

UNIVERSITATEA "ALEXANDRU-IOAN CUZA" DIN IAȘI

FACULTATEA DE INFORMATICĂ



LUCRARE DE LICENȚĂ

**Music Recognition Using Convolutional Neural
Networks**

propusă de

Andrei Vavilov

Sesiunea: Iulie, 2021

Coordonator științific

Conf. Dr. Vitcu Anca

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Motivation

In the age of big data and immense computational power, artificial intelligence has come to be the new standard in the computer science field. Various types of data can be understood, learnt, predicted and even produced by a well-tuned neuronal network, making the principles of machine learning a must for a scientist nowadays.

Applications of neuronal networks can be found in any discipline: from medicine to physics, social sciences and languages. The purpose of this thesis is to depict how artificial intelligence can find it's place and purpose in a previously profoundly human field: art.

Introduction

The current thesis is an attempt to materialize the intersection of artificial intelligence and arts, especially, music.

The application consists of two major parts:

- First, the machine learning component, represented by a neuronal network framework, implemented from scratch (i.e. without the explicit support of existent frameworks or libraries). The module presents the necessary functionalities for constructing a neuronal network: layers, activation functions, metrics, optimizers, models and support for data generation and preprocessing.
- Second, the visual support, compound of two illustrative animations, created in Autodesk Maya 2019, and a Graphical User Interface, in which a hypothetical user can interact with the application by feeding it an YouTube link of a song. As a result, the application will decide whether the input song is played on a piano or not (prediction Piano/Other). Depending on the decision of the neuronal network, the corresponding Maya animation will be played.

0.1 Context

The first analogy between the way computers can process information and the way the human brain works (as we know, at least, to this day), has been made by Warren McCullough and Walter Pitts in 1944, who later became the founding members of what is sometimes called the first cognitive science department [5]. The primary idea

is elegant in its simplicity: the neuron, the basis of the human cognitive apparatus, can be modelled in machine learning as an unit in a network.

Since then, numerous studies and breakthroughs have been made, as well as various frameworks and tools which make implementing a neuronal network accessible without possessing the full mathematical background needed prior.

0.2 Purpose and modus operandi

One of the purposes of this thesis is to implement the functionalities and the logic behind a neuronal network *ab initio*, as well as creating and fine-tuning a model. In order to complete this task, we consulted multiple sources(e.g. [6],[10],[8]) which presented the theoretical and mathematical aspects of constructing the aforementioned classifier. The implementation was constructed with the support of various tools from the Python programming language (e.g. Numpy, Librosa, Tensorflow Keras, Kapre), which will be extensively discussed in the following chapters.

As discussed before, the second component of the thesis regards the visual part of the application. For obtaining a interactive use of the project, we used Autodesk Maya 2019 (for creating the animations) and Python tkinker for creating a simplistic GUI.

The logical flow of the pipeline is as follows: the user feeds an YouTube link of a song to the GUI. The model (previously trained and saved) computes a prediction regarding the category under which the input falls (Piano/Other). Given this result, the corresponding animation is played.

0.3 Personal contributions

Numerous attempts of creating a medium between artificial intelligence and other disciplines have been made since the rise of this field. Arts, especially music, is no exception.

The particularity of the current thesis is the approach we had in completing the task: implementing from scratch the neural network framework, and, implicitly, understanding the mathematical and theoretical subtleties of it, as well as creating the visual aid which aims to touch on (although briefly) 3D animations.

0.4 Structure

The structure of the present thesis follows the major constituent parts described before and is as follows:

1. Chapter One

- Similar applications : in which other akin projects are mentioned;

2. Chapter Two

- Technologies used : in which technologies needed for creating the application and their purpose and functionalities are discussed;

3. Chapter Three

- Architecture and implementation : in which the actual implementation is discussed explicitly. This section contains technical and theoretical aspects of the thesis and showcases relevant code;

4. Chapter Four

- Use cases: in which the functionalities of the project are presented.

0.5 Acknowledgements

I would like to express my special thanks of gratitude to the academical staff of the Faculty of Computer Science Iași for the opportunity to do this project.

Obviously, the present thesis would not be possible without the help and guidance of the professor that directed it, PhD. Anca Vitcu.

Chapter 1

Similar applications

Neural Network Models as well as Machine Learning algorithms provide various versatile and adaptive methods for non-linear problem solving. Because of these reasons, there are numerous applications which perform the task of musical/audio classification using the aforementioned techniques. We will discuss in the following sections about three of those applications.

1.1 Spektrum

Spektrum is a multi-platform music genre classifier and music recommendation system developed in the context of the course "Application Challenges for Machine Learning on the example of IBM Power AI", by the team consisting of Marte Vinje, Moritz Klimmek, Thomas Salzer, Aaron Hümmecke, Lukas Vorwerk [9]. The application consists of two parts:

- Music Genre Classification: Performed by a Convolutional Neural Network Model which analyzes the MEL spectrogram of a given input song.
- Music Recommendation: Generating suggestions by making use of a combination of collaborative filtering and content based filtering.

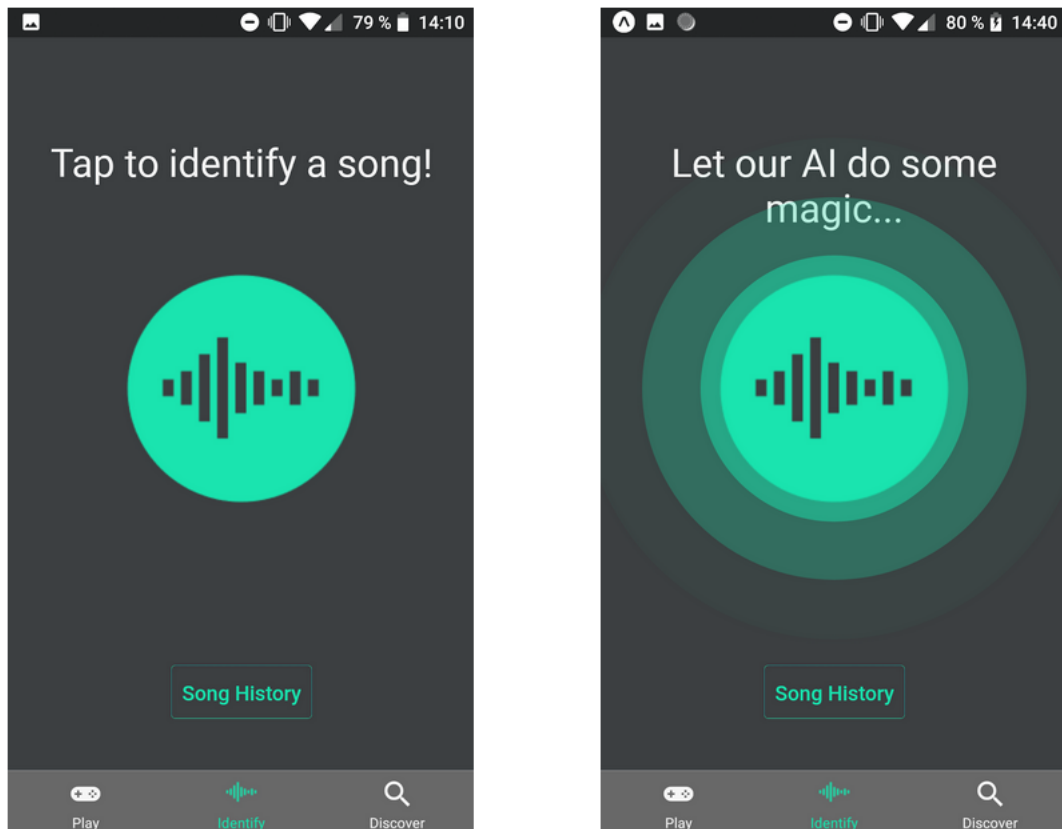


Figure 1.1: The interface of the Spektrum application

1.2 MuseNet

MuseNet is a Deep Neural Network Framework developed by OpenAI that can generate 4-minute snippets of original musical composition with up to 10 different instruments, combining various styles ranging from Mozart to The Beatles. MuseNet creates music by determining the patterns over a given style as input, generating the respective sequence of notes and chords. It also computes a relational graph between the its various current styles, in order to incorporate as many related musical features as possible.

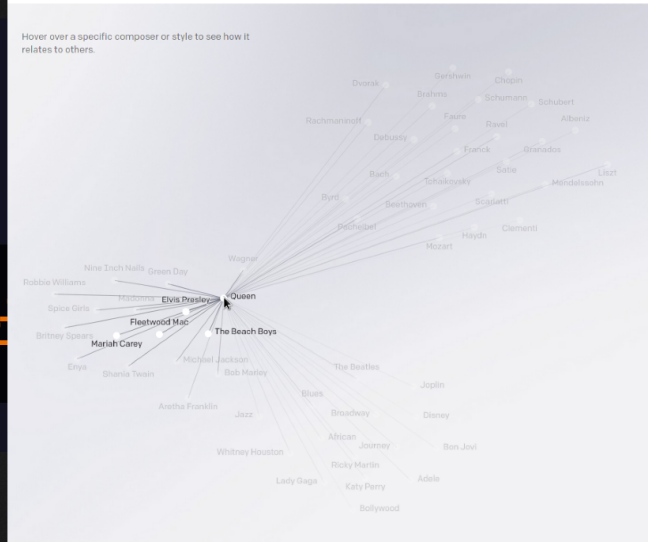


Figure 1.2: An illustration of the MuseNet’s functionalities

1.3 Magenta

Magenta is an open source project (started by a group of engineers from the Google Brain team), based on deep learning and reinforcement learning used for generating and creating art. Based on Google’s Tensorflow as well and Magenta.js, the produce original songs, images, drawings and material textures.

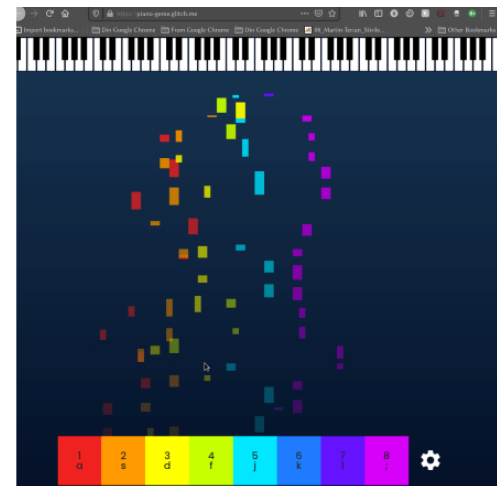
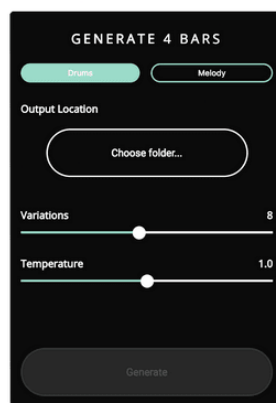
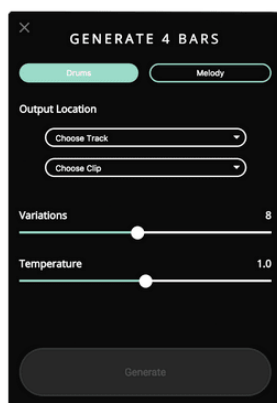


Figure 1.3: Applications based on Magenta

Chapter 2

Technologies used

Python provides many modules for numeric calculus, data processing and manipulation as well as tools for user interaction. In the following chapter we will go through the modules and libraries used for achieving the goal of the application, instrument classification.

2.1 Scipy

SciPy is a Python-based ecosystem of open-source software for mathematics, science, and engineering. The SciPy library contains a collection of numerical algorithms and domain-specific toolboxes, including signal processing, optimization, statistics, and much more.[3] The aforementioned library's submodule *Input and output* provides a variety of ways of reading and writing data in numerous file formats. For the data processing part of the application, we had chosen the *scipy.io.wavfile.read* function because it facilitates the data manipulation process by being compatible with the numpy module, mentioned below (the output of the *scipy.io.wavfile.read* function is a tuple consisting of the sample rate(int) and the audio data(as a *ndarray*).

2.2 NumPy

Numpy is a Python library (part of the SciPy ecosystem) that provides the needed tools in order to perform the mathematical operations behind the Neural Network Framework. The core of the module is the *ndarray* object, which encapsulates an optimized n-dimensional array. Thus, we use *ndarray* objects for computing and storing

the various operations performed throughout the application (e.g. the dot product, the data reshaping). The *ndarray* array is represented as a matrix, whose dimensions are referred as *shape*. Some of the advantages brought by the numpy module are:

- Forward and backward operation compatibility for the various objects contained withing the application
- Fast execution time caused by the optimization of the library [2] (by using pre-compiled C code)
- Support for large number/big data processing.

2.3 TensorFlow

TensorFlow is an end-to-end open source platform for machine learning. It has a comprehensive, flexible ecosystem of tools, libraries and community resources that lets researchers push the state-of-the-art in ML and developers easily build and deploy ML powered applications.[4] From the vast amount of methods and submodules contained within TensorFlow, we used Keras, an utility library built on top of the TensorFlow API. In order for the Neural Network Framework to be able to handle large amounts of data, we inherited in our custom *DataGenerator* class the *tensorflow.keras.utils.Sequence* class. The *Keras DataGenerator* class along with the custom Neural Network Framework *DataGenerator* provide among other functionalities the *getitem* method, which yields a batch of data at a given index, thus overcoming the big data handling problem.

2.4 Kapre

Kapre (Keras Audio Preprocessors) is an audio data preprocessing library built on top of *Keras*, specialized on audio signal handling.[1] Kapre facilitates sound transformation by efficiently performing various operations (e.g. *Short Time Fourier Transformation*, *Magnitude*) in a GPU optimized consistent manner . For the input data modeling we had chosen the *get mel spectrogram layer* function from the module *kapre.composed*. It applies sequentially *STFT*, *Magnitude* and *mel filterbank* transformations over the given input data, returning a mel spectrogram with an user specified sample rate and output shape.

2.5 Librosa

Librosa is a python package for music and audio analysis. It provides the building blocks necessary to create music information retrieval systems. [7] Out of the Librosa API we have chosen the following tools :

- *resample*: A function that resamples an audio file from its original rate to an user given sample rate.
- *to mono*: A function which converts a multi channel signal to mono signaling by averaging the sample values.

2.6 youtube-dl

Youtube-dl is an open source tool for downloading videos and songs from the YouTube platform. In the context of the final application, the youtube-dl module provides the end user with the possibility of classifying a song just by inserting a link to a YouTube video.

2.7 Tkinter

Tkinter is the standard Python GUI Toolkit, a simple yet versatile module was used for the minimalistic Graphic User Interface of the final application.

Chapter 3

Architecture and implementation

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Chapter 4

Use cases

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