fma-selection

This project aims to use techniques in optimization for feature selection. The goal of this project is to expand on the research of Defferrard et. al, and apply the techniques for music analysis.

FMA: A Dataset For Music Analysis

Michaël Defferrard, Kirell Benzi, Pierre Vandergheynst, Xavier Bresson, EPFL LTS2.

The dataset is a dump of the <u>Free Music Archive (FMA)</u>, an interactive library of high-quality, legal audio downloads. The referenced <u>paper</u> is a **pre-publication release**.

Data

All metadata and features for all tracks are distributed in fma_metadata.zip (342 MiB). The below tables can be used with pandas or any other data analysis tool. See the papet or the [usage] notebook for a description. * tracks.csv: per track metadata such as ID, title, artist, genres, tags and play counts, for all 106,574 tracks. * genres.csv: all 163 genre IDs with their name and parent (used to infer the genre hierarchy and top-level genres). * features.csv: common features extracted with librosa. * echonest.csv: audio features provided by Echonest (now Spotify) for a subset of 13,129 tracks.

Then, you got various sizes of MP3-encoded audio data:

- 1. fma_small.zip: 8,000 tracks of 30s, 8 balanced genres (GTZAN-like) (7.2 GiB)
- 2. fma_medium.zip: 25,000 tracks of 30s, 16 unbalanced genres (22 GiB)

Prerequisites

1. Download some data, verify its integrity, and uncompress the archives. $\mathtt{curl} \ \, \text{-0}$

```
https://os.unil.cloud.switch.ch/fma/fma_metadata.zip

curl -O https://os.unil.cloud.switch.ch/fma/fma_small.zip

curl -O https://os.unil.cloud.switch.ch/fma/fma_medium.zip

echo "f0df49ffe5f2a6008d7dc83c6915b31835dfe733 fma_metadata.zip" | sha1sum -c -

echo "ade154f733639d52e35e32f5593efe5be76c6d70 fma_small.zip" | sha1sum -c -

echo "c67b69ea232021025fca9231fc1c7c1a063ab50b fma_medium.zip" | sha1sum -c -

unzip fma_metadata.zip

unzip fma_small.zip

unzip fma_medium.zip
```

- 2. Clone the repository. git clone https://github.com/hfloresr/fma-selection.git cd fma-selection
- 3. Install the Python dependencies from <code>environment.yml</code>. Depending on your usage, you may need to install <code>ffmpeg</code> or <code>graphviz</code>. Install <code>CUDA</code> if you want to train neural networks on GPUs (see <code>Tensorflow's instructions</code>). <code>conda env create --file environment.yml</code>
- 4. Fill in the configuration. ```sh cat .env AUDIO_DIR=/path/to/audio