

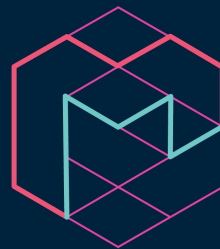
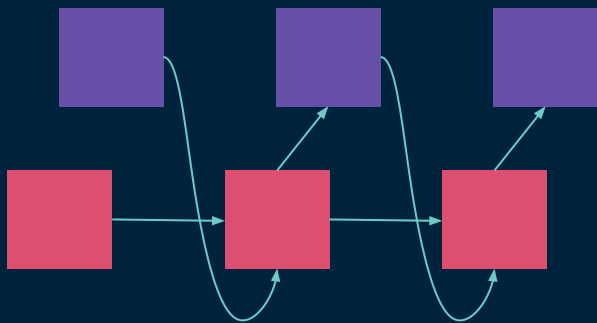
# Generating Event Sequences with RNNs

Kyle Kastner (kkastner@)

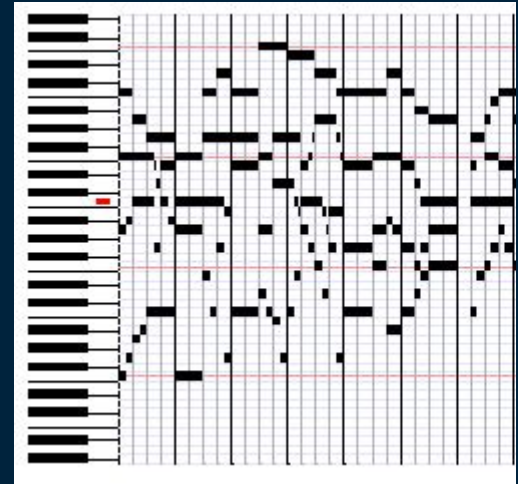
Intern from Université de Montréal - MILA

with Mike Schuster (schuster@)

and Douglas Eck (deck@)

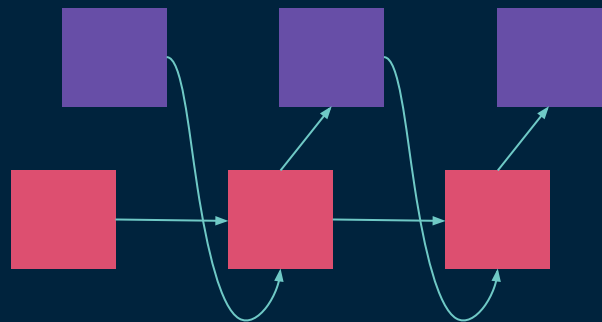


# Piano Roll



For example purposes only, music is NOT the same piece in each image

# Basic Approach: RNN Language Model



## Classic RNNLM

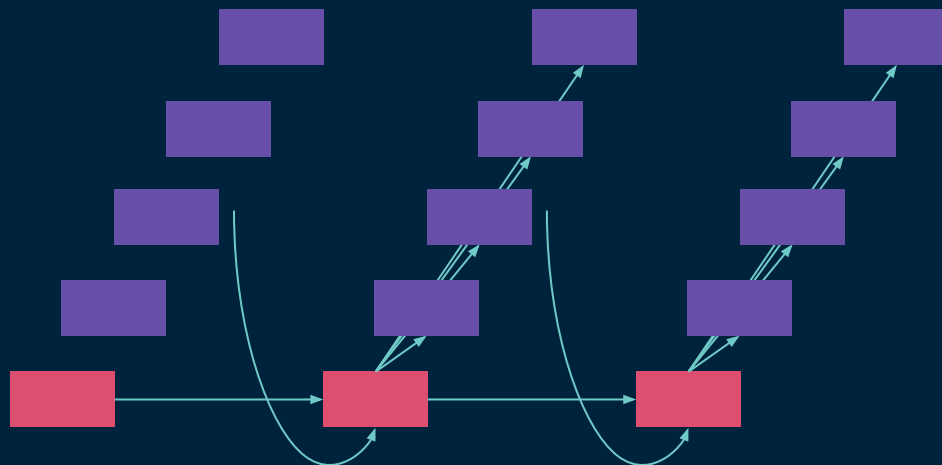
- <START>The cat eats food
- The cat eats food<EOS>

Finding chords is a learning problem itself

- Use notes directly

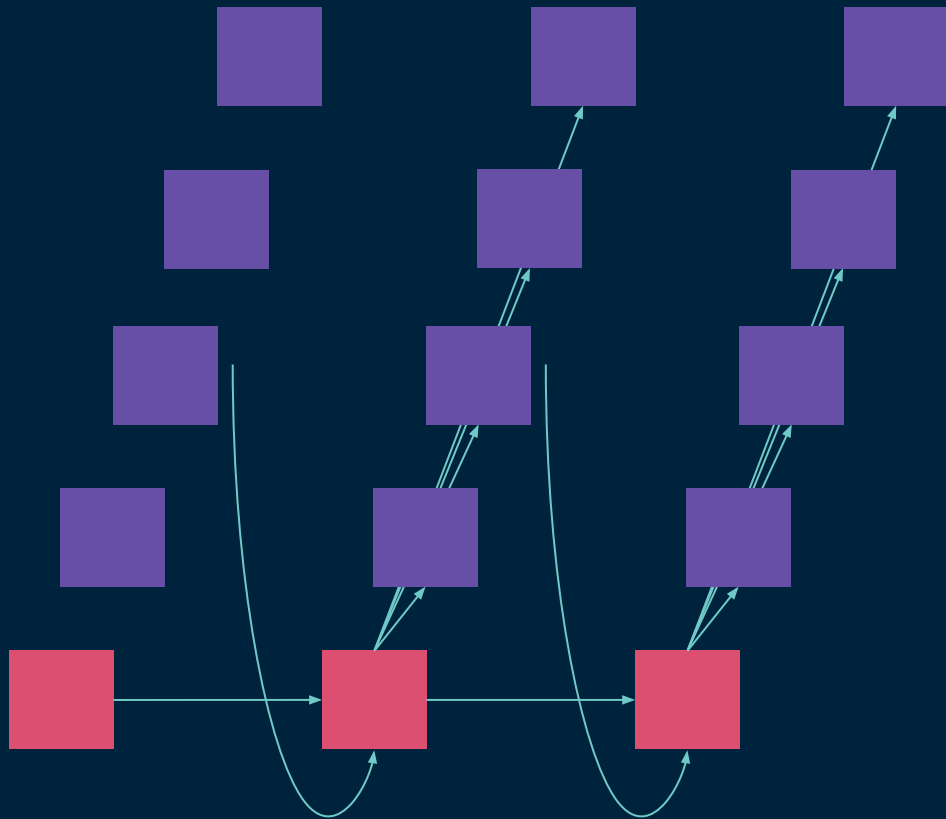
A big problem... how to do multiple outputs per step?

- Add more (softmax) layers



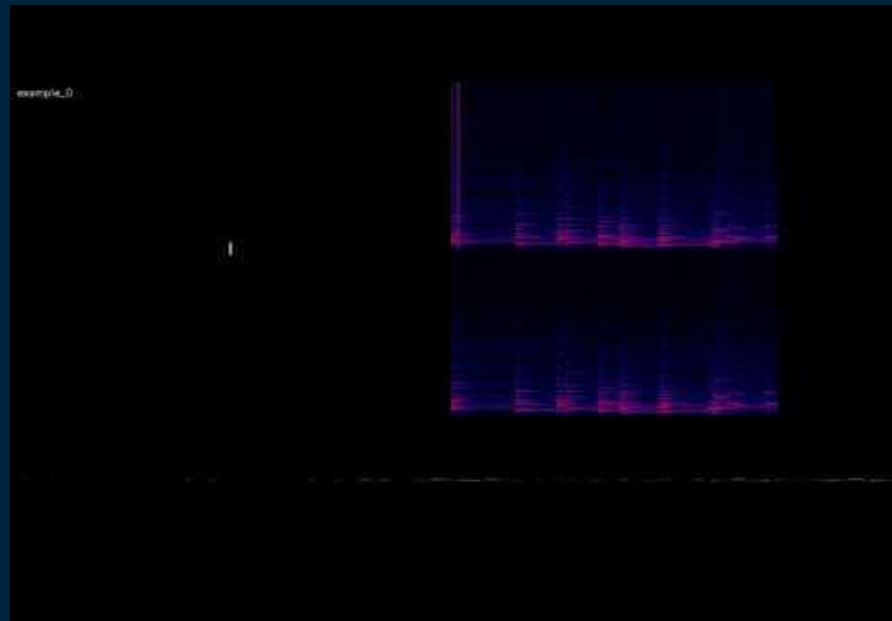
# Goal: Generative modeling of polyphonic music

## Approach (Week 1-4): GRUs on piano roll



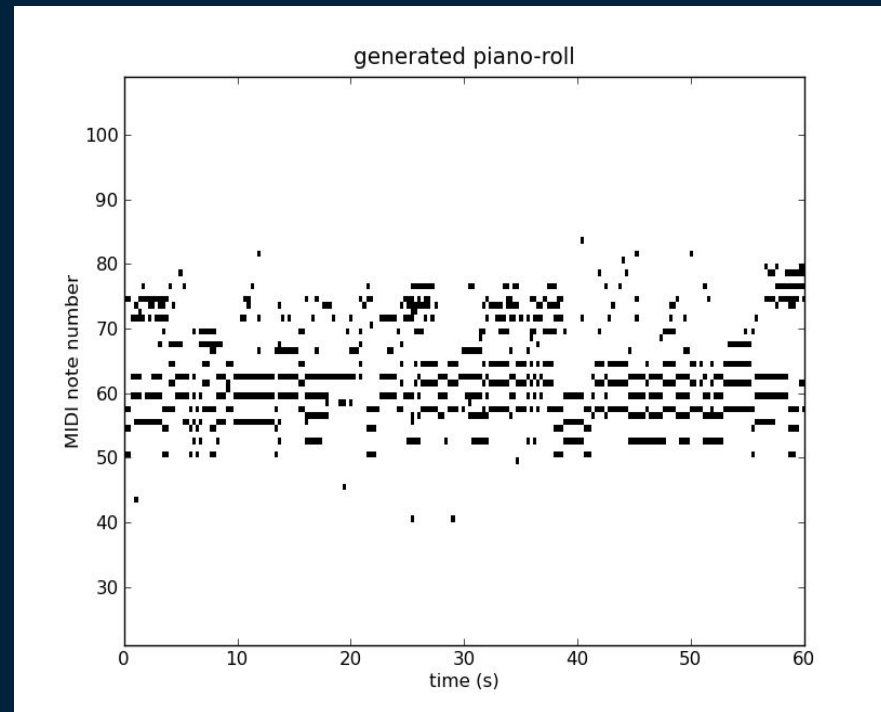
NOTES  
(1 of 89)

- Bach Chorales, ~80k steps, quarter note samp.
- Training time, ~30 minutes



# Major Problems

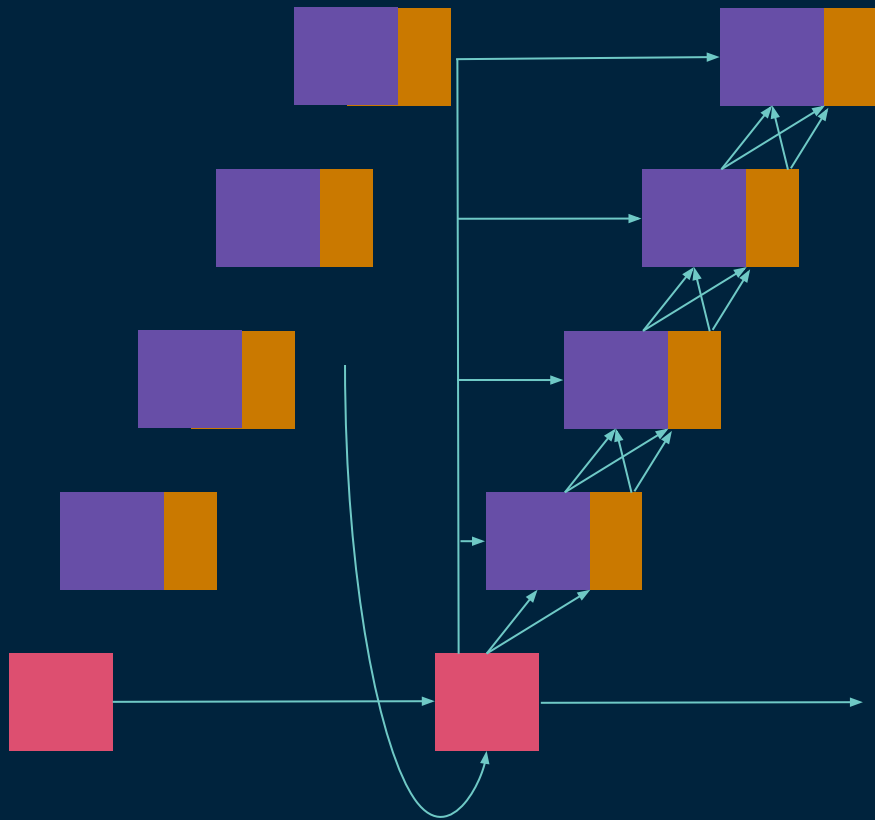
- **No output context at current time**
  - Add chain dependency on output
  - MADE style masked teacher forcing
  - Key component of Pixel RNN
- **“Piano roll” representation**
  - Continuity is *hard* in this format
  - Decisions should be all at once
  - Basic MIDI is an event stream



N Boulanger-Lewandowski, Y. Bengio and P. Vincent, [Modeling Temporal Dependencies in High-Dimensional Sequences: Application to Polyphonic Music Generation and Transcription](#), in Proceedings of the 29th International Conference on Machine Learning (ICML), 2012.

# Goal: Generative modeling of polyphonic music

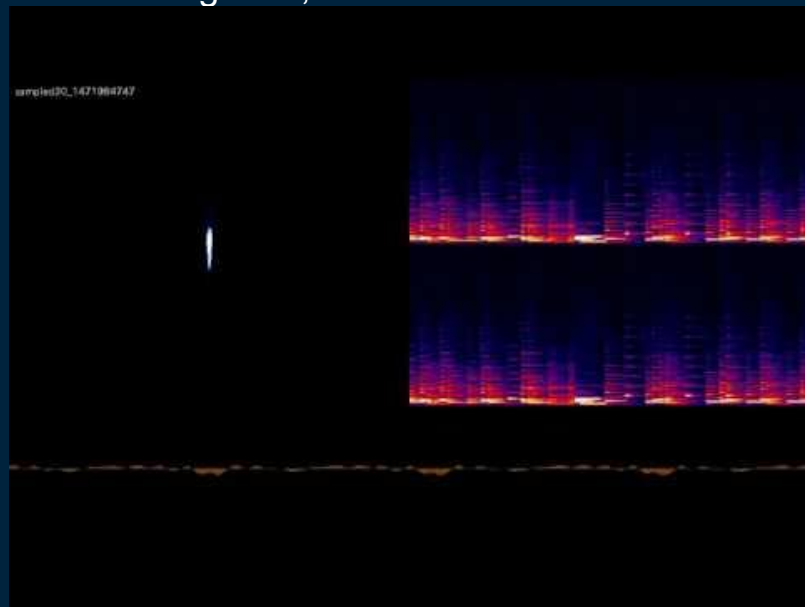
## Approach (Week 5-Now): GRUs on MIDI event stream



NOTES  
(1 of 89)

DURATIONS  
(1 of 12)

- 1 step  $\approx$  1 second of time on average
- Bach Chorales, 41,198 steps
- Training time, 2m14.780s



# Takeaway Lessons / Opinions

- Softmax temperature is a crucial hyperparameter in sampling!
- Condition on ***all*** available information
- Input representation is ***key***
- Better log likelihood does ***not*** mean the samples are better
  - Early stop on validation likelihood worse. EDIT: Was a bug
- Still overparameterized, more parameter sharing could help
- May need beam search, noisy parallel approximate decoding, or other tricks

# Pie in the Sky

- **Event sequences are general**
  - Alerts in a datacenter
  - Events on CAN bus (automotive)
  - Portfolio of financial timeseries
  - Appliance use in homes
- **Haven't seen others using this representation**
  - How to compare likelihoods to existing methods?

## Thanks!

schuster@

deck@

The Magenta team

The Google Brain team

***You***, for looking at these slides

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kastnerkyle on Twitter, GitHub



# References

<http://people.idsia.ch/~juergen/blues/>

From Douglas Eck. LSTM Blues

<https://github.com/tensorflow/magenta>

From The Magenta team, led by Douglas Eck.

<http://deeplearning.net/tutorial/rnnrbm.html>

From Nicolas Boulanger-Lewandowski et al. RNN-RBM and RBM-NADE

<http://www.hexahedria.com/2015/08/03/composing-music...>

From Daniel Johnson. Results speak for themselves

<https://soundcloud.com/bachbot>

From Feynman Liang. A music bot for harmonization and sampling.

<https://arxiv.org/abs/1506.02216>

From Chung et. al (including me and Laurent Dinh). Short name VRNN

<https://soundcloud.com/graphific/pyotr-lstm-tchaikovsky>

From Roelof Pieters. VRNN with some more tricks

<https://arxiv.org/abs/1605.07571>

A followup to VRNN by Sonderby et. al., strong likelihoods

<http://arxiv.org/abs/1412.7927>

From Goel et. al. RNN-DBN

<https://highnoongmt.wordpress.com/2015/05/22/lisls-stis-recurrent-neural-networks-for-folk-music-generation/>

From Sturm et. al. RNN language model on ABC

<http://arxiv.org/abs/0705.2011>

From Graves et. al. Multidimensional RNN, or MDRNN

<https://arxiv.org/abs/1505.00393>

From Visin et. al (including me). Multidimensional RNN, called ReNet

<https://youtu.be/0qnTaAz-xtQ>

From Pachet et. al., Sony CSL. FlowMachines

<https://www.youtube.com/watch?v=2kuY3BrmTfQ>

From David Cope. EMI

Extra Slides

# Previous approaches

- **Tons of Markov chain, HMM, RNN on ABC (Sturm et. al.)**
  - ABC format has limitations, but this is basically “*cheat mode*” - try it!
- **Sigmoid out, Bernoulli cross-entropy cost (LSTM Blues, Eck et. al.)**
  - Might need extremely deep output to get good “mixing” for density estimate
- **RNN-RBM (Boulanger-Lewandowski et. al.)**
  - Better density estimate, but need contrastive divergence training
- **RNN-NADE (Boulanger-Lewandowski et. al.)**
  - Direct gradient descent, close to my final approach
- **RNN-DBN (Goel et. al.)**
  - Even better density, need to do layerwise training
- **Multi-dimensional RNN (Graves, or Visin et. al.)**
  - Works well but naive version can be very slow during training
- **VRNN-based (Pieters, base model from Chung et. al.)**
  - Seems awesome, probably complementary to other output variations here

# Approach (Week 1-2): Markov chain on ABC

- ABC is cheat mode, in a good way
- Encodes musicality
- abc2midi does automated curation!

