

Objectives

The goal of BachBot is to generate 4-part Baroque chorales in the stype of Johann Sebastian Bach. We interpret this as sampling random chorales from a generative probabilistic model of Bach chorales and identify four discrete objectives:

- Melody modeling: Marginal distribution over univariate Soprano melody sequences
- Melody harmonization: Conditional distribution over multivariate (Alto, Tenor, and Bass) harmony parts given fixed Sporano melody
- One-pass polyphonic generation: Sequentially modeling of all parts jointly
- Applications in music analysis: What would Bach do? Enhance "Bach-ness" of inputs.

Data Representation

We will explore how three different representations of music (the first two of which are isomorphic) affect convergence and performance:

- **Per-part tuples** four collections (one for each voice) of note/rest and duration tuples; note that the time interval between two tuples is variable
- **Per-part roll** categorical array $X_{p,t} \in V$ denoting note played by part $p \in \{S, A, T, B\}$ at time $t \in \{1, 2, \dots, T\}$; to distinguish notes held from previous times from notes articulated at the current time, $Y_{p,t} \in \{0,1\}$ indicates if the note is articulated at time t.
- **Piano roll** similar to per-part roll except $X_{n,t} \in \{0,1\}$ denotes if **note** n (rather than part p) is played at time t; this represents cannot express two different parts playing the same note

BachBot: Deep generative modeling of Bach chorales

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Melody Modeling

- Given initial seed, generate melody sequence
- Baseline: N-gram language model perplexity
- Experiments:
- RNN, LSTM, and GRU architectures
- Word level vs note level features, augmentation with expert crafted features
- Constant (roll representation) or varying timestep (tuple representation) per input



Figure 1: Inside of an LSTM cell[1]

One-pass polyphonic generation

- Given initial seed, generate entire chorale
- Baseline: n/a, subjective evaluation
- Experiments:
- Bi-axial and grid architectures
- Convolutional vs recurrent dimensions



Figure 3: Biaxial RNN architecture[3]

Melody Harmonization

- Given melody, generate the harmony parts
- Baseline: HMM-based system[2] accuracy • Experiments:
- Single multivariate vs 4 independent LSTMs with MRF refinement
- Bi-directional LSTM



Figure 2: Bidirectional RNN hidden states

Applications in Music Analysis

• MAP : what would Bach do? Interpreting hidden state activations • Enhance "Bach-ness" of input



Figure 4: 2D grid RNN architecture[4]

- transpose C major/A minor) complete
- torch 2-layer LSTM melody model
- keras/tensorflow bi-axial LSTM modely model Upcoming
- Get baselines for N-gram melody model and HMM-based harmonization [2]
- Augment feature representation with expert crafted features
- Investigate GRUs and vanilla RNNs for melody modeling and harmonization
- Implement and compare biaxial vs grid RNNs for
- Sample outputs and perform subjective evaluation using MTurk



https://github.com/feynmanliang/bachbot • fl350@cam.ac.uk



Project status

Completed

• Preprocessing pipeline (strip markup, extract 4 parts,

harmonization and single pass generation

References

[1] 2013 IEEE Workshop on Automatic Speech Recognition and Understanding, Olomouc, Czech *Republic, December 8-12, 2013.* IEEE, 2013.

[2] Moray Allan and Christopher KI Williams. Harmonising chorales by probabilistic inference. Advances in neural information processing systems, 17:25-32, 2005.

[3] Christopher Olah. Understanding lstm networks, 2015.

[4] Nal Kalchbrenner, Ivo Danihelka, and Alex Graves. Grid long short-term memory.

CoRR, abs/1507.01526, 2015.

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Contact Information

