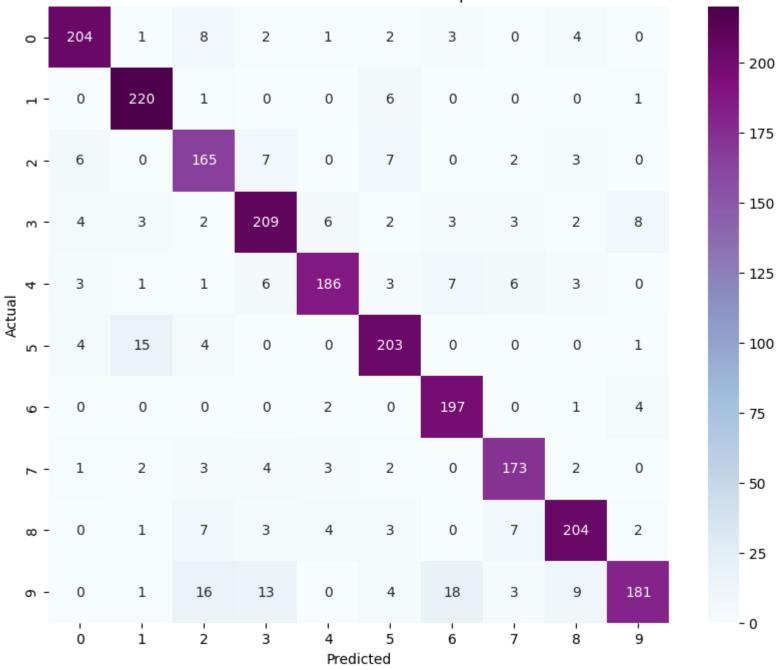
## **Project 1 - Music Genre Classification**

```
In [1]: ## Load necessary Libraries
        import numpy as np
        import os
        import pickle
        import random
        import operator
        import math
        import numpy as np
        from collections import defaultdict
        from python speech features import mfcc
        import scipy.io.wavfile as wav
        import librosa
        import pandas as pd
        import seaborn as sns
        from matplotlib import pyplot as plt
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.neighbors import KNeighborsClassifier
        from sklearn.model selection import train test split, cross val score, GridSearchCV
        from sklearn.metrics import confusion_matrix, accuracy_score
        from sklearn.preprocessing import StandardScaler
        from sklearn.utils import shuffle
In [2]: ## Load first CSV file into a DataFrame
        df1 = pd.read csv('U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/features 3 sec.csv')
In [3]: ## Load second CSV file into another DataFrame
        df2 = pd.read csv('U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/features 30 sec.csv')
```

```
In [4]: |## Merge both DataFrames
        merged df = pd.concat([df1, df2], ignore index=True)
        ## Remove 'filename' column
        merged df = merged df.drop(columns=['filename'])
        ## Split merged DataFrame into x and y, and drop label column
        X = merged df.drop(columns=['label'])
        y = merged df['label']
        ## Split data into training and testing sets
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
        ## Train Random Forest classifier
        clf = RandomForestClassifier(random state=42)
        clf.fit(X train, y train)
        ## Make predictions on test data
        y pred = clf.predict(X test)
        ## Calculate confusion matrix
        conf matrix = confusion matrix(y test, y pred)
        ## Create confusion heatmap matrix
        plt.figure(figsize=(10, 8))
        sns.heatmap(conf matrix, annot=True, fmt='d', cmap='BuPu')
        plt.xlabel('Predicted')
        plt.ylabel('Actual')
        plt.title('Confusion Matrix Heatmap')
        plt.show()
```

## Confusion Matrix Heatmap



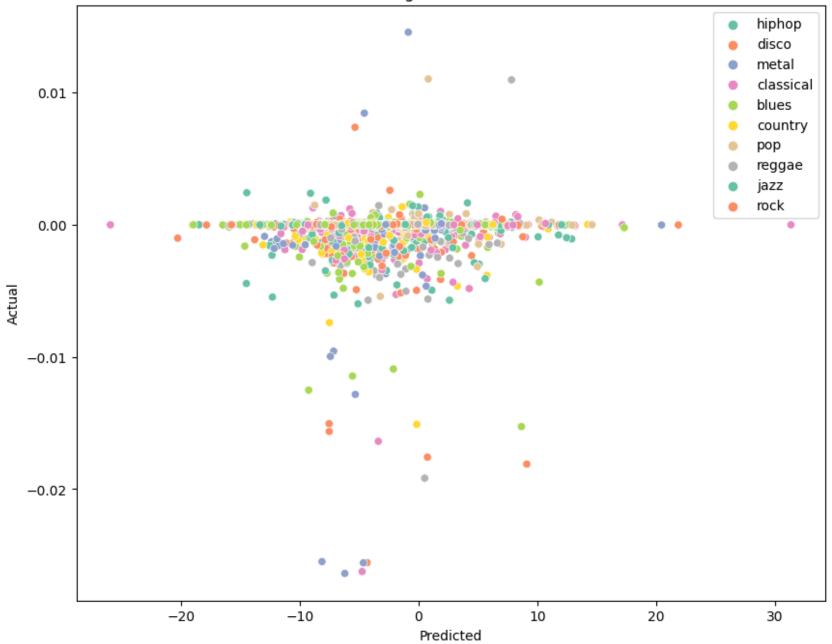
```
In [5]: ## Train KNN Classifier
mknn = KNeighborsClassifier(n_neighbors=5)
mknn.fit(X_train, y_train)

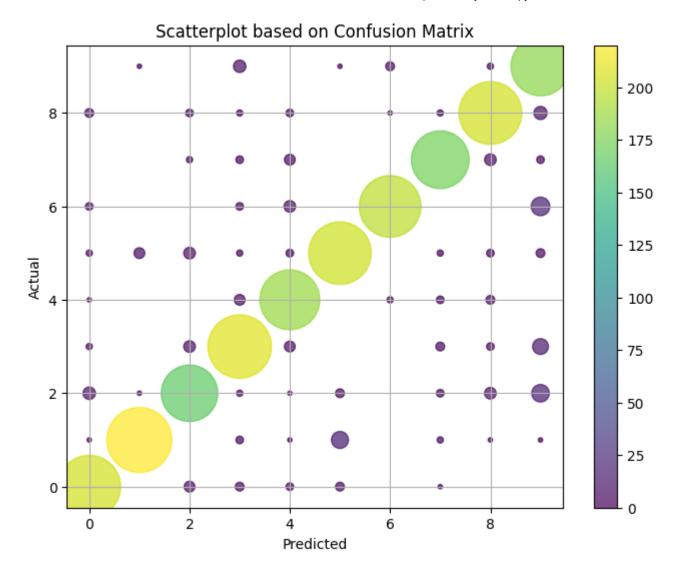
## Make predictions on test data
mknn_predictions = mknn.predict(X_test)

## Randomly choose two columns
column_indices = random.sample(range(X_test.shape[1]), 2)
x_column_index, y_column_index = column_indices

## Visualize KNN clustering from random columns chosen
plt.figure(figsize=(10, 8))
sns.scatterplot(x=X_test.iloc[:, x_column_index], y=X_test.iloc[:, y_column_index], hue=mknn_predictions, pale
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.title('KNN Clustering Plot (Random Columns)')
plt.show()
```

## KNN Clustering Plot (Random Columns)





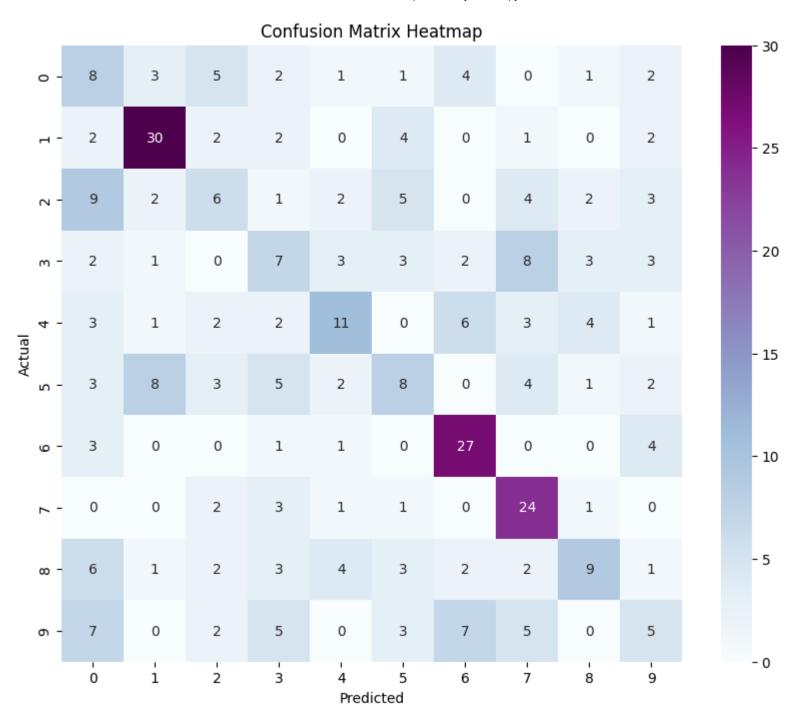
```
In [7]: ## Define function to extract features from an audio file
        def extract features(file path):
            y, sr = librosa.load(file path)
            mfcc feat = librosa.feature.mfcc(y=y, sr=sr, n mfcc=13)
            covariance = np.cov(mfcc feat)
            mean matrix = np.mean(mfcc feat, axis=1)
            return mean matrix, covariance
        ## Define function to load audio data and extract features
        def load and extract features(directory, split ratio=0.66):
            dataset = []
            for folder in os.listdir(directory):
                folder path = os.path.join(directory, folder)
                if os.path.isdir(folder path):
                    for file in os.listdir(folder path):
                        file path = os.path.join(folder path, file)
                        if file.endswith(".wav"):
                            try:
                                 (rate, sig) = wav.read(file path)
                            except Exception as e: ## Shows any .wav file with error
                                print(f"Error reading {file path}: {e}")
                                 continue
                            ## Extract MFCC features
                            mfcc feat = mfcc(sig, rate, winlen=0.020, appendEnergy=False)
                            ## Ensure all feature vectors have same dimensions
                            max dim = 13
                            if mfcc feat.shape[0] < max dim:</pre>
                                ## Pad with zeros to make dimensions consistent
                                padding = np.zeros((max_dim - mfcc_feat.shape[0], mfcc_feat.shape[1]))
                                mfcc_feat = np.vstack((mfcc_feat, padding))
                            elif mfcc feat.shape[0] > max dim:
                                ## Truncate to make dimensions consistent
                                mfcc feat = mfcc feat[:max dim, :]
                             covariance = np.cov(np.transpose(mfcc feat))
                            mean matrix = mfcc feat.mean(0)
                            feature = (mean matrix, covariance, folder)
                            dataset.append(feature)
            ## Shuffle dataset
```

```
random.shuffle(dataset)
    ## Split dataset into training and testing sets
    split index = int(split ratio * len(dataset))
    train set = dataset[:split index]
   test set = dataset[split index:]
   X train = np.array([data[0] for data in train set])
    y train = np.array([data[2] for data in train set])
   X test = np.array([data[0] for data in test set])
   y test = np.array([data[2] for data in test set])
    return X train, X test, y train, y test
## Define function to train and evaluate a model
def train and evaluate model(X train, X test, y train, y test):
    ## Standardize features
    scaler = StandardScaler()
   X train = scaler.fit transform(X train)
   X test = scaler.transform(X test)
    ## Train a Random Forest classifier
    clf = RandomForestClassifier(random_state=42)
    clf.fit(X train, y train)
    ## Evaluate model
   y pred = clf.predict(X test)
    accuracy = accuracy score(y test, y pred)
    conf matrix = confusion matrix(y test, y pred)
    return accuracy, conf matrix
## Main part of script
if __name__ == "__main__":
   ## Set directory path where .wav files are located
    directory = "U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres_original"
    ## Load audio data and extract features
   X train, X test, y train, y test = load and extract features(directory, split ratio=0.66)
    ## Train and evaluate model
    accuracy, conf matrix = train and evaluate model(X train, X test, y train, y test)
```

```
## Print accuracy and visualize confusion matrix
print(f"Accuracy: {accuracy:.2f}")

plt.figure(figsize=(10, 8))
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='BuPu')
plt.xlabel('Predicted')
plt.ylabel('Actual')
plt.title('Confusion Matrix Heatmap')
plt.show()
```

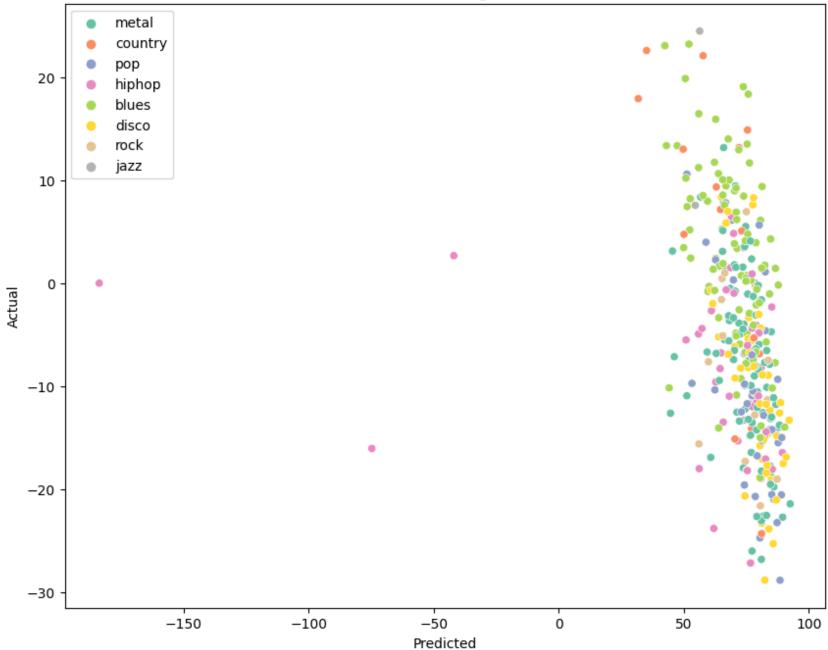
Error reading U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_original\jazz\jazz.00 054.wav: File format b'\xcb\x15\x1e\x16' not understood. Only 'RIFF' and 'RIFX' supported. Accuracy: 0.40



```
In [8]: | ## Define function to train and evaluate a KNN classifier
        def train and evaluate knn(X train, X test, y train, y test, n neighbors=5):
            ## Standardize features using Scaler
            scaler = StandardScaler()
            X train = scaler.fit transform(X train)
            X test = scaler.transform(X test)
            ## Train KNN classifier
            knn = KNeighborsClassifier(n_neighbors=n_neighbors)
            knn.fit(X train, y train)
            ## Evaluate model
            y pred = knn.predict(X test)
            accuracy = accuracy_score(y_test, y_pred)
            return knn, accuracy
        ## Main part of script
        if __name__ == "__main__":
            ## Set directory path where .wav files are located
            directory = "U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres_original"
            ## Load audio data and extract features
            X train, X test, y train, y test = load and extract features(directory, split ratio=0.66)
            ## Train and evaluate model
            knn, accuracy = train and evaluate knn(X train, X test, y train, y test)
            # Print accuracy
            print(f"KNN Accuracy: {accuracy:.2f}")
            # Visualize KNN clustering
            knn predictions = knn.predict(X test)
            plt.figure(figsize=(10, 8))
            sns.scatterplot(x=X_test[:, 0], y=X_test[:, 1], hue=knn_predictions, palette="Set2")
            plt.xlabel('Predicted')
            plt.ylabel('Actual')
            plt.title('KNN Clustering Plot')
            plt.show()
```

Error reading U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_original\jazz\jazz.00 054.wav: File format b'\xcb\x15\x1e\x16' not understood. Only 'RIFF' and 'RIFX' supported. KNN Accuracy: 0.31

## KNN Clustering Plot



```
In [9]: ## Add code to classify a random audio sample from the genres original folder
        results = defaultdict(int)
        i = 1
        for folder in os.listdir(directory):
            results[i] = folder
            i += 1
        ## List all .wav files in the genres original folder
        all wav files = []
        for folder in os.listdir(directory):
            folder path = os.path.join(directory, folder)
            if os.path.isdir(folder path):
                for file in os.listdir(folder path):
                    if file.endswith(".wav"):
                        all wav files.append(os.path.join(folder path, file))
        ## Choose 10 random .wav file for classification
        for _ in range(10):
            random audio file = random.choice(all wav files)
               ## extract genre name from file path
            genre name = os.path.basename(os.path.dirname(random audio file))
            ## print result
            print("Randomly selected audio file:", random audio file)
            print("Classified genre:", genre name)
```

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_original\metal.00009.wav

Classified genre: metal

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\country\country.00025.wav
Classified genre: country

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\disco\disco.00015.wav
Classified genre: disco

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\metal\metal.00094.wav
Classified genre: metal

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\blues\blues.00067.wav
Classified genre: blues

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\disco\disco.00006.wav
Classified genre: disco

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\hiphop\hiphop.00051.wav
Classified genre: hiphop

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\pop\pop.00003.wav
Classified genre: pop

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\classical\classical.00055.wav

Classified genre: classical

Randomly selected audio file: U:/School Homework/Fall 2023/DSC680-Applied Data Science/Project1/genres\_origi

nal\country\country.00074.wav
Classified genre: country