Music Genre Classification

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Abstract

Music genre classification has taken on a new significance in recent years as a result of the significant expansion in the accessibility of music data. We need to correctly index them in order to have better access to them. When working with a large collection of music, using an automatic music genre classification is essential. Researchers have favored machine learning methods for the majority of contemporary music genre classification methods. We used two distinct genre-specific datasets in this investigation. The system is trained and categorized with the help of a Deep Learning technique. Classification and training are carried out with a convolution neural network. The feature extraction step is the most important part of speech analysis. The primary method for extracting audio features is the Mel Frequency Cepstral Coefficient (MFCC). The proposed method divides music into several genres by extracting the feature vector. Based on our findings, our system has an accuracy level of 80%, which will make music genre classification much easier and significantly improve with additional training. The term "music" is a common way to classify the various kinds of music. It must be distinguished from other types of music. There are many different ways that music can be divided into genres. Pop, classical, rock, and other genres make up the music. In the field of music information retrieval, the most difficult task is to classify music by genre. In order to obtain music from a large collection, automatic music genre classification is essential. It has applications in the real world, such as automatically tagging unknown music (which is useful for apps like savaan and wynk etc.)

1. Introduction

With a favorable monsoon, the secondary agro-based market in India, which is an agricultural nation, will remain steady. The length of each year's monsoon depends on how long it rains. A bad monsoon can destroy crops and cause a shortage of some agricultural products, which can lead to food inflation, insecurity, and public unrest. In our analysis, we attempt to comprehend how rainfall in India has changed over time, broken down into months and various subdivisions. Nowadays, rain falls throughout the year, particularly during torrential storms. In addition, one of the primary goals of research on climate change is to determine whether or not extreme shifts in the frequency and occurrence of heavy rainfall are occurring. In this case, we made use of machine learning classification algorithms like SVM, Random Forest, Logistic Regression, and Decision Tree. In hydrological studies, one of the most important concerns has been how accurate the ML models used to predict rainfall based on historical data are. In order to aid in the prevention of destruction and natural disasters, an

accurate ML model could provide early warnings of severe weather. As a result, ML algorithms that can accurately predict rainfall and reduce the error in the climate change model's projected rainfall data set must be developed. Machine learning is a subset of artificial intelligence (AI). Instead of being explicitly programmed to do so, it focuses on teaching computers to learn from data and improve with experience. Algorithms in machine learning are taught to analyze large data sets for patterns and correlations in order to make the best decisions and predictions. Applications for machine learning get better with use and become more accurate with more data. Machine learning is used in everything from our homes to shopping carts to entertainment media to healthcare. It is unquestionably one of the most exciting subfields of artificial intelligence. By providing the machine with specific inputs, it completes the task of learning from data. Understanding what makes Machine Learning work and how it can be used in the future is essential.

The first step in the machine learning process is to feed training data into the algorithm of your choice. The final machine learning algorithm will be developed using either known or unknown training data. The algorithm does have an effect on the type of input training data, and this idea will be discussed more in a moment. The machine learning algorithm is tested to see if it works correctly with new input data. The predictions and outcomes are then compared to one another.

The algorithm is repeatedly retrained until the data scientist achieves the desired result if the prediction and results do not match. This makes it possible for the machine learning algorithm to continuously learn on its own and come up with the best answer, with accuracy gradually increasing over time.



Fig.1 Supervised Machine Learning

2. Literature Review

Anaconda is a scientific computing distribution of the Python and R programming languages that aims to make package management and deployment simpler. Packages for data science that are suitable for Windows, Linux, and macOS are included in the distribution. Anaconda, Inc., which was established in 2012 by Peter Wang and Travis Oliphant, is responsible for its development and upkeep. It is also known as Anaconda Distribution or Anaconda Individual Edition because it is a product of Anaconda, Inc. Other products from the company include Anaconda Team Edition and Anaconda Enterprise Edition, neither of which are free.

The package management system conda is in charge of managing Anaconda's package versions. Because it proved to be useful on its own and for applications other than Python, this package manager was released as its own open-source package. MiniConda is a smaller, bootstrap version of Anaconda that only includes Python, Conda, the packages that they depend on, and a few other packages.

In addition to the conda package and virtual environment manager and over 7,500 additional open-source packages that can be installed from PyPI, the Anaconda distribution comes preinstalled with over 250 packages. Anaconda Navigator, a desktop graphical user interface (GUI) included in the Anaconda distribution, is a graphical alternative to the command line interface (CLI) that enables users to launch applications and manage conda packages, environments, and channels without having to use command-line commands. The default installation of Anaconda2 includes Python 2.7, while Anaconda3 includes Python 3.7. Navigator is able to find packages on Anaconda Cloud or in a local Anaconda Repository, install them in an environment, run the packages, and update them all at the same time. It is compatible with Linux, macOS, and Windows.

For data science and machine learning-related applications, Anaconda Navigator is a free and open-source distribution of the Python and R programming languages. Conda is a cross-platform, open-source package management system that can be installed on Windows, Linux, and macOS. Tools like JupyterLab, Jupiter Notebook, QtConsole, Spider, Glueviz, Orange, Rstudio, and Visual Studio Code are included in Anaconda. We will use Jupiter notebook and spyder for this project. Any Python module has a name that is the base name of the file or the module's name property without the.py extension. A package is a directory of Python modules that includes an additional init.py file. A package is a collection of Python modules. A package is distinguished from a directory that just happens to contain a number of Python scripts by the init.py file. As long as each directory contains its own init.py file, packages can be nested to any depth.

The Python object that Python creates in response to an import of a module or package is always of type module. This indicates that only the file system level distinguishes between package and module. However, keep in mind that when you import a package, sub-packages and modules are not visible, only the variables, functions, and classes in the init.py file.

When adding packages to our Anaconda environment, one common method is to use the "Anaconda Navigator." From the Environments tab, which is located just below the Home tab, we can determine which packages are installed and which are not. Using the Anaconda

Navigator, installing any package is simple: simply search for the necessary package and select it.



3. Proposed System

Supervised learning is a type of machine learning in which machines predict the output based on well-labelled training data and are trained using that data. Some input data already carries the appropriate output tag, as indicated by the labelled data.

The process of providing the machine learning model with both accurate input data and output data is referred to as supervised learning. Finding a mapping function to map the input variable (x) to the output variable (y) is the goal of a supervised learning algorithm. The process of recognizing, comprehending, and organizing concepts and things into predetermined categories or "sub-populations" is known as classification. Machine learning programs use a variety of algorithms to categorize future datasets using pre-categorized training datasets.

When the output variable is categorical, such as Yes-No, Male-Female, True-False, etc., classification algorithms are utilized. Classification is a type of "pattern recognition" in which classification algorithms are applied to the training data in order to locate the same pattern in subsequent data sets. Data augmentation is a method for using your dataset's existing data to expand or enlarge it. In order to assist in better training our model with a large dataset, we employ a variety of methods to expand our dataset. if you use a small dataset to train your model and overfit the data, or if you have a small dataset. Therefore, in order to train our model and improve its capabilities and performance, or in order to make it more applicable to other situations, we require a suitable dataset. This is made possible by Data Augmentation. The input data will go through a series of transformations as part of the data augmentation process. As a result of the variations in the data samples, our dataset will become more rich.



Fig.3 Code Of The Project



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5.Conclusion

Various machine learning algorithms are used to classify music genres, which is a complicated but compatible process that enables software applications to classify millions of songs using schemas created by pre-data sets. We started the project with a basic setup and used MFCC to extract features from audio files. Next, we built a KNN and bagging classifier from scratch that finds k nearest neighbors based on features and the maximum neighbors in a specific class as an output. The amplitude and frequency, which change rapidly, are the most important characteristics for identifying and segmenting the audio. Using librosa, we can easily plot the audio frequency wave in the form and amplitude of a wave plot. By dividing the audio into frames, MFCC removes vocal excitation, separates extracted features, adjusts the loudness and frequency of the sound in accordance with humans, and records the context.

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