**Master Thesis Proposal**

Multi-task Learning of Rhythmic and Tonal Properties in Music

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1 Motivation & Problem Statement

Artificial Intelligence is rapidly becoming the dominant approach for solving a variety of multimedia related tasks, be it image classification, text generation, speech recognition or synthesis. The domain of music is no exception with traditional handcrafted signal processing methods slowly being phased out in favour of those heavily based on machine learning, in particular, methods involving deep learning have garnered considerable attention in recent years [1].

Deep learning solutions for chord recognition, utilizing convolutional neural networks [3], and for audio beat tracking, using recurrent neural networks [2] and dilated convolutional neural networks [1] have achieved state of the art results, surpassing their predecessors by a significant margin. Additionally many music analysis problems overlap in one way or another and it was shown by Sebastian Böck, Matthew Davies and Peter Knees that for example creating a joint model for the task of beat tracking and tempo estimation can have benefits for each individual task [4]. By training the model on both beat and tempo information, evaluating it and then providing additional tempo only information for further training, the beat tracking capabilities of the model were shown to have also further improved. This has two important implications: first of all jointly learning musical properties can have a positive effect on the outcome of each individual task but also a purely practical benefit is that while tempo information is relatively cheap to produce and is more widely available, high quality beat information is much harder to obtain and by being able to train the model on large amounts of the former it is possible to somewhat mitigate the scarcity of the latter.

In a similar fashion learning a joint model for beat tracking and chord recognition should yield improved results for both. Moreover since high quality annotations for both these tasks are expensive to produce, a joint model could greatly benefit from having the ability to work with either kind of annotations.

2 Expected Results

The aim of this work is to investigate the potential of a deep learning architecture for the joint task of beat tracking and chord recognition.

The proposed model will be first trained and evaluated on beat and chord information separately to demonstrate its feasibility, then a similar experiment will be conducted with both beat and chord information present in order to investigate whether there is any benefit from multi-task learning in this case and finally the model will be further trained on beat information only, to evaluate if its chord recognition capability has improved, and the other way around, it will be further trained on chord information only, to evaluate if its beat tracking capability has improved.

3 Methodological Approach

* Literature review:

Will be roughly divided into three parts.

Information must be gathered and reviewed about current deep learning architectures for audio beat tracking, the same will be done for chord recognition and finally information on possible approaches for multi-task learning have to be gathered and analyzed.

* Model:

Three deep learning models will be proposed.

Two models approximating the state of the art solutions for beat tracking and chord recognition and a third model for solving the joint problem.

* Implementation:

The models described above will be implemented as standalone solutions for their respective tasks.

* Evaluation:

The models will be evaluated using k-fold cross validation and analysis of the results will be performed to determine whether learning a joint model presents an advantage over learning separate models for each task and whether providing one type of information additionally as training data yields improvement for the unrelated task.

4 State of the Art

5 Relevance to the Curriculum of Logic and Computation (066 931)

This work is closely tied to the module Knowledge Representation and Artificial Intelligence, most notably “184.702 Machine Learning”, “188.501 Similarity Modeling 1” and “188.498 Similarity Modeling 2” as they deal with supervised machine learning which includes deep learning, but also “188.502 Media and the Brain 1”, “188.499 Media and the Brain 2”, “194.039 Intelligent Audio and Music Analysis” (Probably shouldn’t be included? Not part of Logic and computation) and “188.413 Self-Organizing Systems”

References

[1] Temporal convolutional networks for musical audio beat tracking (boeck 1)

[2] Boeck 2 RNNs and stuff

[3] Chord recognition Filip Korzeniowksi and some1 else

[4] Boeck, Knees ad one more guy (Beats + Tempo)