

Collaboration of Two AI Musicians: Challenges and Possibilities

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Environment

- Fixed music scores
- Real-time interaction
- Live performance
- Acoustic communication

Objectives

- Each AI can function alone
- Together, two AIs “interpret” their respective parts in sync

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- Each AI can function alone
 - can play its part (i.e. score → audio)
 - can at the same time collaborate with a human
- Together, two AIs “interpret” their respective parts in sync
 - each AI does not know the partner's identity
 - the overall effect should be human-like

Relevant capabilities

- Playing
- Listening

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 - turn a score into audio in real-time, with flexibility in timing and dynamics (relatively easy for MIDI piano, more difficult for voice)
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- Listening
 - identifying input
 - score following (“alignment”)

Relevant capabilities

- Playing

- For MIDI piano, we have a system that:
 - writes a score as an array of dimension $2N \times 6$
each note consists of 1 note-on, 1 note-off event
event format:
[index, type, note, score position, time, velocity]
 - can adjust (time, velocity) of each note
at any time, following “musical intent”
 - outputs the array in real-time
as MIDI performance

Relevant capabilities

- Listening

- Identifying input
- Score alignment

Relevant capabilities

- Listening

- Identifying input
 - we want essentially an Audio to MIDI module, aka “transcription”
- Score alignment
 - from the MIDI data, we can fill in the score array
 - our new “search-by-beat” protocol, used for automating data processing (for SMC submission) works reasonably well

Relevant capabilities

- Listening

- Identifying input
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Relevant capabilities

- Listening

- Identifying input
Pitch detection

Relevant capabilities

- Listening

- Identifying input

Pitch detection

- Problems:

noise

Two simultaneous parts present in audio
Pitch detection algorithm “confused”

echo

Peaks at 200Hz, 300Hz
→ Pitch detection returns 100Hz?

single pitch

human intonation inaccuracy

MIDI generated thus will not be a
good representation of intent

Relevant capabilities

- Listening

- Can we circumvent this problem?
- How do human musicians do it?

Relevant capabilities

- Listening

- Can we circumvent this problem?
- How do human musicians do it?
- One piece of information we have not used:
the score
- I think humans do not fully transcribe; rather,
they start from the score,
identify salient events on both sides (score, audio),
confirm a match upon “reasonably certain” detection

Relevant capabilities

- Listening

Idea for “query-based” listening:

no separation of tasks into detection, alignment

- At every time t , calculate initial hypothesis for $P(t)$, “score position corresponding to the current time”, by extrapolating assuming constant tempo
- Salient note-on events in the vicinity of $P(t)$ are checked against the audio (“salient” = “unique”)
- Find probability that the audio contains the event
- If threshold not met, initial hypothesis holds;
If threshold met, $P(t)$ is set to the score position of the matched event

Relevant capabilities

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Issues

- “Probability that the audio contains the event”: absolute or time-based?
- Acoustic effects: it will always get slower

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Issues

- “Probability that the audio contains the event”: absolute or time-based?
- Acoustic effects: it will always get slower
 - we do not know the attack curve
 - so when a match with score event E occurs, we can only surmise that $P(t) \geq p(E)$
 - thus $P(t)$ should be adjusted only positively
 - stability problem? “Mutual acceleration”?

Relevant capabilities

- Listening

- Of course, we can do even better
- More available information: the listener's own performance
- If we can “subtract” the expected acoustic effect from the audio, we could isolate the partner's performance

Putting it together

- Two machines, each following this listening protocol
- Each perceives a series of identified couplets (time, score position)
- By linear interpolation & Kuramoto model, it generates a fitted accompaniment