify sad tracks are ind

```
In [22]:
         FileLink('spotify_mood_audio_features.csv')
```

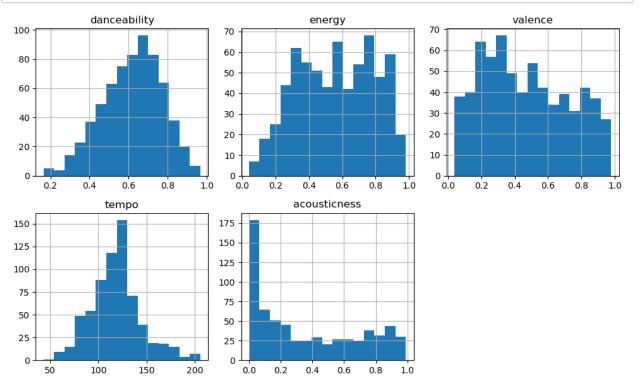
Out [22]: spotify mood audio features.csv (spotify mood audio features.csv)

```
In [23]: # Display the first few rows to verify the content
         print(df.head())
                           track_id
                                     danceability
                                                            valence
                                                   energy
                                                                       tempo
            3EslM7q3nwkGRvY7HNbGBx
                                            0.754
                                                     0.543
                                                              0.458
                                                                     104.678
         1
            2okho7vU7Nsq1UZD0kgIMi
                                            0.598
                                                     0.714
                                                              0.467
                                                                      99.979
            53Iv1sdaJ8TYFfhtTBGyv0
                                            0.728
                                                     0.491
                                                              0.700
                                                                     129.988
         3
            77sMIMlNaSURUAXq5coCxE
                                            0.569
                                                     0.741
                                                              0.430
                                                                     100.118
            00a9mH0nH14GZ4toWwkf4c
                                            0.572
                                                     0.778
                                                              0.540
                                                                     168.073
            acousticness
                            mood
         0
                 0.09400
                           Happy
         1
                 0.03640
                           Happy
         2
                 0.84100
                           Happy
         3
                 0.01220
                           Happy
         4
                 0.00638
                           Happy
In [24]: # Display the shape of the dataframe
         print("Shape of the DataFrame:", df.shape)
         # Check for missing values
         print("Missing values in each column:\n", df.isnull().sum())
         Shape of the DataFrame: (661, 7)
         Missing values in each column:
          track id
                           0
         danceability
                          0
         energy
                          0
         valence
                          0
                          0
         tempo
         acousticness
                          0
         mood
                          0
         dtype: int64
```

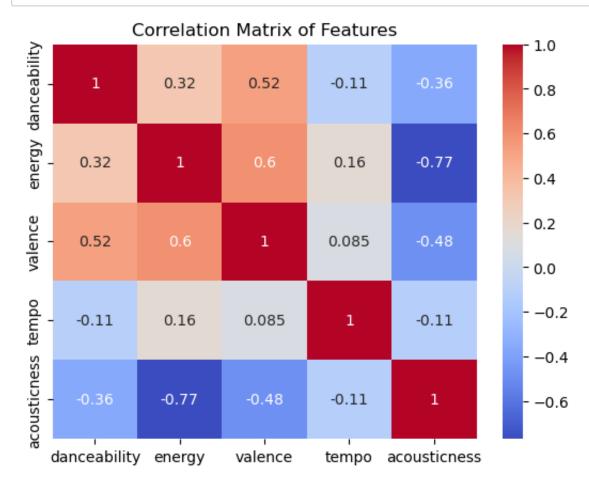
```
In [26]: import matplotlib.pyplot as plt
# Plotting the distribution of moods
df['mood'].value_counts().plot(kind='bar')
plt.title('Distribution of Moods')
plt.xlabel('Mood')
plt.ylabel('Frequency')
plt.show()
```



In [27]: # Visualizing distributions of numeric features
 features = ['danceability', 'energy', 'valence', 'tempo', 'acousticnes
 df[features].hist(bins=15, figsize=(10, 6), layout=(2, 3))
 plt.tight\_layout()
 plt.show()



## In [29]: import seaborn as sns # Heatmap of feature correlations sns.heatmap(df[features].corr(), annot=True, cmap='coolwarm') plt.title('Correlation Matrix of Features') plt.show()



```
In [31]: from sklearn.preprocessing import LabelEncoder

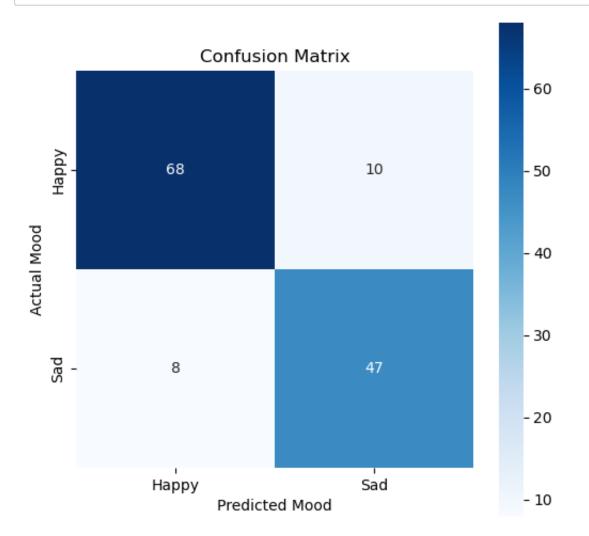
# Encoding the 'mood' column
df['mood_encoded'] = LabelEncoder().fit_transform(df['mood'])
```

In [32]: import pandas as pd
from sklearn.model\_selection import train\_test\_split
from sklearn.ensemble import RandomForestClassifier

```
In [33]: # Splitting the dataset
X_train, X_test, y_train, y_test = train_test_split(df[features], df['
# Initializing and training the Random Forest model
model_rf = RandomForestClassifier(n_estimators=100, random_state=42)
model_rf.fit(X_train, y_train)

# Making predictions
y_pred_rf = model_rf.predict(X_test)
```

> Accuracy of the Random Forest classifier: 0.8646616541353384 precision recall f1-score support 0.87 0.88 0 0.89 78 1 0.82 0.85 0.84 55 0.86 133 accuracy 0.86 0.86 0.86 133 macro avg weighted avg 0.86 0.86 133 0.87

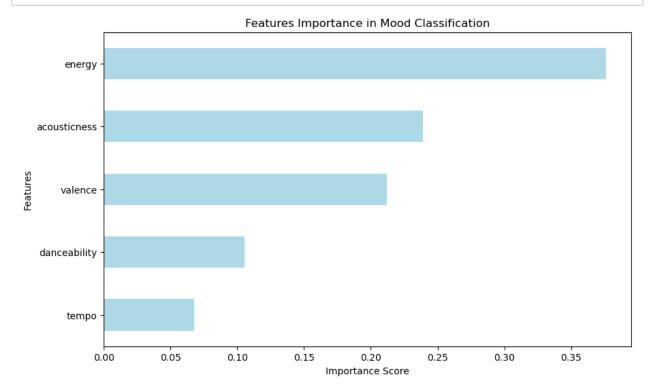


```
In [40]: # Extract feature importances
    feature_importances = model_rf.feature_importances_

# Create a pandas series to hold the feature names and their importance
importances = pd.Series(feature_importances, index=features)

# Sort the feature importances in descending order
importances_sorted = importances.sort_values()

# Visualize the feature importances
plt.figure(figsize=(10, 6))
importances_sorted.plot(kind='barh', color='lightblue')
plt.title('Features Importance in Mood Classification')
plt.xlabel('Importance Score')
plt.ylabel('Features')
plt.show()
```



```
In [ ]:
```