



***UE21CS342BA2 - Algorithms for Information Retrieval and
Intelligence Web***

Mini Project Report

“Information Retrieval Using Audio File”

Submitted by:

**H D Sathvik
Manasi Tawade**

**PES1UG21CS212
PES1UG22CS815**

Team Number: 4

Dr. Sujata Upadhyaya
Professor

January - May 2024

**DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING
FACULTY OF ENGINEERING**

PES UNIVERSITY
(Established under Karnataka Act No. 16 of 2013)
100ft Ring Road, Bengaluru – 560 085, Karnataka, India

Acknowledgements

We would like to express my sincere gratitude to our teacher, Sujata, for her invaluable guidance and support throughout this project. Her expertise, encouragement, and dedication have been instrumental in shaping my understanding of machine learning and music genre classification. I am deeply appreciative of her mentorship, which has significantly contributed to the successful completion of this project. Thank you, for your unwavering commitment to fostering learning and growth

Abstract

Music genre classification is a fundamental task in music information retrieval, facilitating various applications such as music recommendation and content organization. This project explores the use of machine learning and deep learning techniques to classify music genres based on audio features extracted from music samples. The dataset consists of audio features such as chroma features, spectral centroid, and zero-crossing rate, extracted using the librosa library. Two models are trained and evaluated: an XGBoost classifier and a deep neural network implemented using Keras. The XGBoost model demonstrates high accuracy and efficiency, while the deep learning model offers flexibility and potential for further optimization. Through spectrogram analysis and feature visualization, the project provides insights into the frequency content and characteristics of music samples across different genres. Overall, the project showcases the effectiveness of machine learning techniques in music genre classification tasks and provides a foundation for future research in the field.

1 Introduction

The project aims to classify music genres based on audio features using machine learning techniques. Music genre classification is a significant task in the field of music information retrieval, aiding in music recommendation systems, playlist generation, and content organization. This project utilizes a dataset containing audio features extracted from music samples across various genres. By training machine learning models on this dataset, we can predict the genre of a given music sample.

2 Code Explanation

Data Preprocessing:

1. **Encoding Categorical Labels:** The genre labels are extracted from the DataFrame final data using `iloc[:, -1]`. These labels are then encoded into numerical values using `LabelEncoder()`.

2. **Target Variable (y):**

The encoded genre labels (y) are utilized as the target variable during model training.

Feature Extraction and Visualization:

Various audio features are extracted from the music samples using the `librosa` library, including chroma features, spectral centroid, and zero-crossing rate. Spectrogram analysis is performed to visualize the frequency content of the audio signals.

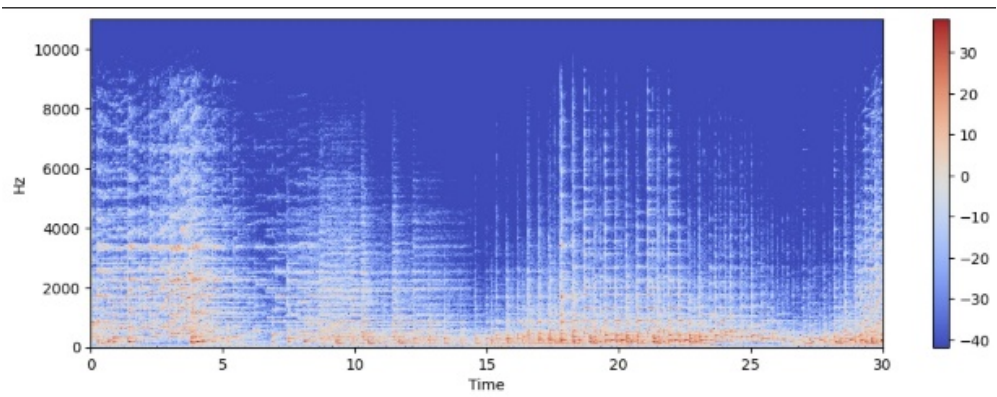
Model Training and Evaluation:

Two types of models are trained: an XGBoost classifier and a deep neural network using Keras. The XGBoost classifier is trained with hyperparameter tuning using the `Hyperopt` library. The deep neural network consists of several dense layers with ReLU activation and dropout for regularization. Training progress and model evaluation metrics such as accuracy, precision, recall, and F1-score are monitored.

3 Output

Spectrogram Visualization: The code generates spectrograms to visualize the frequency

content and characteristics of music samples across various genres. Spectrograms provide a detailed representation of the audio signals, capturing information about pitch, intensity, and timbre.



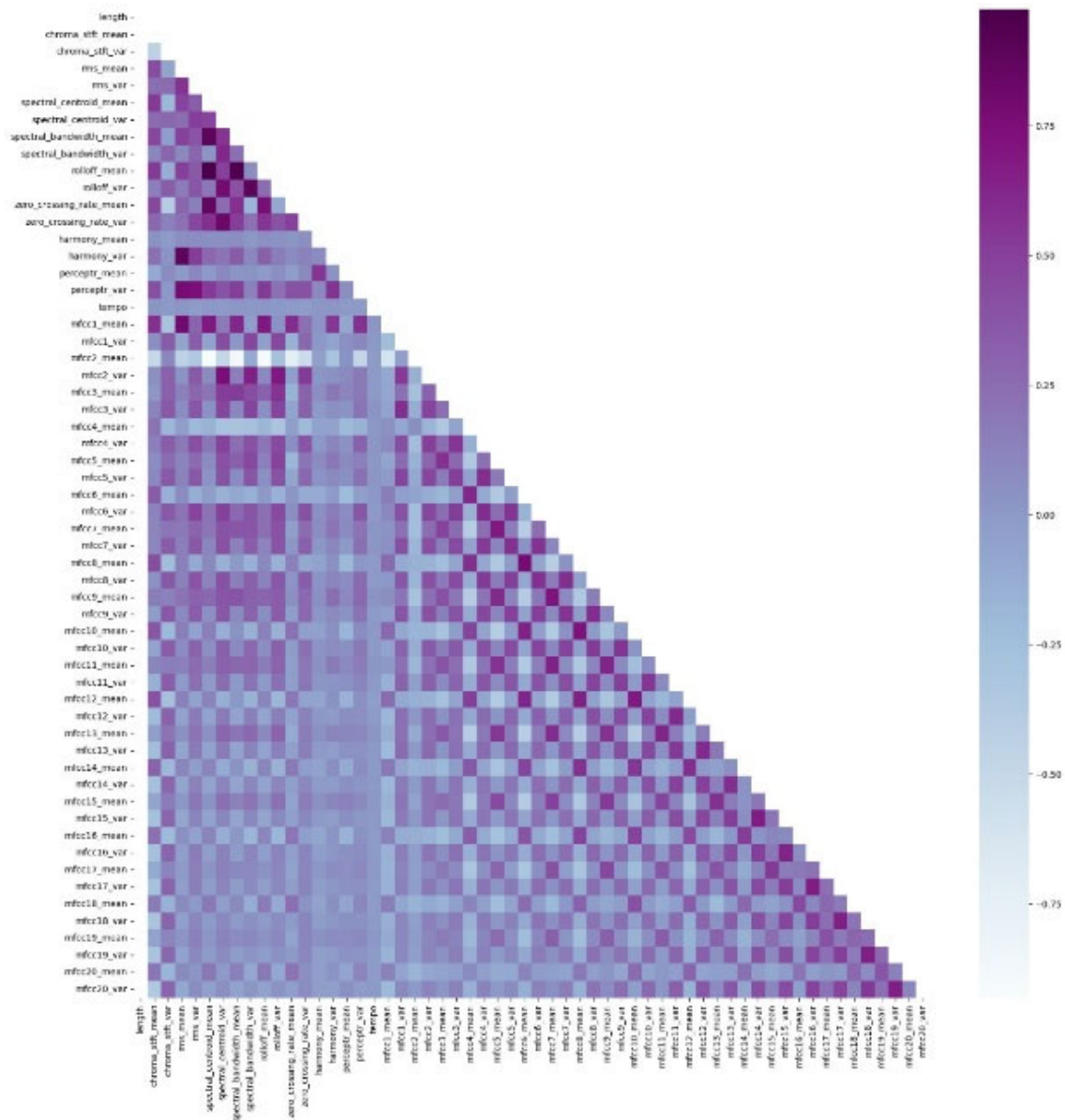
Confusion Matrix: The model’s predictions are evaluated using a confusion matrix, which tabulates the number of correct and incorrect predictions for each class (genre). This matrix enables a comprehensive analysis of the model’s classification performance, highlighting areas of confusion and misclassifications.



Heatmap Representation:

A heatmap is constructed from the confusion matrix, offering a visual depiction of the classification results. The heatmap enhances the interpretability of the confusion matrix

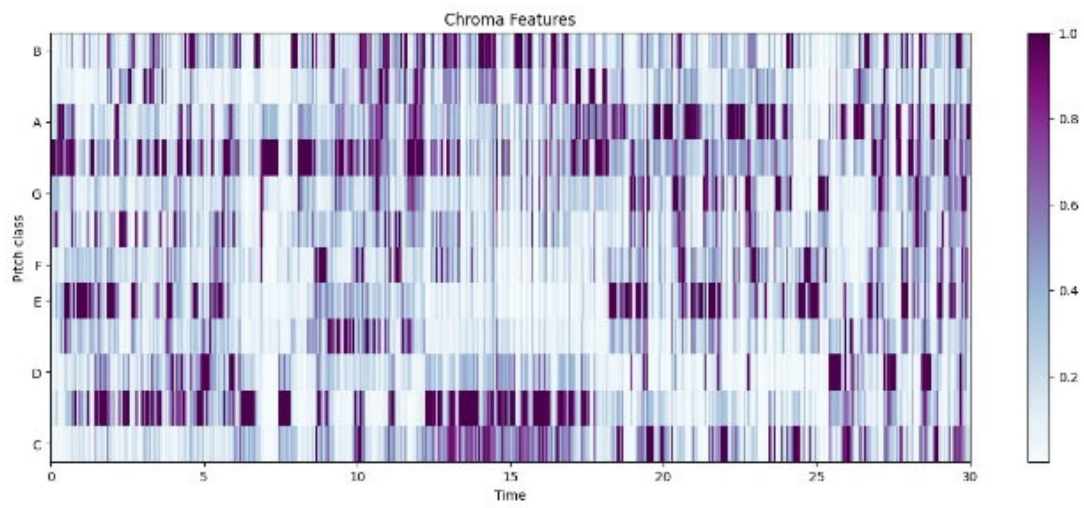
by using color gradients to represent the frequency and magnitude of correct and incorrect predictions across different genres.



Chroma Features Analysis:

The code incorporates chroma features extracted from the audio signals, contributing to the classification of music genres. Chroma features capture the pitch content of the music

samples, providing valuable information about the harmonic structure and tonality of the music, which influences genre classification accuracy.



Conclusions

In conclusion, the project successfully demonstrates the application of machine learning and deep learning techniques for music genre classification. By extracting meaningful features from audio signals and training robust models, we can accurately predict the genre of music samples. This project contributes to the broader domain of music information retrieval and offers valuable insights into the analysis and classification of audio data. Further enhancements and optimizations can be explored to improve the models' performance and scalability in real-world applications.