



Session 3: Optical Music Recognition

Chairs:

Nina Hirata (University of São Paulo) Josep Lladós (CVC, Universitat Autònoma de Barcelona)



Session outline (each paper: 10 min presentation)

- On the Potential of Fully Convolutional Neural Networks for Musical Symbol Detection. Matthias Dorfer, Jan Hajič, Gerhard Widmer
- Towards a Universal Music Symbol Classifier. Alexander Pacha, Horst Eidenberger
- Bootstrapping Samples of Accidentals in Dense Piano Scores for CNN-Based Detection. Kwon-Young Choi, Bertrand Coüasnon, Richard Zanibbi, Yann Ricquebourg
- Optical Music Recognition by Recurrent Neural Networks. Arnau Baró,
 Pau Riba, Jorge Calvo Zaragoza, Alicia Fornés
- Pen-based Music Document Transcription. Javier Sober-Mira, Jorge Calvo Zaragoza, David Rizo, Jose Manuel Inesta.





Introduction to the topic

- Optical music recognition (OMR) can be seen as the application of optical character recognition to interpret sheet music or printed scores into editable or playable form.
- Traditionally, given a musical score image, it follows the traditional pipeline of layout analysis (staff removal, and symbol segmentation), and classification.
- The use of syntactic models is highly important.
- One of the major challenges is the recognition of handwritten scores, both off line (e.g. historical musical documents) or on line (creation and edition).





Paper 1:

On the Potential of Fully Convolutional Neural Networks for Musical Symbol Detection. Matthias Dorfer, Jan Hajič, Gerhard Widmer

- Notehead detection on handwritten scores
- CNN that produces a pixel level probability map, followed by searching of local peeks in the map (considered centroids of the noteheads)
- high F-score both for writer dependent and independent tests
- adaptable to the detection of other types of symbols





Paper 2:

Towards a Universal Music Symbol Classifier. Alexander Pacha, Horst Eidenberger

- building of a large musical symbol dataset, by unifying existing datasets
- unification of class names and treatment of ambiguous symbols
- preliminary classification tests give error rates below 2%





Paper 3:

Bootstrapping Samples of Accidentals in Dense Piano Scores for CNN-Based Detection. Kwon-Young Choi, Bertrand Coüasnon, Richard Zanibbi, Yann Ricquebourg

- A CNN based detector (visual attention) to localize and classify three accidental symbols associated with a note head, or the note head if there is no accidental.
- To cope with "few data samples", a data augmentation bootstrapping method is used.
- Complex and damaged piano scores.





Paper 4:

Optical Music Recognition by Recurrent Neural Networks. Arnau Baró, Pau Riba, Jorge Calvo Zaragoza, Alicia Fornés

- Recognition of musical scores as a sequence using BLSTM Recurrent Neural Networks.
- Context can be introduced.
- Use of a training synthetic dataset of more than 40.000 images labeled at primitive level.





Paper 5:

Pen-based Music Document Transcription. Javier Sober-Mira, Jorge Calvo Zaragoza, David Rizo, Jose Manuel Inesta.

- Human-machine interaction task for music notation creation.
- Combination of two modalities for the recognition: on-line and off-line data.
- Convolutional Neural Network approach.





Discussion

Open questions:

- Overall summary: The papers address problems from symbol detection and classification, to transcription of music scores.
- Deep learning models are the current choice (which model works best?)
- Many works are proposing the introduction of deep neural network approaches, but still for a particular step in the pipeline. Can we think in end-to-end systems?
- What about the "personal" component of music interpretation? Each composer has his/her own style. Can be use this type of contextual knowledge? How?
- Shall we approach other communities (e.g. NLP)?
- Availability and imbalance of training data is perceived as an important concern. What about databases? What is required? How to obtain it?
- Which is the "big challenge"?