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Introduction

In the research on Music Information Retrieval, it is attempted to automatically classify a piece of music based on the raw audio-file, without the manual work of a human listener. To achieve this, computational feature extraction is needed. For this *essentia* is an ideal tool.

Essentia is a C++ library with Python bindings for audio analysis. The software is fairly young (first version released in 2008, newest version 2.1 beta3 released on September 2016) and allows to add new algorithms easily. For this reason, and because *essentia* is open-source, continuing additions are made, making it an extensive library with many up-to-date and experimental algorithms. *Essentia* is mainly supported by the Music Technology Group of the Universitat Pompeu Fabra.

This report tries to demonstrate the usage of *essentia* with both Python and C++.

General Structure of *essentia*

Essentia provides two operating modes: The standard mode and the streaming mode. While the standard mode provides maximum control in C++, it is also recommended for research with Python, due to its interactive environment. The streaming mode allows to easily port your code from Python to C++ and provides easy-to-write extractors in C++ and Python.

The *essentia*-library covers a wide range of algorithms: From spectral, tonal and pitch analysis over loudness, dynamics and rhythm analysis to filters, extraction and segmentation of an audio

file, just to name a few.

These functionalities are provided as individual processing blocks, which are called Algorithms, or *essentia*-algorithms in this report. An *essentia*-algorithm has three different types of attributes: Parameters, Inputs and Outputs. Every Algorithm can have any number of those attributes, including none at all. The Parameters can have default values, and are set when an Algorithm is instantiated. An instantiated Algorithm receives the Input-variables as parameters and returns the Output-variables after finishing its calculation.

Using *essentia* with Python

After importing the *essentia*-library:

```
import essentia
import essentia
import essentia.standard
import essentia.streaming
```

its Algorithms can be accessed. To use an *essentia*-algorithm from the library one first always needs to create an instance of it. Some Algorithms require Parameter-values, whilst other Algorithms have default values for their Parameters, which can be configured when creating the Algorithm. For example the Algorithm *EasyLoader*, which loads a music file, requires the Parameter *filename* to be instantiated with the path to the music file. Optionally, *startTime* and *endTime* can be set to choose the time frame of the extracted audio.

```

        instantiate EasyLoader
loader =
    essentia.standard.EasyLoader(
        filename = "../music/audio.mp3",
        startTime = musicStart,
        endTime = musicEnd)

```

The result of this instantiation is a Python function, which can be used like any other function in Python. Depending on the algorithm, multiple Input parameters need to be passed to the function for the calculation. In the case of *EasyLoader* we have no Input and one Output:

```

        use EasyLoader
audio = loader()

```

On the *essentia* website all Algorithms are documented with a detailed description of their Parameters, Inputs and Outputs as well as the calculation they are performing.

Adding a new Algorithm to the library

In the following, I will describe how to add a new *essentia*-algorithm to the library with the example of *DanceabilityDetailed*. I added this Algorithm because the original Algorithm *Danceability* was not returning the full information that could be retrieved from the calculation.

To add a new Algorithm, first the two aptly named files

- *danceabilityDetailed.h* and
- *danceabilityDetailed.cpp*

need to be created.

The new Algorithm is a subclass of *essentia::Algorithms*, inheriting functions to handle the Input, Output and Parameter attributes. In *danceabilityDetailed.h* first all protected Input and Output variables are declared. In the constructor declaration those variables are set as Inputs and Outputs with *declareInput(...)* and *declareOutput(...)*. With *declareParameters()*

the Parameters are declared. The two functions *compute()* and *configure()* and the variables *name*, *category* and *description* are inherited from *essentia::Algorithms*. Finally, functions and variables specific for the Algorithm are declared.

```

        danceabilityDetailed.h
#ifndef ESSENTIA_DANCEABILITYDETAILED_H
#define ESSENTIA_DANCEABILITYDETAILED_H
#include "algorithm.h"
#include "essentiamath.h"

namespace essentia {
namespace standard {

class DanceabilityDetailed :
    public Algorithm{

protected:
    Input<std::vector<Real>> _signal;
    Output<Real> _danceability;
    Output<std::vector<Real>>
        _dfaExponents;
    Output<std::vector<Real>> _dfaTaus;
    int _preferredSize, _actualSize;

public:
    DanceabilityDetailed() {
        //as seen in the paper
        _preferredSize = 36;
        _actualSize = _preferredSize;
        declareInput(
            _signal,
            "signal", "#d");
        declareOutput(
            _dfaTaus,
            "dfaTaus", "#d");
    }

void declareParameters() {
    declareParameter(
        "minTau", "#d",
        "(0,inf)", 310.);
    declareParameter(
        "maxTau", "#d",
        "(0,inf)", 8800.);
}

```

```

void compute();
void configure();

static const char* name;
static const char* category;
static const char* description;

protected:
std::vector<int> _tau;
Real stddev(const std::vector
    <Real>& array, int start,
    int end) const;
};

} // namespace standard
} // namespace essentia

```

Note, that most of the above code needs to be repeated in a similar fashion for the streaming mode of *essentia*. In the *danceabilityDetailed.cpp*-file the inherited variables and all functions are filled with meaning. Note that *configure* is called when the Algorithm is created. Here the Parameters are used to set corresponding internal variables. *compute()* is called when the created Algorithm is executed, here the actual calculation is defined.

```

danceabilityDetailed.cpp
#include "danceabilityDetailed.h"

using namespace std;
namespace essentia {
namespace standard {

const char* DanceabilityDetailed
    ::name= "DanceabilityDetailed";
const char* DanceabilityDetailed
    ::category = "Rhythm";
const char* DanceabilityDetailed
    ::description
    = DOC("Long\n\nDescription");

void DanceabilityDetailed
    ::configure() {
    Real minTau =
        parameter("minTau").toReal();
    Real maxTau =
        parameter("maxTau").toReal();
    Real tauIncrement =
        parameter(
            "tauMultiplier"
        ).toReal();

    if (minTau > maxTau) {
        throw EssentiaException(
            "Danceability: minTau cannot \
            be larger than maximumTauInMs"
        );
    }

    // tau is the number of blocks of 10ms
    // we calculate each time
    _tau.clear();
    for (Real tau = minTau;
        tau <= maxTau;
        tau += tauIncrement) {
        _tau.push_back(int(tau / 10.0));
    }
}

void DanceabilityDetailed::compute() {
    //using _tau
    ...
}
} // namespace standard
} // namespace essentia

```

Again, most of the above code needs to be repeated similarly for the streaming mode of *essentia*.

For more information on the *danceability*-Algorithm, please refer to Akshay Paranjape's report.

Outlook: One possible application of *essentia*

Considering the wide range of Algorithms *essentia* provides to analyse music, a possible application could be a DJ-program: With *essentia*

certain features of the songs can be extracted, such as beats per minute, meter, and danceability. These features can be used to classify the songs, possibly with machine learning. The DJ-program can use these classifications to determine which song to play next or to create playlists with certain themes. While songs are already classified by e.g. genre or composer, with *essentia* it is possible to automatically classify pieces of music by their raw audio-input and its features.

To gain first-hand experience in the use of *essentia*, I wrote a Python-code (see Appendix) with this application in mind, demonstrating the use of a few Algorithms that should be useful for such feature extractions. The Python code extracts beat positions, beats per minute, meter (currently just experimental), rubato, and novelty.

Conclusion

Essentia offers a huge variety of Algorithms for feature extraction and a great documentation on its website <http://essentia.upf.edu> with detailed descriptions of each algorithm, the use of *essentia* in both Python and C++, and how to create Algorithms.

The analysis of a song is quite fast: Executed on a laptop, my Python code takes only a few seconds per song, naturally depending on the song's length.

Overall, using the *essentia*-library was a worthwhile experience, as it is both easy to use and easy to extend. It has many potential applications in Music Information Retrieval.

References

- [1] Full *essentia* homepage, root-site: <http://essentia.upf.edu>. Last visited: July 3rd, 2017.
- [2] S. Streich, and P. Herrera. *Detrended Fluctuation Analysis of Music Signals: Dance-*

ability Estimation and further Semantic Characterization. Proceedings of the AES 118th Convention, Barcelona, Spain, 2005.

Appendix 1

Python code for feature extraction

```
# -*- coding: utf-8 -*-
"""
Created on Mon May 1 20:19:04 2017

@author: andrea
"""

# import the essentia module. It is aptly named 'essentia'
import numpy as np
import math
import essentia
import essentia.standard
import essentia.streaming
import matplotlib.pyplot as plt
import time

from essentia.standard import*

playMusic = 'yes' # yes or no
musicIndex= -6

myMusicFiles = ['HouseLoop2016.wav', 'SherlockWhoYouReallyAre.wav',
                'PeriodicTableSong.wav', 'LastSled.wav', 'ItsGonnaBeOKAY.wav',
                'SingleLadies.wav', 'Thriller.wav',
                'TchaikovskyFlowersWaltz.wav', 'TchaikovskyFlowersWaltz.wav',
                'SherlockWhoYouReallyAre.wav', 'SherlockWhoYouReallyAre.wav',
                'ChaChaMusic.mp3', 'rumba.mp3', 'Tango.mp3']
myMusicStarts = [0,0, 8, 9, 0, 85, 14, 67, 334, 0, 64, 30, 0, 0]
myMusicEnds = [180,200, 164, 344, 222, 190, 344, 206, 377, 64,
               88, 90, 60, 60]
musicfile = '../audio/' + myMusicFiles[musicIndex]
musicStart = myMusicStarts[musicIndex]
musicEnd = myMusicEnds[musicIndex]

samplingrate = 44100

# let's define a small utility function
def play(audiofile):
    import os, sys

    # NB: this only works with linux!! mplayer rocks!
```

```

    if sys.platform == 'linux2':
        if playMusic == 'yes':
            cmd = 'mplayer_' + audiofile + '_-ss_'
            + str(myMusicStarts[musicIndex]) + '_-endpos_'
            + str(myMusicEnds[musicIndex]-myMusicStarts[musicIndex])
            os.system(cmd)
        else:
            print 'Not playing audio because you asked me not to do so'
    else:
        print 'Not playing audio ...'

print 'analyzing the audio_' + musicfile

#play(musicfile)
loader = essentia.standard.EasyLoader(
    filename = musicfile, startTime = musicStart,
    endTime = musicEnd)

# and then we actually perform the loading:
audio = loader()
audioSize = audio.size

analysisStart = 0*samplingrate
analysisEnd = audioSize
analysisDelta = 1024#2205
numFreqIter = int(math.floor((analysisEnd-analysisStart)/analysisDelta))

#instantiate all algorithms that don't need yet unknown parameters:
getWindow = Windowing(type = 'hann')
# FFT() would give the complex FFT,
# here we just want the magnitude spectrum
getSpectrum = Spectrum()
getFrequencyBands = FrequencyBands()
getNoveltyCurve = NoveltyCurve()
getBpmHistogram = BpmHistogram()
getBeatTrackerDegara = BeatTrackerDegara()
getBeatogram = Beatogram()
getRythmExtractor = RhythmExtractor2013()
getMeter = Meter()
getBpmRubato = BpmRubato()

#####
## get rythm, bpm: ##
#####

'''

```

```

bpm (real) – the tempo estimation [bpm]
ticks (vector_real) – the estimated tick locations [s]
confidence (real) – confidence with which the ticks are
detected (ignore this value if using 'degara' method)
estimates (vector_real) – the list of bpm estimates
characterizing the bpm distribution for the signal [bpm]
bpmIntervals (vector_real) – list of beats interval [s]
',',

(bpm, ticks, confidence, estimates, bpmIntervals) = getRythmExtractor(audio)
print 'Beats_per_Minute:'
print bpm

#####
## get meter !!!Meter is only experimental ##
#####
ticksDegara = getBeatTrackerDegara(audio)
getBeatsLoudness = BeatsLoudness(beats=ticksDegara)
(loudness, loudnessBandRatio) = getBeatsLoudness(audio)
beatogram = getBeatogram(loudness, loudnessBandRatio)
meter = getMeter(beatogram)

print 'meter:_'
print meter

#####
# get rubato for bpm: ##
#####

',',

rubatoStart (vector_real) – list of timestamps where
the start of a rubato region was detected [s]
rubatoStop (vector_real) – list of timestamps where
the end of a rubato region was detected [s]
rubatoNumber (integer) – number of detected rubato regions
',',

(rubatoStart, rubatoStop, rubatoNumber) = getBpmRubato(ticksDegara)
print 'rubato:_'
print rubatoStart
#####
# get novelty ##
#####

i=0
frame = audio[analysisStart + i*analysisDelta : analysisStart
+ (i+1)*analysisDelta]

```

```

spectrum = getSpectrum(getWindow(frame))
allFrequencyBands = getFrequencyBands(spectrum)

for i in range(1,numFreqIter-1):
    frame = audio[analysisStart + i*analysisDelta : analysisStart
                  + (i+1)*analysisDelta]
    spectrum = getSpectrum(getWindow(frame))
    allFrequencyBands = np.append(allFrequencyBands ,
                                  getFrequencyBands(spectrum))

allFrequencyBands = allFrequencyBands.reshape(numFreqIter-1, -1)
novelty = getNoveltyCurve(allFrequencyBands)

#####
## Do something usefull with the data ##
#####

from matplotlib.pyplot import plot , draw , show
#ion() # enables interactive mode

y1 = [0 , novelty.size]
x1 = [0 , 0]

fig = plt.figure(figsize=(15,5))
ax = fig.add_subplot(111)
ax.set_xlim(0 , novelty.size+1)
ax.set_ylim(audio.min()*novelty.max()/audio.max() , novelty.max())
plt.show(block=False)
#plt.show()

ax.plot(np.arange(0 , audio.size)*novelty.size/audio.size ,
        audio*novelty.max()/audio.max() , label=myMusicFiles[musicIndex] ,
        color='dodgerblue')

for i in range(0,rubatoNumber):
    xc = (rubatoStart[i]*novelty.size)/(musicEnd-musicStart)
    ax.axvline(x=xc , color="green")
    xc = (rubatoStop[i]*novelty.size)/(musicEnd-musicStart)
    ax.axvline(x=xc , color="green")
ax.plot(novelty , label="Novelty" , color="red" , linewidth=2.)

ax.set_ylim(audio.min()*novelty.max()/audio.max() , novelty.max())

show()

```



```

# get the canvas object
canvas = ax.figure.canvas
background = canvas.copy_from_bbox(ax.bbox)
plt.subplots_adjust(left=0.1, right=0.9, top=0.75, bottom=0.1)
leg = ax.legend(bbox_to_anchor=(0, 1.3), loc=2, borderaxespad=0.)

# add the progress line.
line = ax.axvline(x=0, color='r', animated=True)

starttime=time.time()
mytimer=0
mytimer_ref=0

def update(canvas, line, ax):
    # revert the canvas to the state before any progress line was drawn
    ax.lines = ax.lines[:2+2*rubatoNumber]
    t = time.time() - starttime #- (musicEnd - musicStart)
    mytimer = ((t + mytimer_ref)*novelty.size)/(musicEnd-musicStart)
    # update the progress line with its new position
    ax.axvline(x=mytimer, color='r')
    canvas.blit(ax.bbox)

def startGraph():
    global starttime
    starttime=time.time()
    global mytimer_ref
    mytimer_ref=0#event.xdata
    print "starttime",starttime, mytimer_ref
    timer.start()

def onclick(event):
    play(musicfile)
    startGraph()

timer=fig.canvas.new_timer(interval=5)
args=[canvas, line, ax]
timer.add_callback(update,*args) # every 100ms it calls update function
# when I click the mouse over a point, line goes to that point and start moving
cid1=line.figure.canvas.mpl_connect('button_press_event',onclick)

#fig.savefig("analysis.png")
plt.show()
print 'plot_finished'

```