



Multitrack Music Transformer

Hao-Wen Dong Ke Chen Shlomo Dubnov Julian McAuley Taylor Berg-Kirkpatrick
University of California San Diego



UC San Diego

Overview

Generate **orchestral** music

- of **diverse instruments**
- using a **new compact representation**
- with a **multi-dimensional transformer**



Demo



(Source: Vienna Mozart Orchestra)



Related Work (Transformers for Music Generation)

Model	Multitrack	Instrument control	Compound tokens	Generative modeling
REMI [5]				✓
MMM [10]	✓			✓
CP [6]			✓	✓
MusicBERT [15]	✓		✓	
FIGARO [11]	✓			✓
MMT (ours)	✓	✓	✓	✓

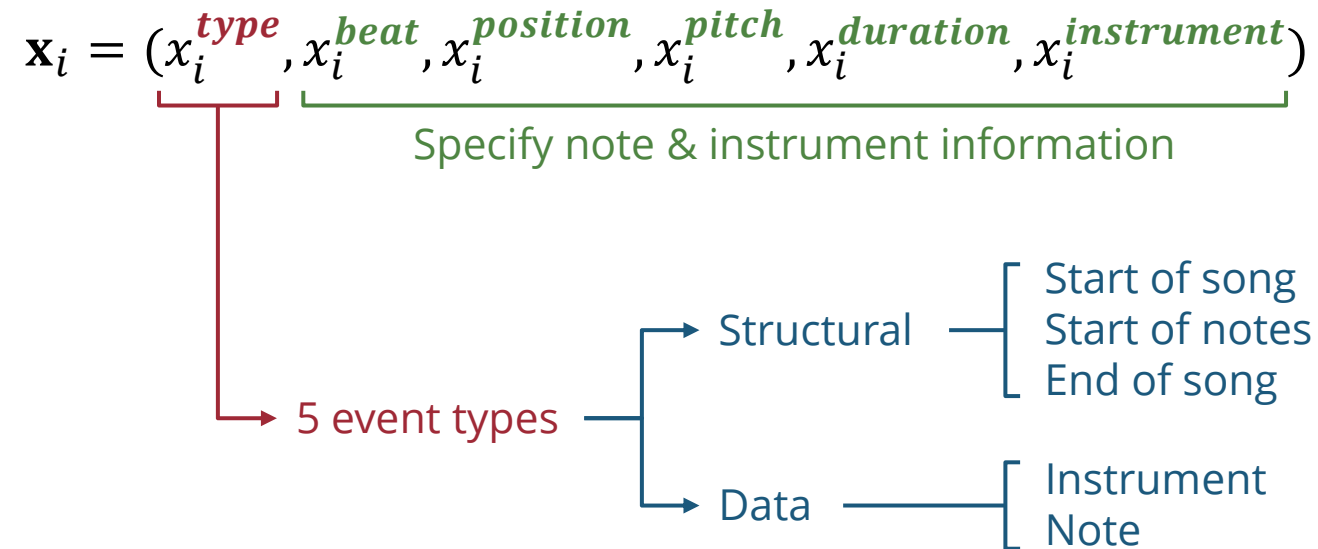
Huang and Yang, "Pop Music Transformer: Beat-based Modeling and Generation of Expressive Pop Piano Compositions," *Proc. MM*, 2020.
Ens and Pasquier, "MMM : Exploring Conditional Multi-Track Music Generation with the Transformer," *arXiv preprint arXiv:2008.06048*, 2020.
Hsiao et al., "Compound Word Transformer: Learning to Compose Full-Song Music over Dynamic Directed Hypergraphs," *Proc. AAAI*, 2023.
Zeng et al., "MusicBERT: Symbolic Music Understanding with Large-Scale Pre-Training," *Proc. Findings of ACL*, 2021.
von Rütte et al., "FIGARO: Controllable Music Generation using Learned and Expert Features," *Proc. ICLR*, 2023.

Representation

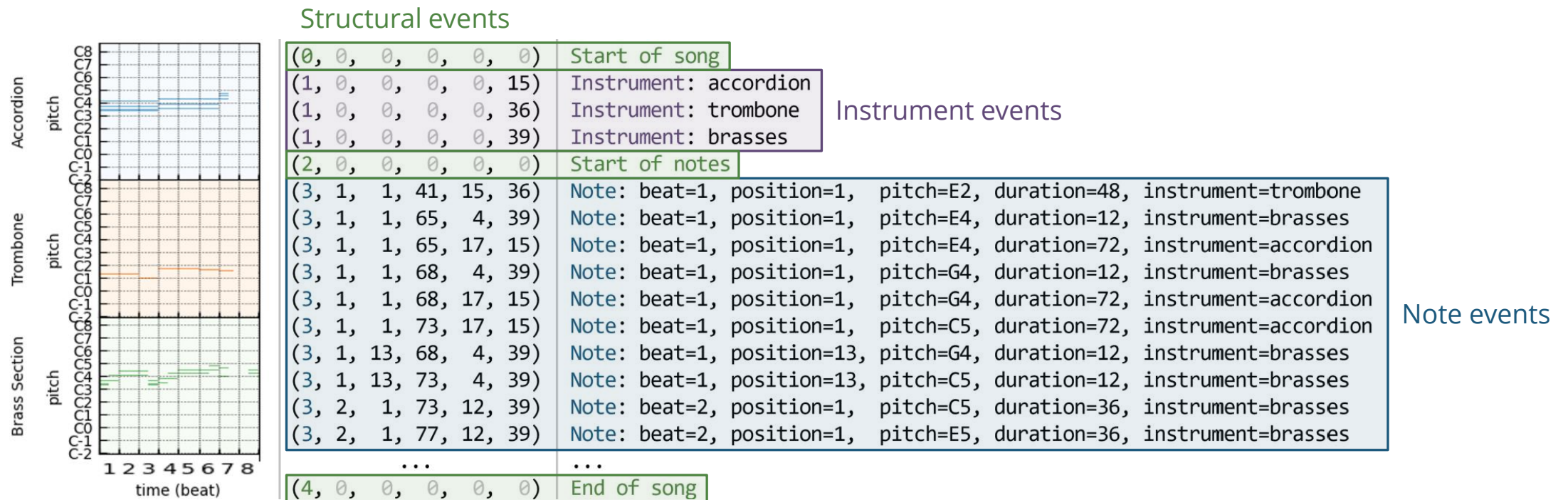
- We represent a music piece as a sequence of events

$$\mathbf{x} = (\mathbf{x}_1, \dots, \mathbf{x}_n)$$

- Each event \mathbf{x}_i is encoded as

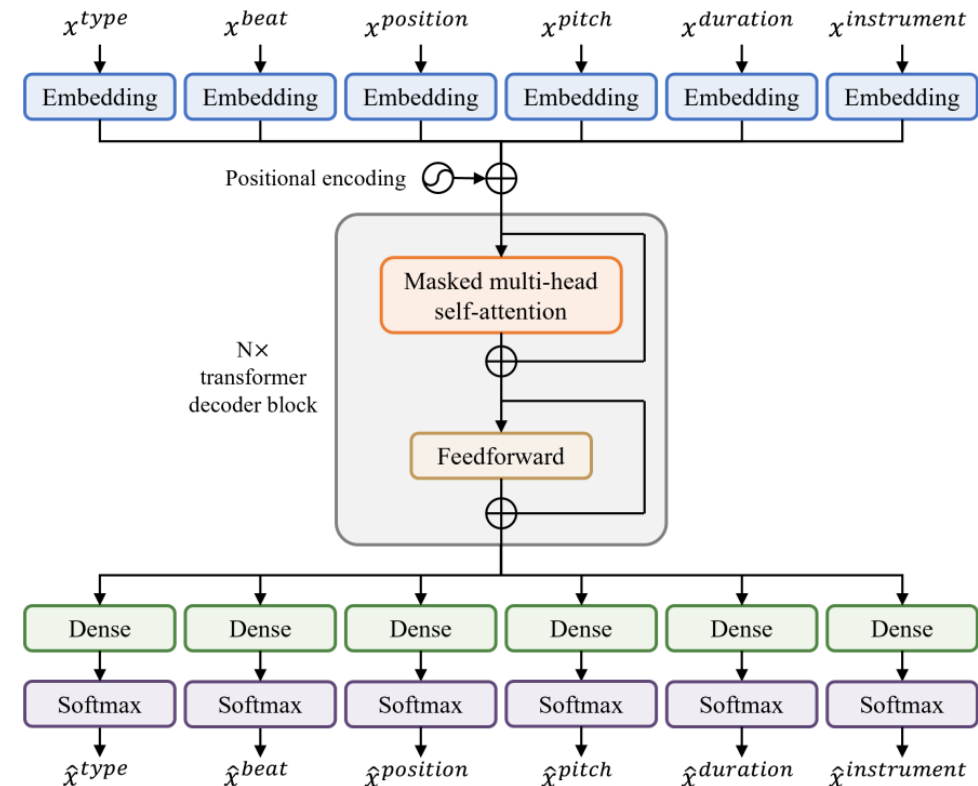


Representation (An Example)



Multitrack Music Transformer

- A multi-dimensional decoder-only transformer model
 - Predict six fields *at the same time*
- Trained autoregressively
 - Predict the next event given past events
- At inference time, illegal values are assigned zero probabilities
 - Violate the ordering of structural events
 - Violate the hierarchical sorting of events



Three Sampling Modes

Unconditional generation

Input

(0, 0, 0, 0, 0, 0)	Start of song
(1, 0, 0, 0, 0, 15)	Instrument: accordion
(1, 0, 0, 0, 0, 36)	Instrument: trombone
(1, 0, 0, 0, 0, 39)	Instrument: brasses
(2, 0, 0, 0, 0, 0)	Start of notes
(3, 1, 1, 41, 15, 36)	Note: beat=1, position=1, pitch=E2, duration=48, instrument=trombone
(3, 1, 1, 65, 4, 39)	Note: beat=1, position=1, pitch=E4, duration=12, instrument=brasses
(3, 1, 1, 65, 17, 15)	Note: beat=1, position=1, pitch=E4, duration=72, instrument=accordion
(3, 1, 1, 68, 4, 39)	Note: beat=1, position=1, pitch=G4, duration=12, instrument=brasses
(3, 1, 1, 68, 17, 15)	Note: beat=1, position=1, pitch=G4, duration=72, instrument=accordion
(3, 1, 1, 73, 17, 15)	Note: beat=1, position=1, pitch=C5, duration=72, instrument=accordion
(3, 1, 13, 68, 4, 39)	Note: beat=1, position=13, pitch=G4, duration=12, instrument=brasses
(3, 1, 13, 73, 4, 39)	Note: beat=1, position=13, pitch=C5, duration=12, instrument=brasses
(3, 2, 1, 73, 12, 39)	Note: beat=2, position=1, pitch=C5, duration=36, instrument=brasses
(3, 2, 1, 77, 12, 39)	Note: beat=2, position=1, pitch=E5, duration=36, instrument=brasses
...	...
(4, 0, 0, 0, 0, 0)	End of song

Only need to train ONE model!

Instrument-informed generation

Input

(0, 0, 0, 0, 0, 0)	Start of song
(1, 0, 0, 0, 0, 15)	Instrument: accordion
(1, 0, 0, 0, 0, 36)	Instrument: trombone
(1, 0, 0, 0, 0, 39)	Instrument: brasses
(2, 0, 0, 0, 0, 0)	Start of notes
(3, 1, 1, 41, 15, 36)	Note: beat=1, position=1, pitch=E2, duration=48, instrument=trombone
(3, 1, 1, 65, 4, 39)	Note: beat=1, position=1, pitch=E4, duration=12, instrument=brasses
(3, 1, 1, 65, 17, 15)	Note: beat=1, position=1, pitch=E4, duration=72, instrument=accordion
(3, 1, 1, 68, 4, 39)	Note: beat=1, position=1, pitch=G4, duration=12, instrument=brasses
(3, 1, 1, 68, 17, 15)	Note: beat=1, position=1, pitch=G4, duration=72, instrument=accordion
(3, 1, 1, 73, 17, 15)	Note: beat=1, position=1, pitch=C5, duration=72, instrument=accordion
(3, 1, 13, 68, 4, 39)	Note: beat=1, position=13, pitch=G4, duration=12, instrument=brasses
(3, 1, 13, 73, 4, 39)	Note: beat=1, position=13, pitch=C5, duration=12, instrument=brasses
(3, 2, 1, 73, 12, 39)	Note: beat=2, position=1, pitch=C5, duration=36, instrument=brasses
(3, 2, 1, 77, 12, 39)	Note: beat=2, position=1, pitch=E5, duration=36, instrument=brasses
...	...
(4, 0, 0, 0, 0, 0)	End of song

N-beat continuation

Input

(0, 0, 0, 0, 0, 0)	Start of song
(1, 0, 0, 0, 0, 15)	Instrument: accordion
(1, 0, 0, 0, 0, 36)	Instrument: trombone
(1, 0, 0, 0, 0, 39)	Instrument: brasses
(2, 0, 0, 0, 0, 0)	Start of notes
(3, 1, 1, 41, 15, 36)	Note: beat=1, position=1, pitch=E2, duration=48, instrument=trombone
(3, 1, 1, 65, 4, 39)	Note: beat=1, position=1, pitch=E4, duration=12, instrument=brasses
(3, 1, 1, 65, 17, 15)	Note: beat=1, position=1, pitch=E4, duration=72, instrument=accordion
(3, 1, 1, 68, 4, 39)	Note: beat=1, position=1, pitch=G4, duration=12, instrument=brasses
(3, 1, 1, 68, 17, 15)	Note: beat=1, position=1, pitch=G4, duration=72, instrument=accordion
(3, 1, 1, 73, 17, 15)	Note: beat=1, position=1, pitch=C5, duration=72, instrument=accordion
(3, 1, 13, 68, 4, 39)	Note: beat=1, position=13, pitch=G4, duration=12, instrument=brasses
(3, 1, 13, 73, 4, 39)	Note: beat=1, position=13, pitch=C5, duration=12, instrument=brasses
(3, 2, 1, 73, 12, 39)	Note: beat=2, position=1, pitch=C5, duration=36, instrument=brasses
(3, 2, 1, 77, 12, 39)	Note: beat=2, position=1, pitch=E5, duration=36, instrument=brasses
...	...
(4, 0, 0, 0, 0, 0)	End of song

Experimental Setup

Data

- Symbolic Orchestral Database (SOD)
(Crestel et al., 2017)
 - 5,743 songs, 357 hours
- Temporal resolution: 12 time steps per quarter note
- 80% training, 10% validation, 10% test
- Data augmentation
 - Randomly shift for -5~6 semitones
 - Randomly select a starting beat

Model & Training

- 6 transformer decoder blocks
- 8 attention heads
- Model dimension: 512
- Sequence length: 1024
- Maximum number of beats: 256
- Maximum training steps: 200,000

Example Results

Unconditional
generation



Instrument-informed
generation



church-organ, viola,
contrabass, strings,
voices, horn, oboe

4-beat continuation



Wolfgang Amadeus Mozart's
Eine kleine Nachtmusik



More audio samples



salu133445.github.io/mmt/

Subjective Listening Test Results

	Number of parameters	Average sample length (sec)	Inference speed (notes per second)	Subjective listening test results			
				Coherence	Richness	Arrangement	Overall
MMM [10]	19.81 M	38.69	5.66	3.48 ± 0.35	3.05 ± 0.38	3.28 ± 0.37	3.17 ± 0.43
REMI+ [11]	20.72 M	28.69	3.58	3.90 ± 0.52	3.74 ± 0.21	3.74 ± 0.44	3.77 ± 0.41
MMT (ours)	19.94 M	100.42	11.79	3.55 ± 0.46	3.53 ± 0.35	3.40 ± 0.44	3.33 ± 0.47

2.6x/3.5x longer
generated samples
(within the same sequence length)

2.1x/3.3x faster
inference speed

Higher quality than MMM
Lower quality than REMI+

Analyzing Self-attention

- *Mean relative attention* for a field d :

$$\gamma_k^{(d)} = \frac{\sum_{x \in \mathcal{D}} \sum_{s > t} \boxed{a_{s,t}(\mathbf{x})} \boxed{1_{x_t^{(d)} - x_s^{(d)} = k}}}{\sum_{x \in \mathcal{D}} \sum_{s > t} a_{s,t}(\mathbf{x})}$$

Attention weight
Whether the field value is of difference k

(0, 0, 0, 0, 0, 0)	Start of song
(1, 0, 0, 0, 0, 15)	Instrument: accordion
(1, 0, 0, 0, 0, 36)	Instrument: trombone
(1, 0, 0, 0, 0, 39)	Instrument: brasses
(2, 0, 0, 0, 0, 0)	Start of notes
(3, 1, 1, 41, 15, 36)	Note: beat=1, position=1, pitch=E2, duration=48, instrument=trombone
(3, 1, 1, 65, 4, 39)	Note: beat=1, position=1, pitch=E4, duration=12, instrument=brasses
(3, 1, 1, 65, 17, 15)	Note: beat=1, position=1, pitch=E4, duration=72, instrument=accordion
(3, 1, 1, 68, 4, 39)	Note: beat=1, position=1, pitch=G4, duration=12, instrument=brasses
(3, 1, 1, 68, 17, 15)	Note: beat=1, position=1, pitch=G4, duration=72, instrument=accordion
(3, 1, 1, 73, 17, 15)	Note: beat=1, position=1, pitch=C5, duration=72, instrument=accordion

$\gamma_{-8}^{(pitch)}$ (blue arrow from row 3 to row 6)
 $\gamma_{-5}^{(pitch)}$ (green arrow from row 5 to row 8)

Analyzing Self-attention

- *Mean relative attention* for a field d :

$$\gamma_k^{(d)} = \frac{\sum_{x \in \mathcal{D}} \sum_{s > t} a_{s,t}(\mathbf{x}) \mathbf{1}_{x_t^{(d)} - x_s^{(d)} = k}}{\sum_{x \in \mathcal{D}} \sum_{s > t} a_{s,t}(\mathbf{x})}$$

Biased towards difference that occurred more frequently!

- *Mean relative attention gain* for a field d :

$$\tilde{\gamma}_k^{(d)} = \gamma_k^{(d)} - \frac{\sum_{x \in \mathcal{D}} \sum_{s > t} \mathbf{1}_{x_t^{(d)} - x_s^{(d)} = k}}{\sum_{x \in \mathcal{D}} \sum_{s > t} \boxed{1}}$$

Assuming a uniform attention matrix

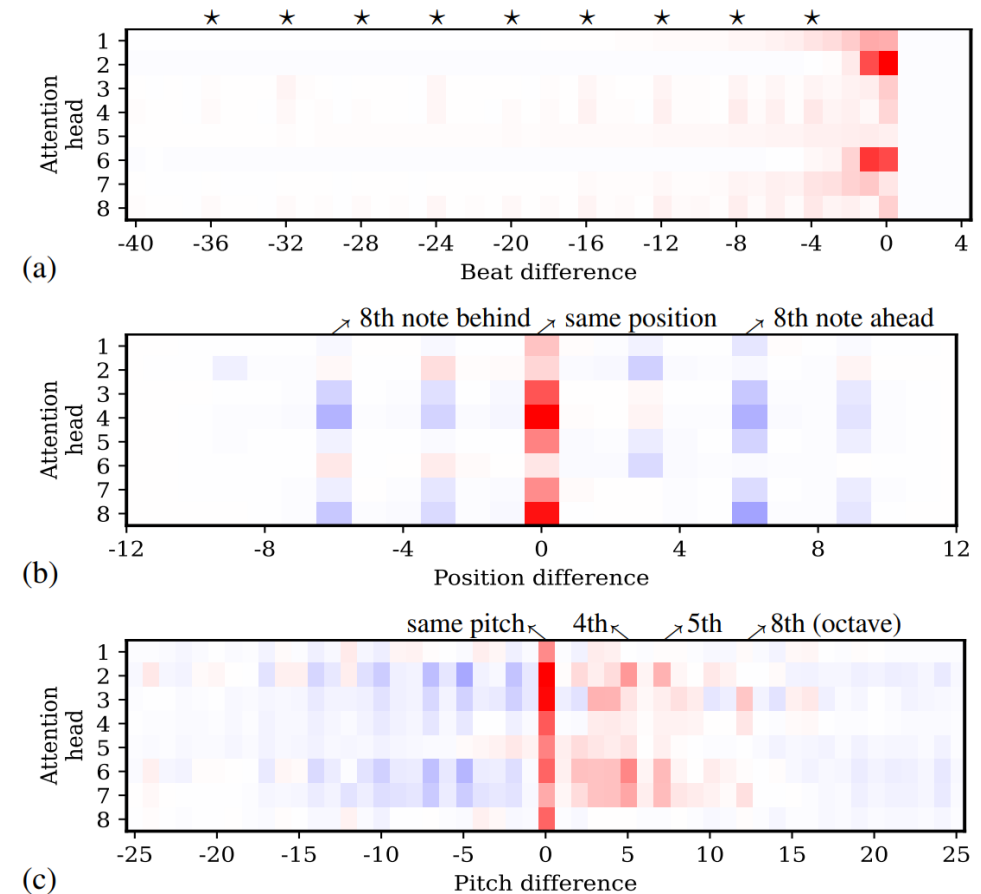
Musical Self-attention

The MMT model attends more to notes

- that are $4N$ beats away in the past
- that have the same position as the current note (A note on beat attends more to a note on beat; a note off beat attends more to a note off beat.)
- that has a pitch in an octave above which forms a consonant interval

→ MMT learns a **relative self-attention** for certain aspects of music, specifically, **beat**, **position** and **pitch**.

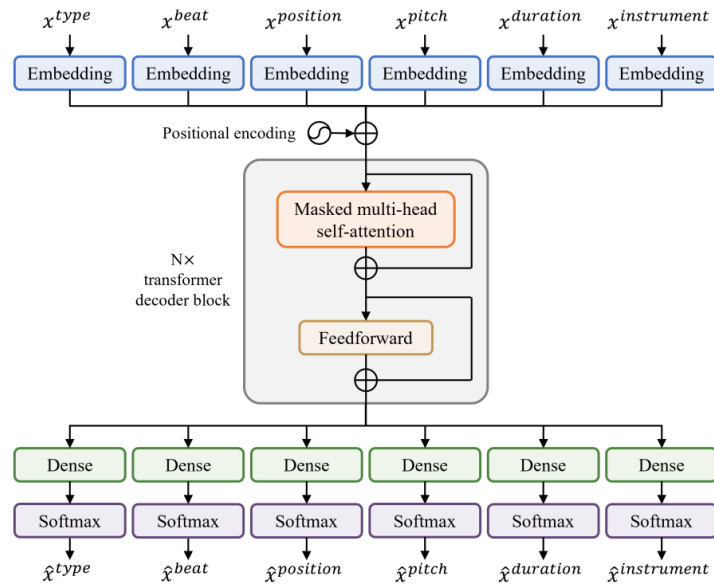
Positive and negative mean relative attention gain



Summary

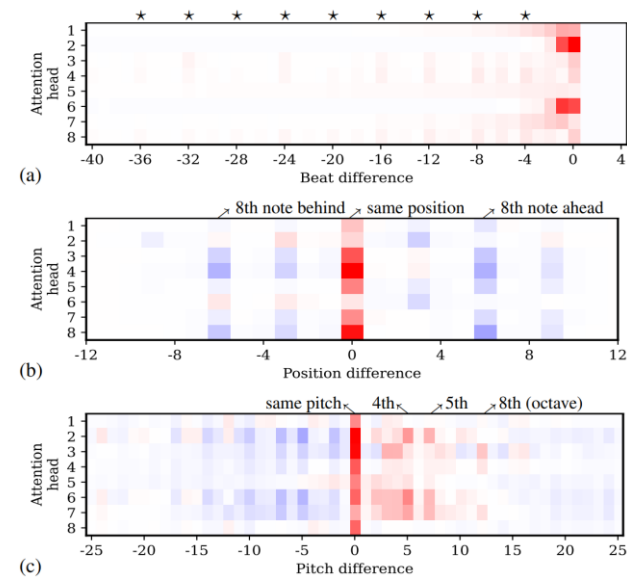
Multitrack Music Transformer

Proposed an efficient representation and model for multitrack music generation



Musical Self-attention

Presented the first systematic analysis of musical self-attention



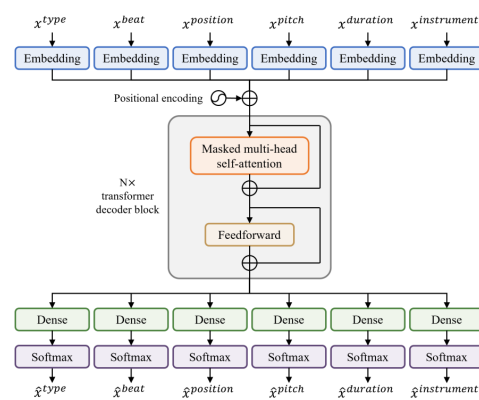
Acknowledgements

- Hao-Wen thanks J. Yang and Family Foundation and Taiwan Ministry of Education for supporting his PhD study.
- This project has received funding from the European Research Council (ERC REACH) under the European Union's Horizon 2020 research and innovation programme (Grant agreement #883313).

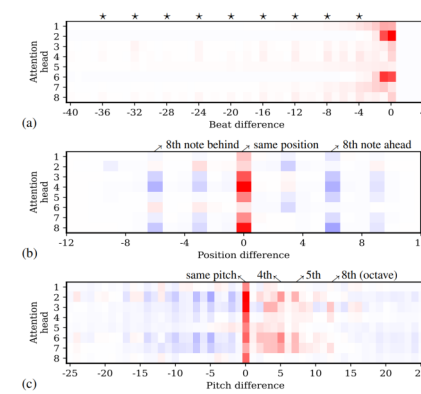


Thank you!

Multitrack Music Transformer



Musical Self-attention



Paper: arxiv.org/abs/2207.06983 Demo: salu133445.github.io/mmt/ Code: github.com/salu133445/mmt