

# **Recognizing Printed Melodies using Object Detection and a Convolution Neural Network**

## **Report on the Lecture Machine Learning using MATLAB**

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## **1 Introduction**

The goal of Optical Music Recognition (OMR) is to reproduce a sequence of document-written musical notation by a machine. Since the language of musical notation contains hundreds of symbols, this problem becomes really complex. Our aim is to treat only a fragment of the musical notation language, namely the fragment of American and Europe folk music. It has a huge restriction on the alphabet, since it does not contain chords, for example, and is more appropriate of our use case. Because we want to apply OMR to an application that photocopies a tune of a song-book and transforms it later to an audio file. This can be further used to learn how to sing this tune.

Available OMR Datasets such as DeepScores (Tugener, 2021) contains around 30 000 000 sheets of written music with close to a hundred millions of small objects. This would really blow up our project. That's why we generate our own Dataset.

We further follow the general framework to OMR that contains according to Rabela et al. (2012) the following steps.

1. image pre-processing
2. recognition of the musical symbols;
3. reconstruction of the musical information in order to build a logical description of musical notation; and
4. construction of a musical notation model to be represented as a symbolic description of the musical sheet.

OMR really has difficulties on point 2 (Pacha et al., 2018). The authors applied Faster R-CNN, RetinaNet and U-Net on various datasets. The performance on DeepScores, which is not hand-written and therefore easier to be trained, is not promising as the mean average precisions of the intersection over union ratios are 19.6%, 9.8%, and 24.8%, respectively.

An other approach by Metaj and Magnolfi (2021) is more promising, but has really good results around 95% only for a selection where the bounding boxes have at least an intersection over union ratio of 0.50.

That's why we invent a new approach on the object detection that is not based on a neural net.

## **2 Musical Notation Background**

foo

## **3 Training Data Sampling**

foo

## **4 Object Detection**

foo

## **5 Image Classification**

foo

## **6 Post Processing**

foo

## Bibliography

- Metaj, S. and Magnolfi, F. (2021). MNR\_MUSCIMA\_Notes\_Recognition. [https://github.com/StivenMetaj/MNR\\_MUSCIMA\\_Notes\\_Recognition](https://github.com/StivenMetaj/MNR_MUSCIMA_Notes_Recognition).
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