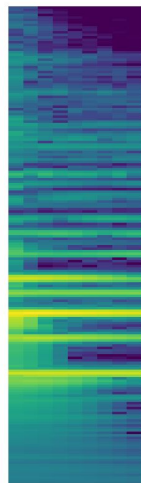


A Data-Driven Methodology for Considering Feasibility and Pairwise Likelihood in Deep Learning Based Guitar Tablature Transcription Systems

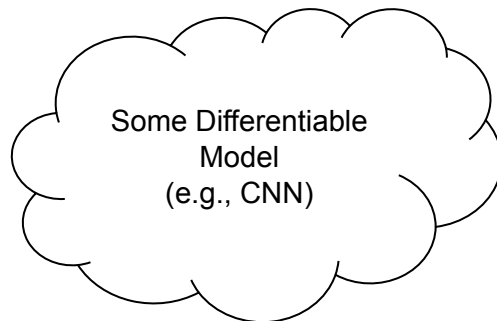
Frank Cwitkowitz, Jonathan Driedger, and Zhiyao Duan

Guitar Tablature Transcription

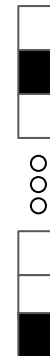
Some Spectral Features
(e.g., CQT)



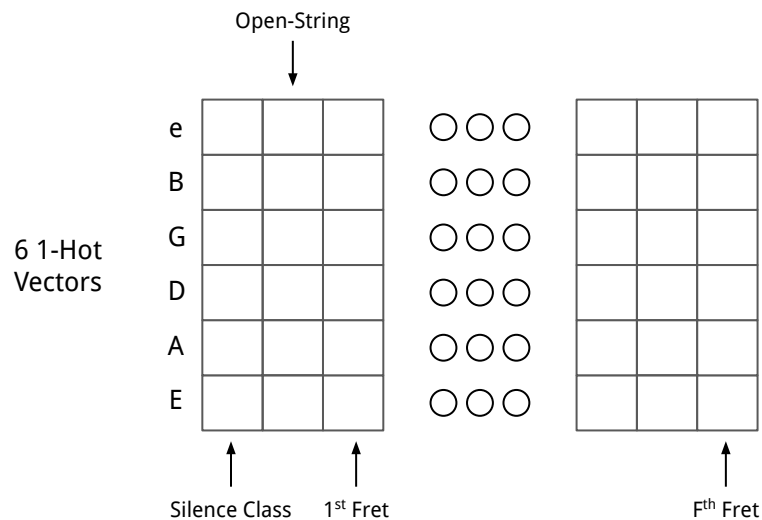
| ← context → |



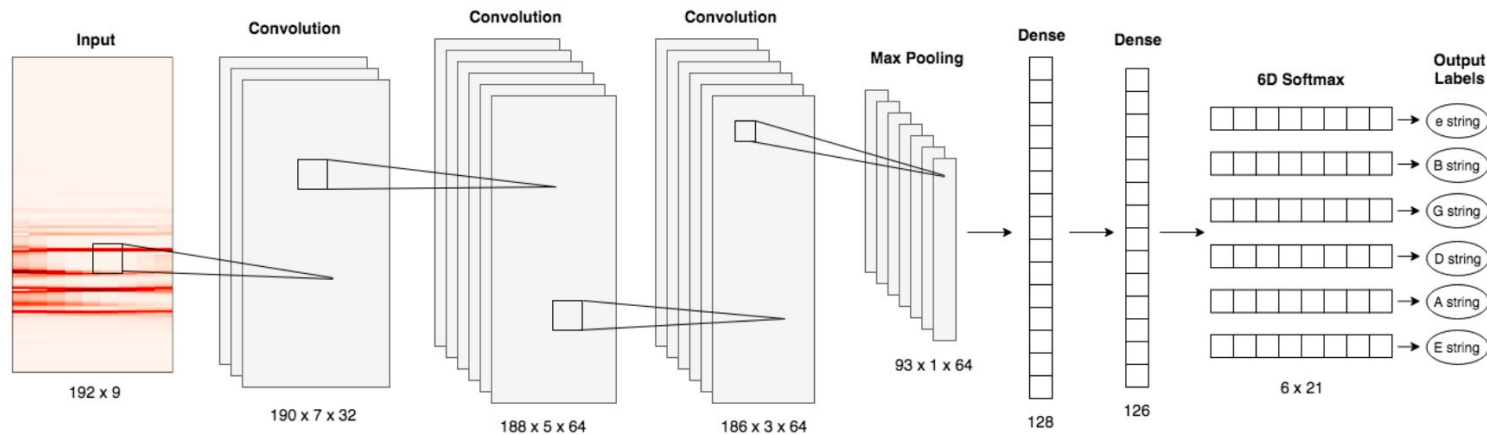
6-Hot Vector



Guitar Tablature Transcription



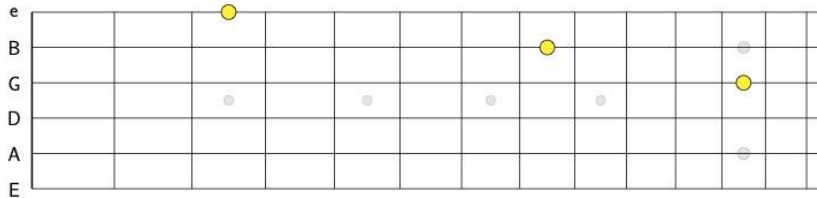
Example - TabCNN¹



¹Andrew Wiggins and Youngmoo Kim. "Guitar Tablature Estimation with a Convolutional Neural Network". In: Proceedings of ISMIR. 2019.

6D Softmax Approach

- Always generates valid predictions
- 6 independent classification tasks
- Ignores high correlation of fretting between strings
 - Physical limitations
 - Musical tendencies



The main obstacle for guitar
tablature transcription is
labeled data!

Symbolic Tablature

- Prescriptive playing notation for guitar
- Large digital collections available (e.g. DadaGP²)

A musical score for guitar. The top staff is a treble clef with a 4/4 time signature. It contains a melody with various note values and rests. The bottom staff is a guitar tablature with six lines. It includes fret numbers (0, 2, 5, 7, 12) and chord diagrams (X-X-X). The score is divided into four measures, numbered 1, 2, 3, and 4. Measure 1 starts with a treble clef and a 4/4 time signature. Measure 2 has a 7 (slide) and a 12 (bend). Measure 3 has a 3 (triple) and a 5 (fret). Measure 4 has a 3 (triple) and a 5 (fret).

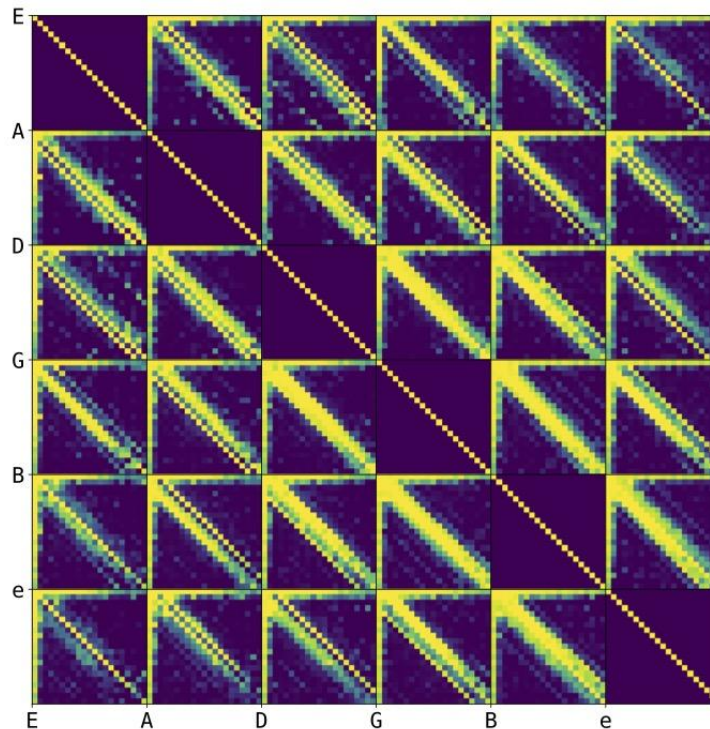
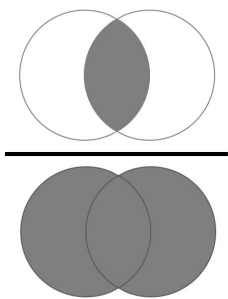
²Pedro Sarmento et al. "DadaGP: A Dataset of Tokenized GuitarPro Songs for Sequence Models". In: Proceedings of ISMIR. 2021

Proposed Methodology

- Leverage tablature data
- Re-formulate output layer (sigmoid)
- Incorporate a novel inhibition loss
 - Discourage co-activation of unlikely note pairs
 - Informed by pairwise likelihoods estimated from data

Estimating Pairwise Likelihood

- Compute frame-level IOU of all string/fret (S/F) pairs



Inhibition Loss

$$w(c_i, c_j) = (1 - \boxed{IoU(i, j)})^b$$

Estimated likelihood for S/F i and S/F j .

Inhibition Loss

$$w(c_i, c_j) = (1 - IoU(i, j))^{\boxed{b}}$$

Boosting parameter to relax inhibition.

Inhibition Loss

$$w(c_i, c_j) = (1 - IoU(i, j))^b$$

Inhibition weight for S/F i and S/F j .

Inhibition Loss

$$w(c_i, c_j) = (1 - IoU(i, j))^b$$

$$L_{inh} = \frac{1}{2N} \sum_{n=1}^{\boxed{N}} \sum_{i=1}^C \sum_{j=1}^C z_{c_i, n} z_{c_j, n} w(c_i, c_j)$$

Total number of frames.

Inhibition Loss

$$w(c_i, c_j) = (1 - IoU(i, j))^b$$

$$L_{inh} = \frac{1}{2N} \sum_{n=1}^N \sum_{i=1}^{\boxed{C}} \sum_{j=1}^{\boxed{C}} z_{c_i, n} z_{c_j, n} w(c_i, c_j)$$

Total number of distinct S/F combinations.

Inhibition Loss

$$w(c_i, c_j) = (1 - IoU(i, j))^b$$

$$L_{inh} = \frac{1}{2N} \sum_{n=1}^N \sum_{i=1}^C \sum_{j=1}^C z_{c_i, n} z_{c_j, n} w(c_i, c_j)$$

Activation produced by model for S/F i and S/F j during frame n .

Inhibition Loss

$$w(c_i, c_j) = (1 - IoU(i, j))^b$$

$$L_{inh} = \frac{1}{2N} \sum_{n=1}^N \sum_{i=1}^C \sum_{j=1}^C z_{c_i, n} z_{c_j, n} w(c_i, c_j)$$

Inhibition weight for S/F i and S/F j .

Experiments

➤ Employ TabCNN as a baseline (extending the model)

- 6-fold cross-validation on GuitarSet³
- Evaluate with same transcription metrics
- Introduce “distribution” metrics
- Ablation study w.r.t. the inhibition loss

³Qingyang Xi et al. “GuitarSet: A Dataset for Guitar Transcription”. In: Proceedings of ISMIR, 2018.

Discussion

- Tablature performance does not increase significantly
 - Small size of GuitarSet
 - Some label noise in GuitarSet
- Distribution of predictions more closely matches DadaGP

Summary

- Pairwise likelihoods of S/Fs estimated using DadaGP
- Propose novel inhibition loss incorporating likelihoods
- TabCNN (baseline) augmented with the inhibition loss

All code is available at

<https://github.com/cwitkowitz/guitar-transcription-with-inhibition>