

RECURRENT NEURAL NETWORKS FOR DRUM TRANSCRIPTION

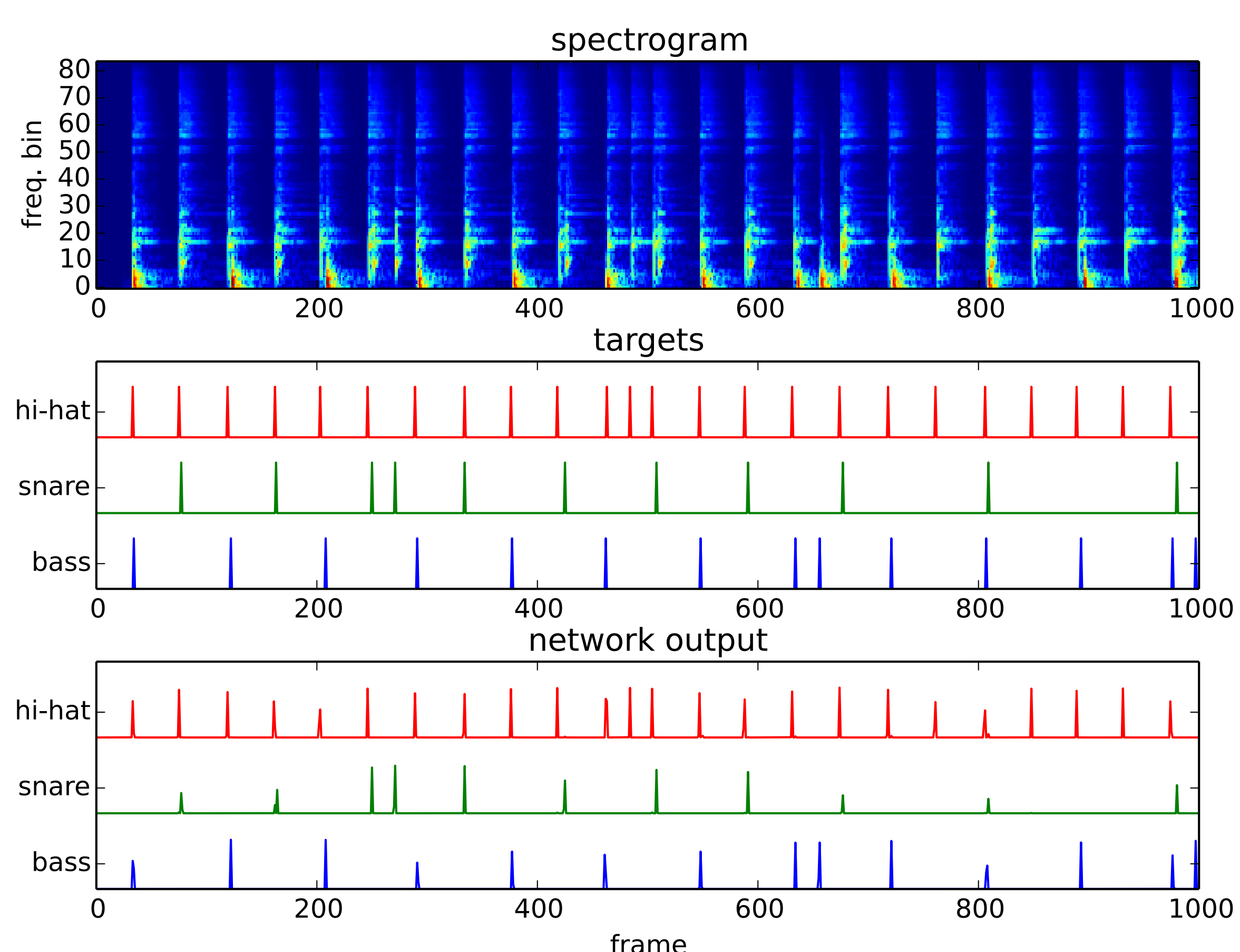
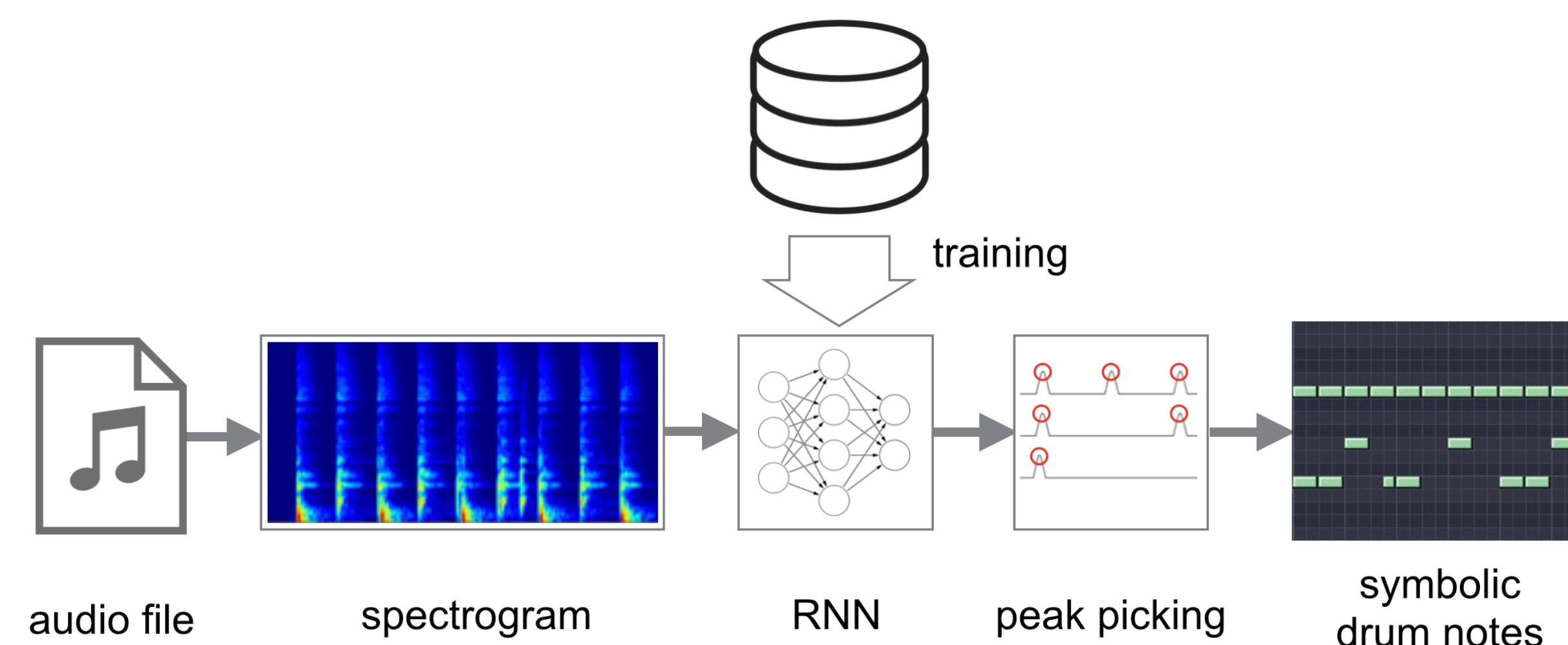
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1 introduction

- > method for transcription of solo drum audio recordings based on neural networks
- > comparison of four RNN architectures for drum transcription
- > evaluation on two well-known datasets

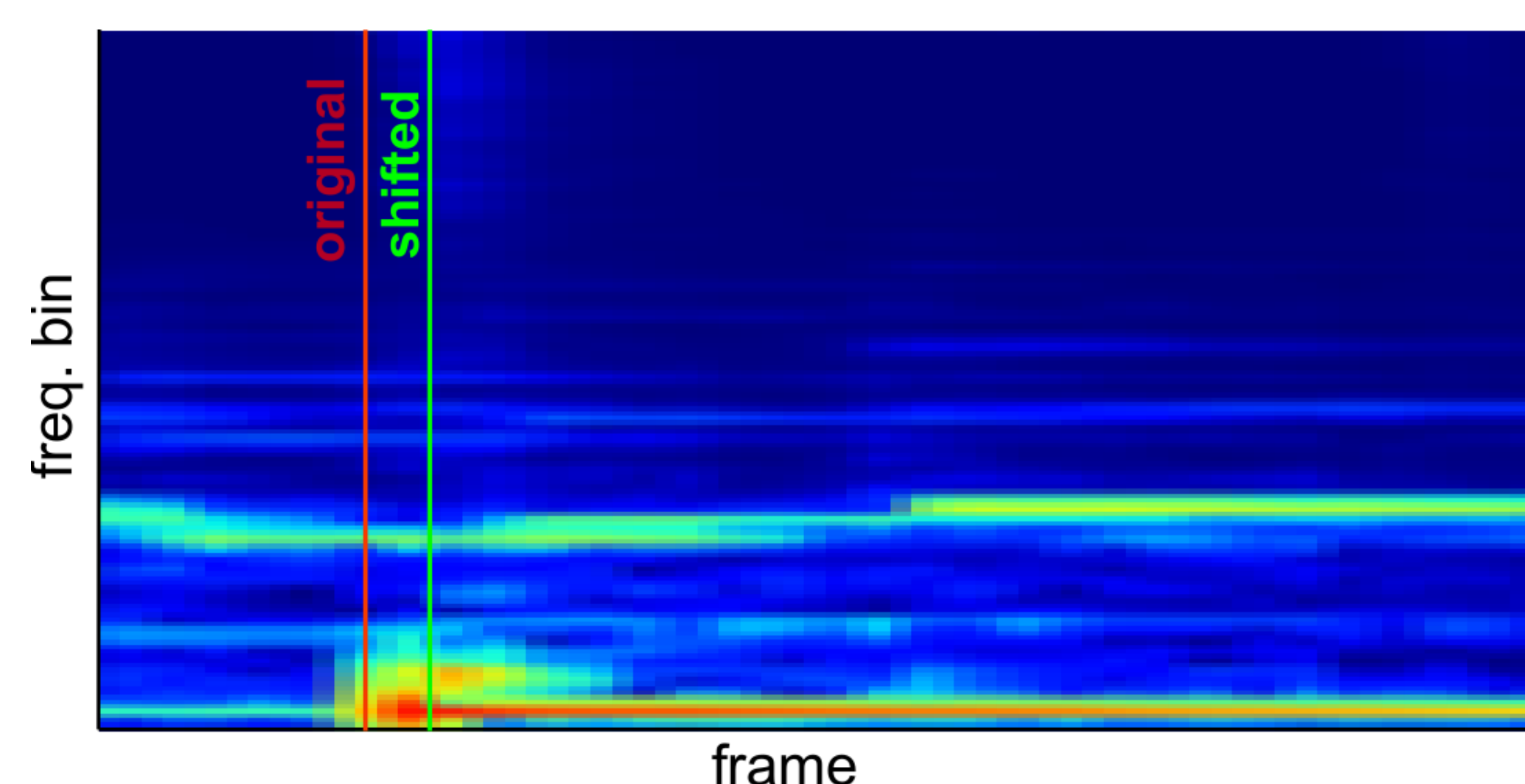
overview of method



2 method

- > extraction of a log. 84-bin power spectrogram @ 100Hz
- > single layer RNNs with 200 nodes
- > output of three detection functions for bass drum, snare drum, and hi-hat
- > peak picking yields notes for the three instruments

time shift of labels



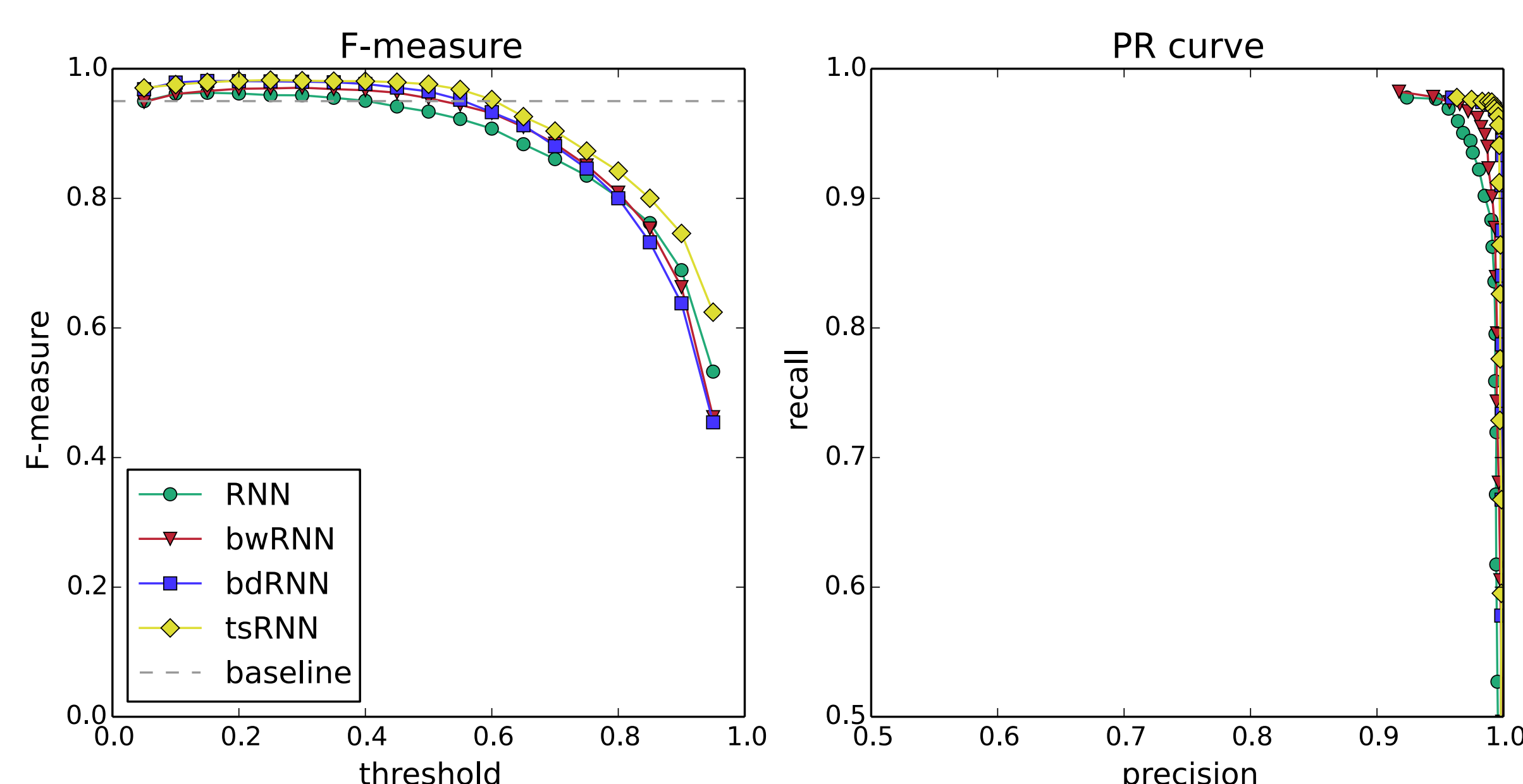
3 evaluated architectures

- > simple RNN
- > backward RNN (bwRNN)
- > bidirectional RNN (bdRNN)
- > RNN with time shift (tsRNN)

results

Results for IDMT-SMT-Drums

| algorithm | best F-measure [%] | at threshold |
|-----------|--------------------|--------------|
| RNN | 96.3 | 0.15 |
| bwRNN | 97.1 | 0.30 |
| bdRNN | 98.1 | 0.15 |
| tsRNN | 98.2 | 0.25 |
| NMF [5] | 95.0 | - |



Results for ENST-Drums (trained on SMT)

| algorithm | best F-measure [%] | at threshold |
|-----------|--------------------|--------------|
| RNN | 69.3 | 0.05 |
| bwRNN | 64.4 | 0.15 |
| bdRNN | 70.3 | 0.05 |
| tsRNN | 73.1 | 0.10 |
| HMM [24] | 81.5 | - |

