



CHRIST
(DEEMED TO BE UNIVERSITY)
BANGALORE, INDIA

AUDIO GENRE CLASSIFICATION

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MISSION

CHRIST is a nurturing ground for an individual's holistic development to make effective contribution to the society in a dynamic environment

VISION

Excellence and Service

CORE VALUES

Faith in God | Moral Uprightness
Love of Fellow Beings
Social Responsibility | Pursuit of Excellence

Agenda

- Introduction to Audio Genre Classification
- Dataset
- Features
- Algorithms Implemented
- Comparing Results
- Conclusion

Introduction to Audio Genre Classification

- What is it?

Audio genre classification uses machine learning to automatically categorize music based on its sonic characteristics.

- How does it work?

1. Feature extraction: Audio signals are converted into numerical data describing rhythm, tempo, timbre, and other musical elements.

2. Model training: Machine learning algorithms learn patterns from labeled audio samples across various genres.

3. Prediction: New audio is analyzed, and its genre is predicted based on the trained model.

Dataset - GTZAN

- The GTZAN dataset is the most-used public dataset for evaluation in machine listening research for music genre recognition (MGR).

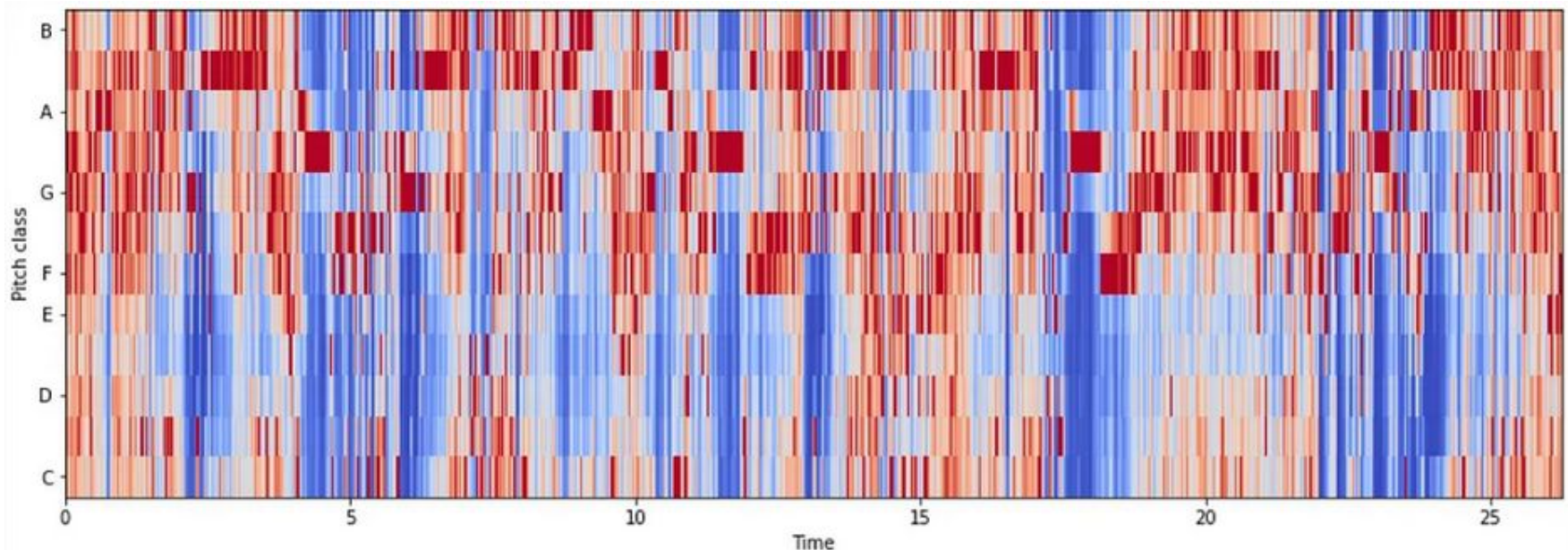
The dataset has the following types of data:

- **genres original** - A collection of 10 genres with 100 audio files each.
- **images original** - A visual representation (Mel Spectrograms) for each audio file.
- **2 CSV files** - Containing features of the audio files. One file has features of 30 sec clips, the other has features of 3 sec clips (i.e., the second CSV has 10 times more data than the first)

Features

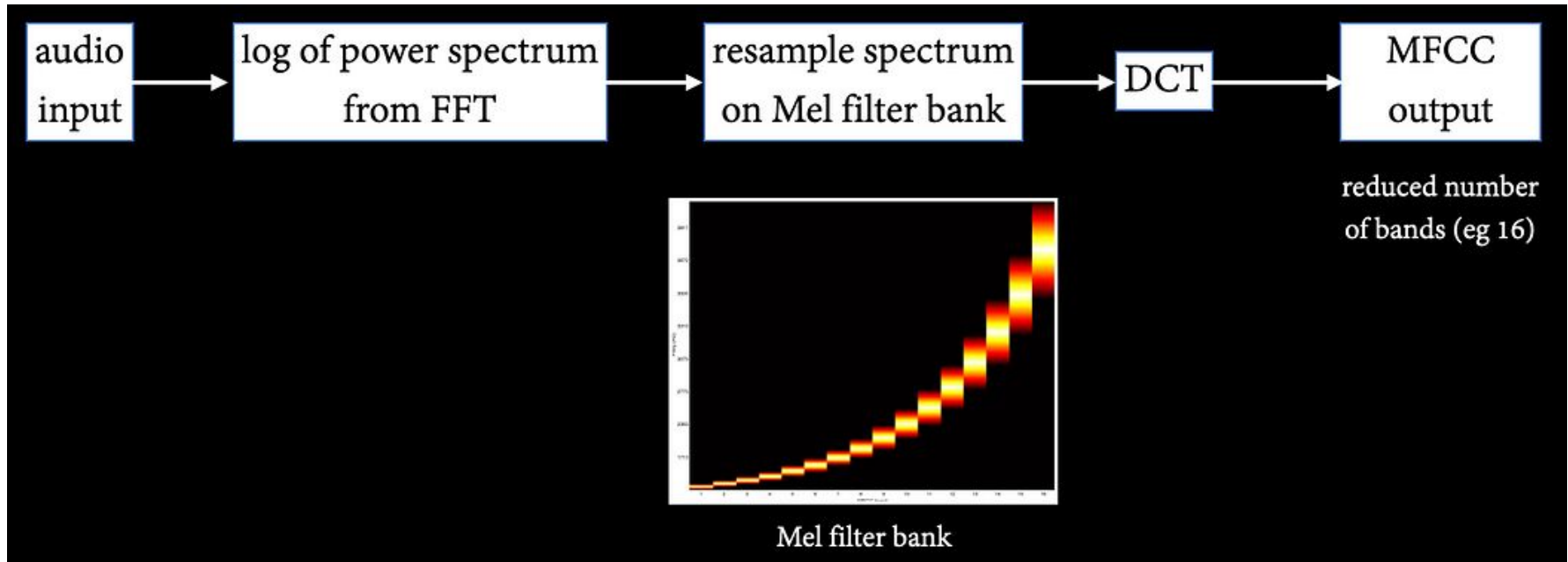
- **Chroma Frequencies**

It is an effective method for analyzing music where pitch can be adequately categorized (generally into twelve categories) and whose intonation approximates the equitable scale.



- Mel Frequency Cepstral Coefficients

The mel frequency cepstral coefficients (MFCCs) of an audio signal are a small set of features (usually about 10–20) which describe the overall shape of the spectral envelope.



- **Spectral Rolloff**

The spectral rolloff is a parameter that describes the characteristics of a signal's shape. It indicates the frequency at which the high frequencies start to diminish and eventually reach zero.

- **Spectral Bandwidth**

Spectral bandwidth is a measure of the range of frequencies present in a signal.

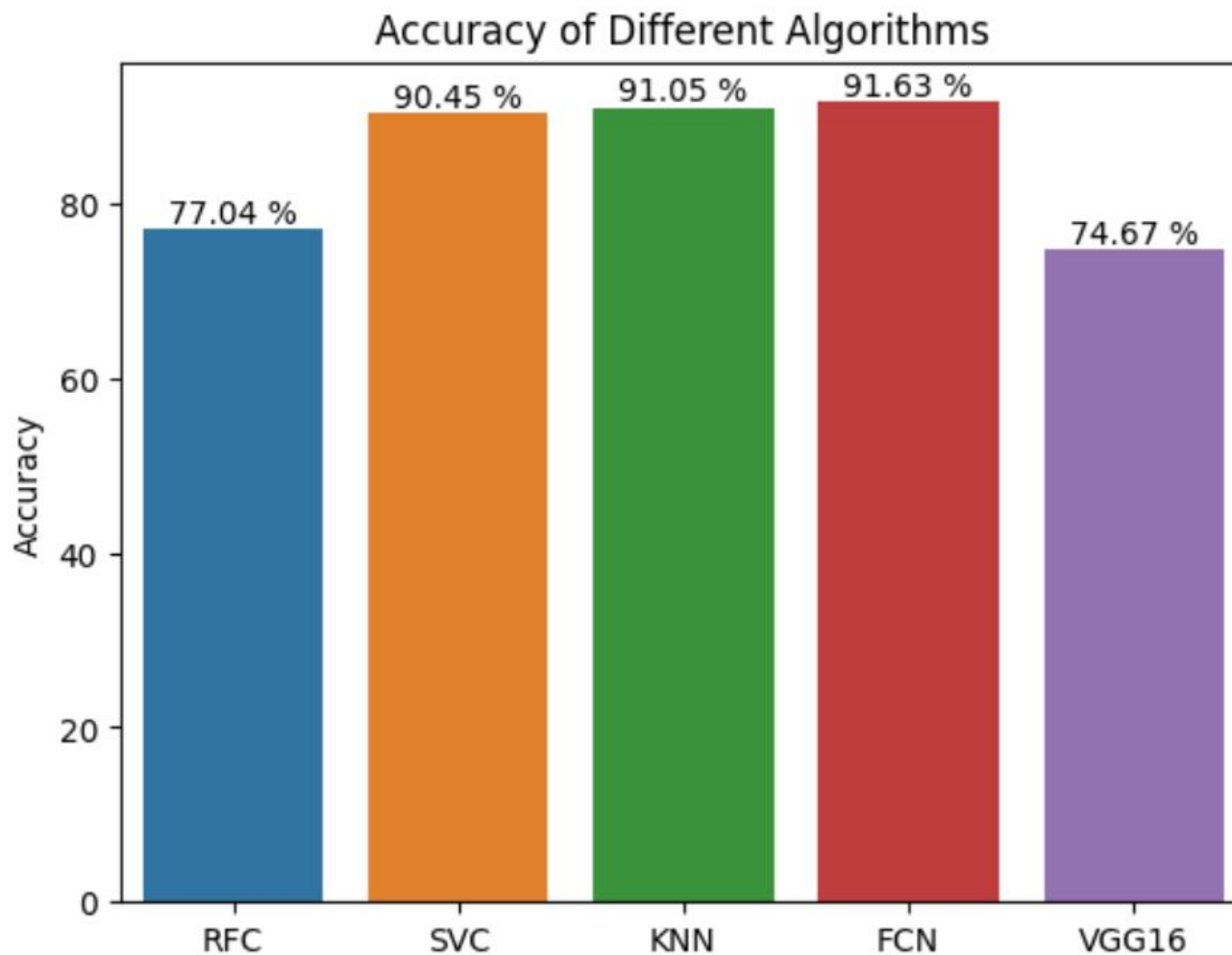
- **Zero Crossing Rate**

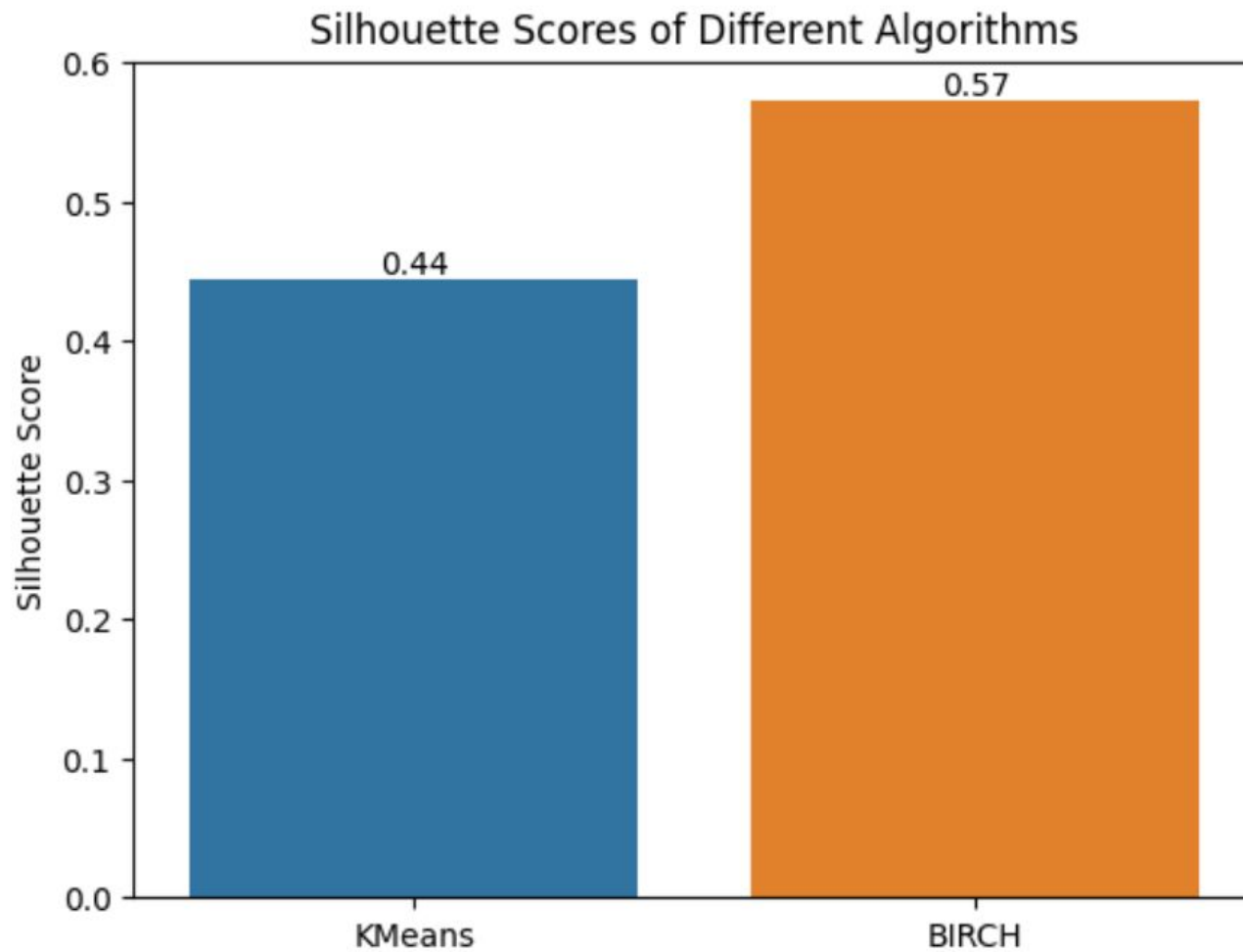
The term refers to the speed at which a signal transitions from a positive value to zero, then to a negative value, or from a negative value to zero, and then to a positive value.

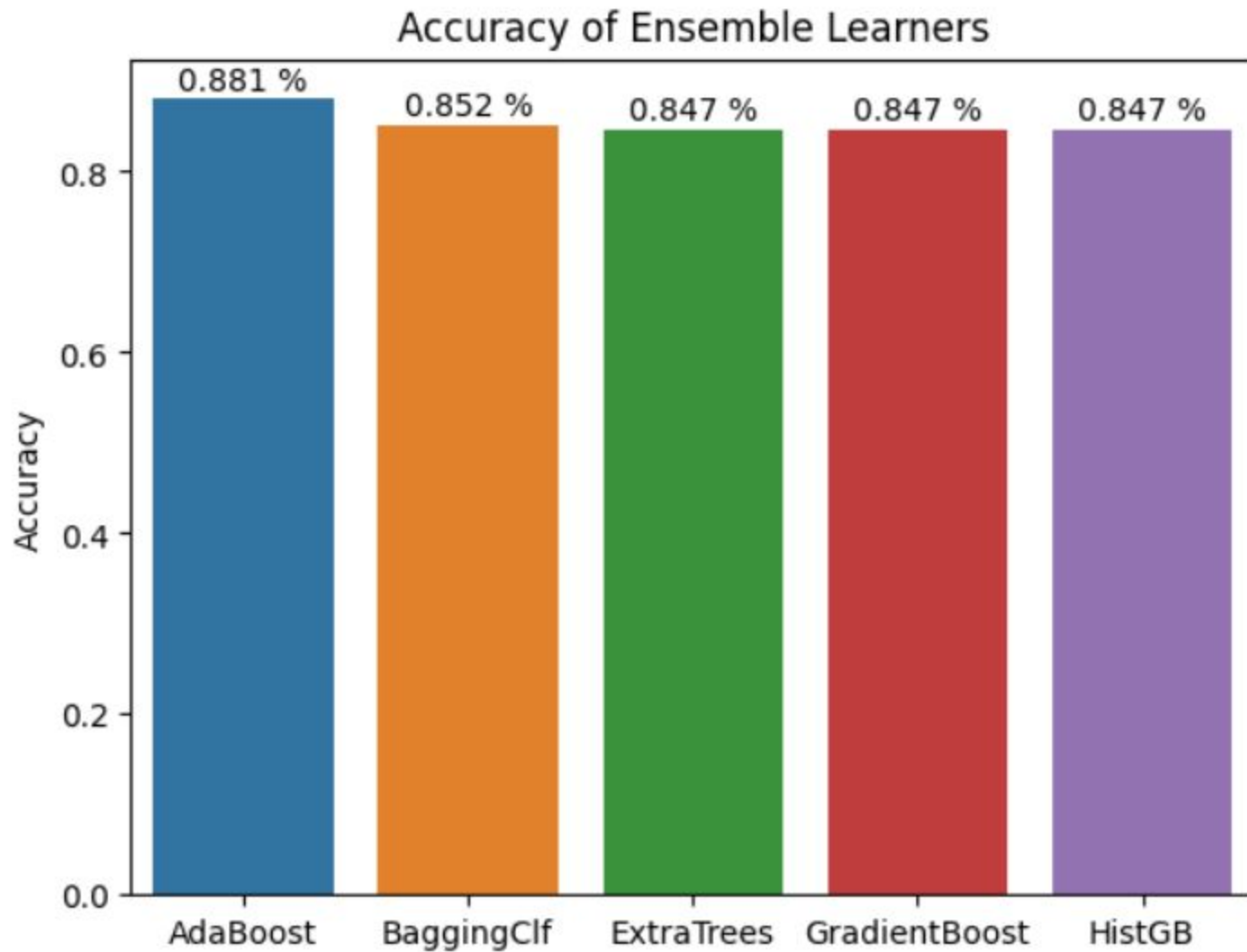
Algorithms Implemented

- **Unsupervised Learning**
- **Dimensionality Reduction**
- **Supervised Learning**
- **Deep Learning**

Comparing Results







Conclusion

The models that performed the best included Support Vector Classifier, K-Nearest Neighbours, and Fully Connected Network.

Moreover, the models can be fine-tuned if one has access to higher computational power to yield results that are state-of-the-art.

Audio Genre Classification is a sub-problem that occurs while performing various different tasks. With the help of machine learning, we can now classify songs as belonging to a particular genre without the need of human interruption.