Machine Learning:

Music Genre Classification





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01 - Introduction



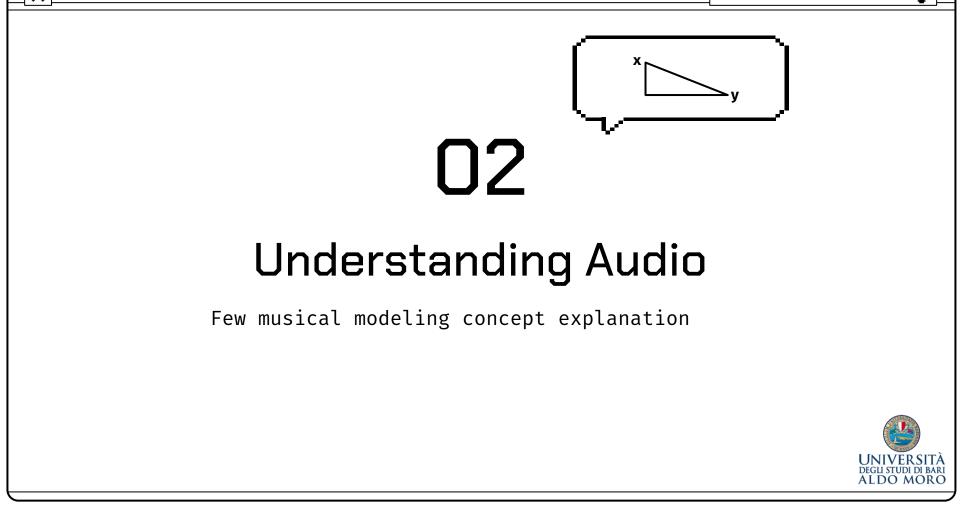
A music genre is a style or a type of music. There are numerous music genres, such as hip-hop, rock, country, pop, etc.

The goal of this project is to build a predictive machine learning model to classify the genre of a given song.









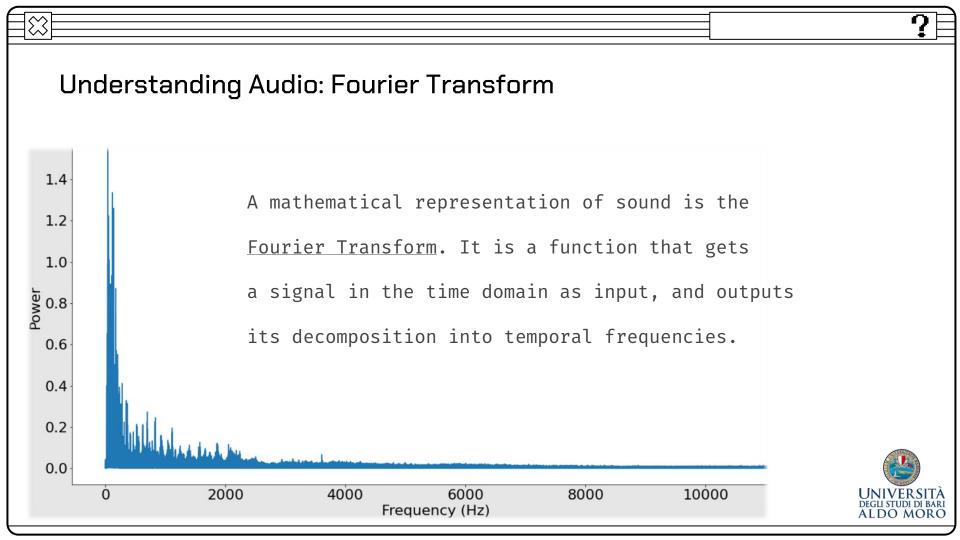
Understanding Audio

Sound is defined as vibrations that travel through the air or another medium as an audible mechanical wave

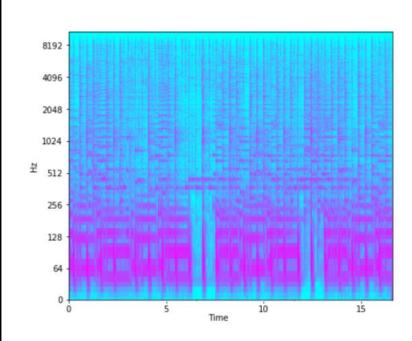


SOUND WAVE





Understanding Audio: Spectrogram

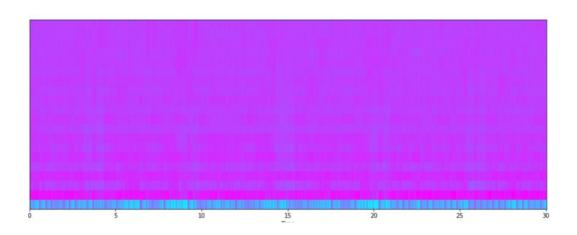


Then we have the <u>Spectrogram</u> which is a visual representation of the spectrum of frequencies of a signal as it varies with time.



Understanding Audio: MFCCs

The Mel frequency cepstral coefficients (MFCCs) of a signal are a small set of features (usually about 10-20) which concisely describe the overall shape of a spectrogram envelope.







03



DATASET

The data used in this project

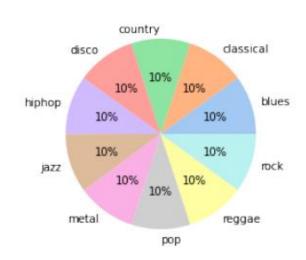


Dataset: GTZAN

The original GTZAN dataset consists of:

- 1000 audio files
- 10 genre classes (disco, classical, ...)
- Each song is 30 seconds in duration
- Each song is described by 60 features

In order to have more entries in the dataset, each song is split in 10 songs of 3 seconds each. In this way, the dataset has been expanded to 10,000 songs.









The MFCCs dataset is inspired by the GTZAN dataset and it tries to represent a song by using only information present in Mel Frequency Cepstral Coefficients (MFCCs) in order to isolate a representative time series information.

It is created by analysing the 3-seconds GTZAN songs and thanks to librosa python package it was possible to extract 13 set of features (MFCCs) per song.

The overall size of the two datasets are the same: 10000 entries.





04

Experiments

Experiments conducted



Enviroment

- Google Colab Pro
- GPU Enviroment
- Scikit-Learn
- Keras
- Librosa



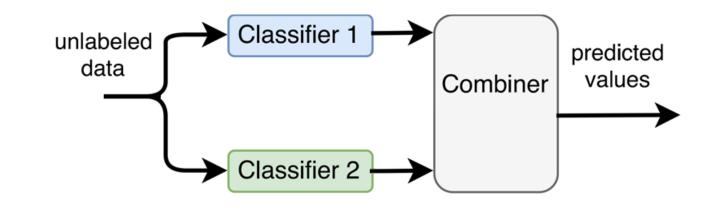




Main Idea 🖀

The main idea is to train two different models, one per dataset and combine the results of this two classifiers as the final model.

The two chosen classifiers are the best performing models, based on the accuracy on the test set.



Loss Functions

Sparse Categorical Cross-entropy

It is used when class labels are mutually exclusive for each data, meaning each data entry can only belong to one class

$$J(w) = \sum_{i=1}^{size_o utput} y_i * \log \hat{y_i}$$

- y_i is the true label
- \hat{y}_i is the predicted label



GTZAN Experiments Overview

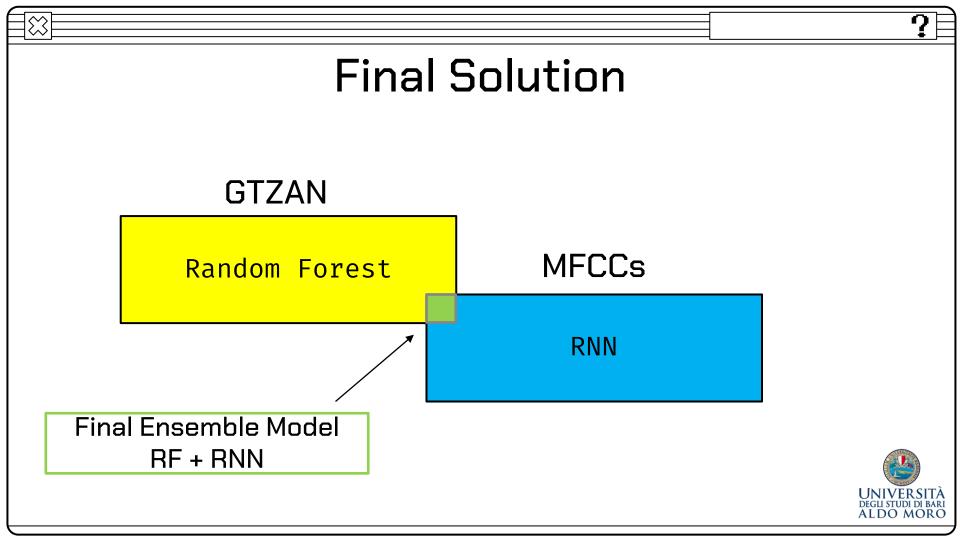


Naive Bayes		42%
Nearest Neighbors		28%
Logistic Regression	max_iter = 10000	45%
Random Forest	n_estimatorsint, default=100	86%
Ada Boost	n_estimators = 100	40%
Gradient Boosting	n_estimatorsint, default=100	82%
Decision Tree		60%
Support Vector Classifier	max_iter = 10000	23%
MLP	hidden_layer_sizes=(128, 64, 8), max_iter=250, learning_rate_init=0.000001	11%
FNN	Dense (32, 16, 10), epoch = 100 relu on dense layer, the last one with SoftMax	81%

MFCCs Experiments Overview

FNN	Dense128, Dropout 0.4, Dense 64, Dense 32, Dense 16, Dense 10	42%
CNN	Conv 2D 8, Maxpooling (2,2), Conv 2D 32, Maxpooling (2,2), Flatten, Dense 128, Dense 64, Dense 10	63%
RNN	LSTM 64, LSTM 64, Dense 64, Dropout 0.3, Dense 10	89%







Thanks!



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