

声学大讲堂 _{音频产业创新技术公益讲座}

して 博音听力



张逸霄 伦敦玛丽女王大学 2024.9.2



声学大讲堂 音频产业创新技术公益讲座

助力引领区建设科创中国浦东行



张逸霄

研究方向

可控音乐生成、跨模态音乐编辑,以及大语言模型在音乐中的应用

教育经历

- 2020-2024, 伦敦玛丽女王大学, 博士生
 - 导师: Prof. Simon Dixon & Dr. Mark Levy (Apple Inc.)
- 2015-2019, 电子科技大学, 工学学士

实习经历:

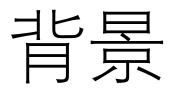
- 2024.5-至今, Stability Al
- 2023.10-2024.5, Sony Al Tokyo
- 2023.6-2023.9, Yamaha R&D
- 2019.9-2020.10, NYU Shanghai



目录

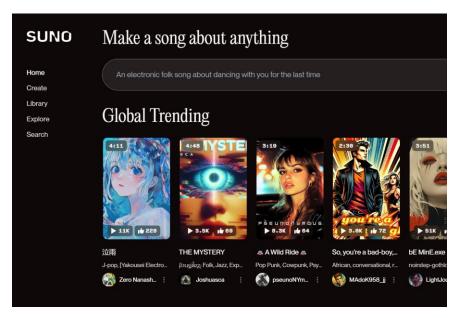
• 背景

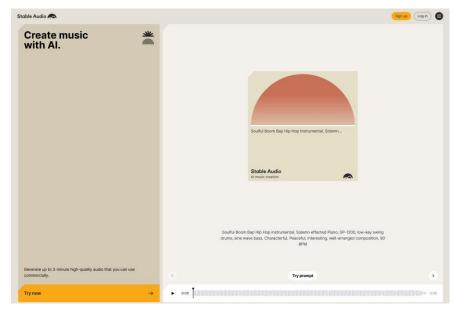
- 文本到音乐生成:任务描述
- 符号还是音频: 两类表示法
- 大模型时代的两种范式
- 为什么需要增强模型的控制能力?
- 四种可行的思路
- 预训练模型: MusicGen
- 音乐大语言模型的内容控制
 - Content-based Controls For Music Large Language Modeling, ISMIR 2024
 - Liwei Lin, Gus Xia, Junyan Jiang, Yixiao Zhang
- 编配、内绘和润色: 通过基于内容的控制来指导长期音乐音频生成和编辑
 - Arrange, Inpaint, and Refine: Steerable Long-term Music Audio Generation and Editing via Content-based Controls, IJCAI 2024
 - Liwei Lin, Gus Xia, <u>Yixiao Zhang</u>, Junyan Jiang
- 指令微调的音乐大模型: 多任务音乐编辑
 - Instruct-MusicGen: Unlocking Text-to-Music Editing for Music Language Models via Instruction Tuning, arXiv/2405.18386
 - Yixiao Zhang, Yukara Ikemiya, Naoki Murata, Woosung Choi, Marco Martínez, Liwei Lin, Gus Xia, Wei-Hsiang Liao, Yuki Mitsufuji, Simon Dixon
- 音乐大模型为什么仍然不能像人类一样合作?
 - The Interpretation Gap in Text-to-Music Generation Models, NLP4MusA Workshop @ ISMIR 2024
 - Yongyi Zang*, Yixiao Zhang*
- Q&A



文本到音乐生成: 任务描述

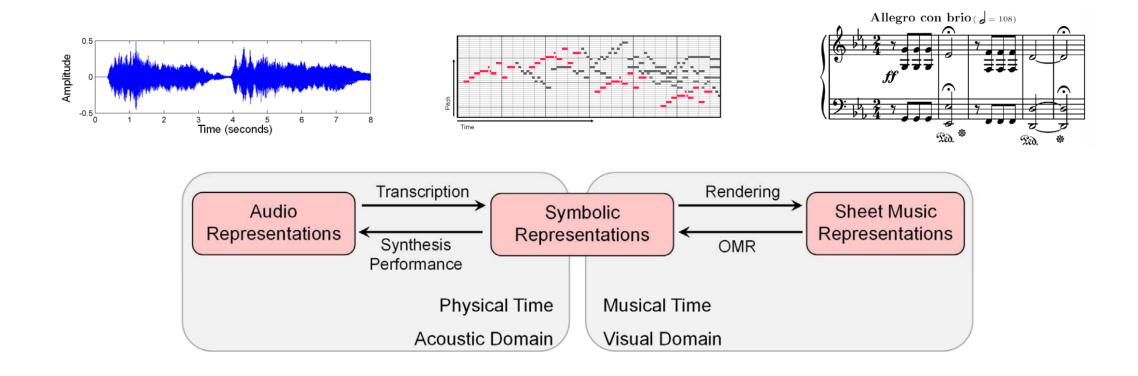
• 给定音乐的文本描述,模型输出符合描述的音乐片段



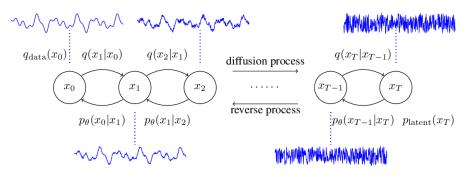


Suno Al Stable Audio 2.0

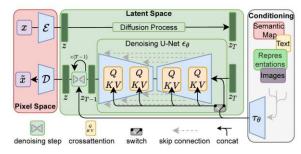
符号还是音频: 两类音乐表示法



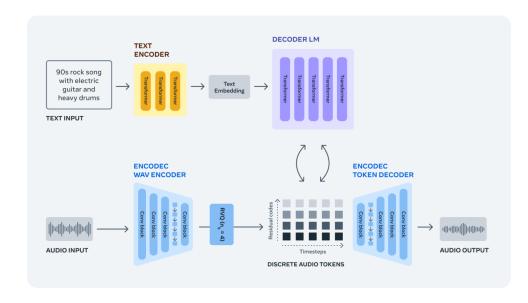
大模型时代的两种流行范式



[2] Kong, Z., Ping, W., Huang, J., Zhao, K., & Catanzaro, B. DiffWave: A Versatile Diffusion Model for Audio Synthesis. In International Conference on Learning Representations.



[3] Rombach, R., Blattmann, A., Lorenz, D., Esser, P., & Ommer, B. (2022). High-resolution image synthesis with latent diffusion models. In Proceedings of the IEEE/CVF conference on computer vision and pattern recognition (pp. 10684-10695).



[4] Copet, J., Kreuk, F., Gat, I., Remez, T., Kant, D., Synnaeve, G., ... & Défossez, A. (2024). Simple and controllable music generation. Advances in Neural Information Processing Systems, 36.

扩散模型 (Diffusion Models)

自回归语言模型 (Autoregressive Language Models)

为什么需要增强模型的可控能力?

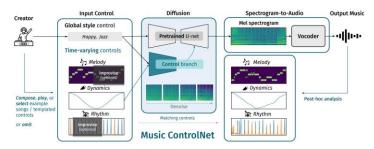
• 文本到音乐生成模型,只接受文本描述作为输入

- 对于音乐创作,用户需要更多控制:
 - 和弦进行、主旋律、鼓点、音色…

四种可行的思路

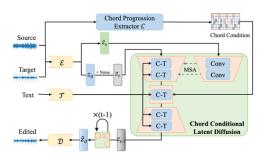
• 在预训练大模型上增加控制模块

- 在预训练阶段增加
- 在微调阶段增加



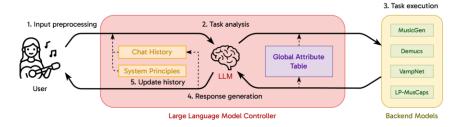
[5] Wu, S. L., Donahue, C., Watanabe, S., & Bryan, N. J. (2024). Music controlnet: Multiple time-varying controls for music generation. IEEE/ACM Transactions on Audio, Speech, and Language Processing, 32, 2692-2703.

• 训练单独的音乐编辑模型



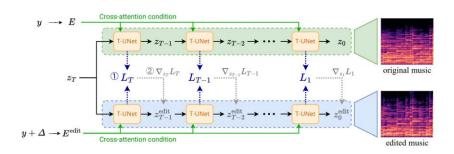
[6] Han, B., Dai, J., Hao, W., He, X., Guo, D., Chen, J., ... & Song, X. (2023). InstructME: An instruction guided music edit and remix framework with latent diffusion models. IJCAI 2024.

• 以代理(Agent)方式协调多种大模型



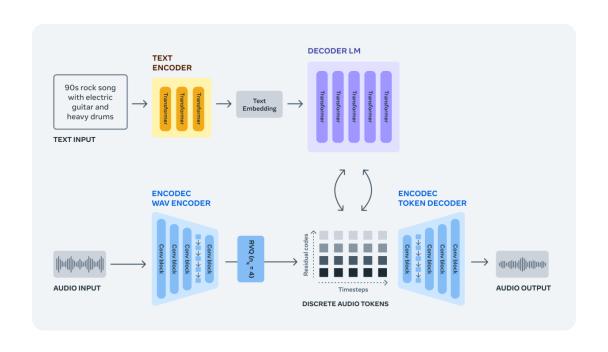
[7] Zhang, Y., Maezawa, A., Xia, G., Yamamoto, K., & Dixon, S. (2023). Loop copilot: Conducting ai ensembles for music generation and iterative editing. arXiv preprint arXiv:2310.12404.

• 在推理阶段介入控制



[8] Zhang, Y., Ikemiya, Y., Xia, G., Murata, N., Martínez, M., Liao, W. H., ... & Dixon, S. (2024). MusicMagus: Zero-shot text-to-music editing via diffusion models. *arXiv preprint* arXiv:2402.06178.

预训练模型: MusicGen

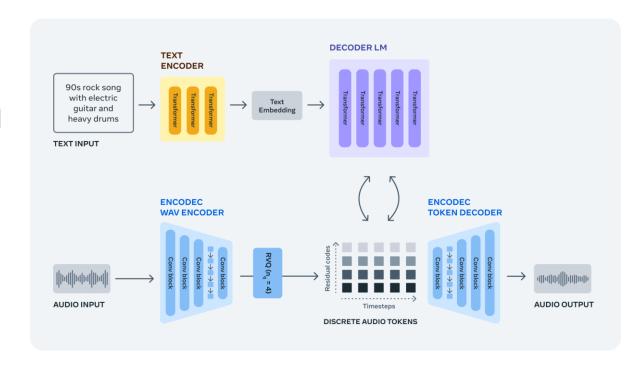


MusicGen包括三个组件:

- 1. 文本编码器
- 2. EnCodec

3. 潜空间的Transformer

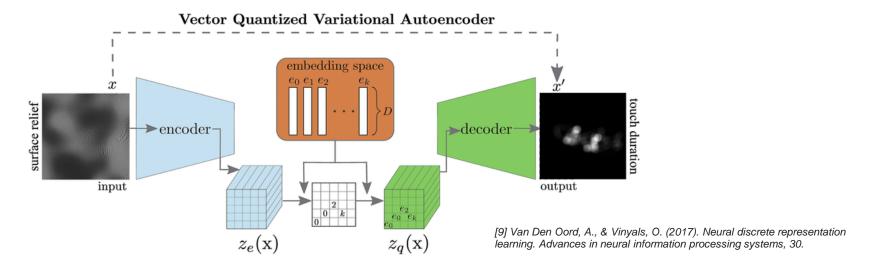
预训练模型: MusicGen

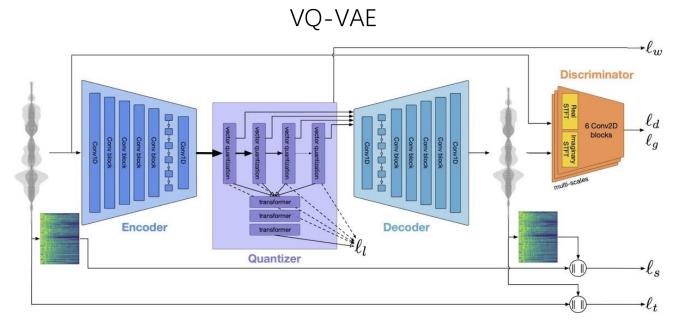


MusicGen包括**三个组件**:

- 1. 文本编码器:将字符串翻译为编码向量
- 2. EnCodec: 将**连续的音乐波形**(32kHz)压缩成**离散的表示**(50Hz)
- 3. 潜空间的Transformer: 负责建模音乐序列

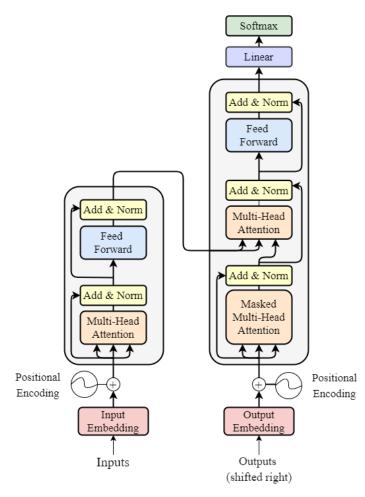
EnCodec





EnCodec [10] Défossez, A., Copet, Audio Compression. Tran

Transformer



每一层包含:

• 一个自注意力层

• 一个交叉注意力层

• 一个前馈层

音乐大语言模型的内容控制

Content-based Controls For Music Large Language Modeling, ISMIR 2024 Liwei Lin, Gus Xia, Junyan Jiang, <u>Yixiao Zhang</u>

论文主要贡献

• Coco-mulla在MusicGen上增加了旋律、鼓点、和弦进行的控制

• 仅在300首歌上微调原模型4%数量的新参数, 5小时完成训练

We equip <u>MusicGen</u>, a text-to-music generation model, with direct and content-based controls on innate music languages such as pitch, chords and drum track. To this end, we contribute Coco-Mulla, a **co**ntent-based **co**ntrol method for **music large la**nguage modeling. It uses a parameter-efficient fine-tuning (PEFT) method tailored for Transformer-based audio models. Our approach achieved low-resource semi-supervised learning, tuning with less than 4% parameters compared to the original model and training on a small dataset set with fewer than 300 unannotated songs. We illustrate the **chords** and **rhythms** control power of the model. Moreover, by combining **piano roll** and text descriptions, our system enables flexible music variation generation and style transfer.

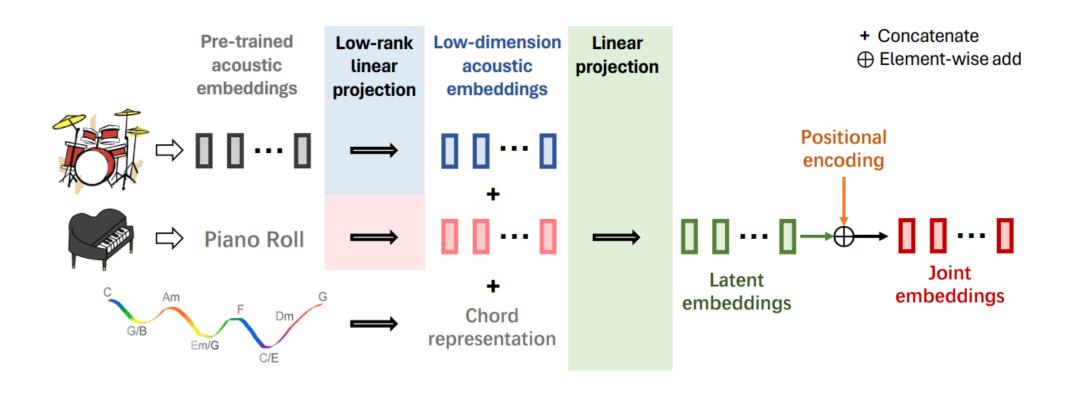
See more details in our paper and try our model via github [coming soon] or huggingface [coming soon].

Chord and Drums Condition

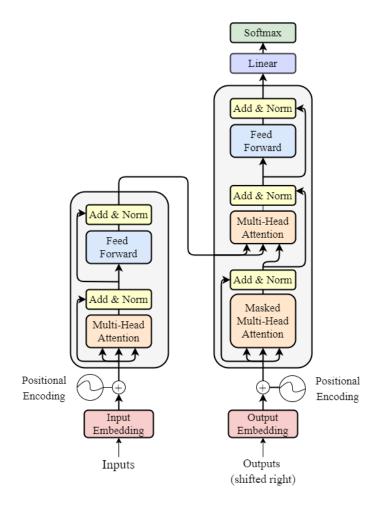
Chord and rhythm controls via symbolic chord representation and acoustic drum tracks. We transcribe chord progression of the generated audio samples using a chord recognition model.

	Bb maj	Fmaj	Eb maj	Bb maj	Fmaj	Eb maj		Gmin	Fmaj	Eb maj	ВЬ
timated Chords of the Generated Sample	6min 00:00 Bb mai	-	F1.			-	Bb maj/3 Cmin Bb maj	Gmin		Et	ВЬ
Conditioned Chords	Gmin	Fmaj	Eb maj	Bb maj	Fmaj	Eb maj	Bb maj/3 Cmin Bb mai	Gmin	Fmaj	Eb maj	ВБ
	Conditioned Drums		captivati effortles skillfully melody	composition features a ing saxophone solo that sly melds with piano chords, weaving its way through the with languid grace. Instrument ine, piano, drums.	acoustic guitar guitar, drums	with a flute solo and chords. instrument: fl		o with a swing melody. piano, drums	thunderou fanfares, a	chestral arrangemer s percussion, epic br nd soaring strings, co atmosphere fit for a	ass reating a
Sample 001	▶ 0:00 / 0:20 —	• •	▶ 0:0	00 / 0:20	: 0:00 / 0:20	•	▶ 0:00 / 0:2	0 —— •)	▶ 0:00 / 0:20	<u> </u>	• :
Sample 002	► 0:00 / 0:20 —		▶ 0:0	00 / 0:20	▶ 0:00 / 0:20	•	 	•) 	• :
Sample 003	▶ 0:00 / 0:20	• •	▶ 0:0	00 / 0:20	▶ 0:00 / 0:20	•	▶ 0:00 / 0:2	0 —— •)	▶ 0:00 / 0:20		• :
Sample 004	► 0:00 / 0:20 —	• • :	▶ 0:0	00 / 0:20	▶ 0:00 / 0:20	•	▶ 0:00 / 0:2	0 ——	▶ 0:00 / 0:20		• :
Sample 005	► 0:00 / 0:20 -	— • :	▶ 0:0	00 / 0:20	:	•	. 0:00 / 0:2	0 ——	₽ 0:00 / 0:20		• :

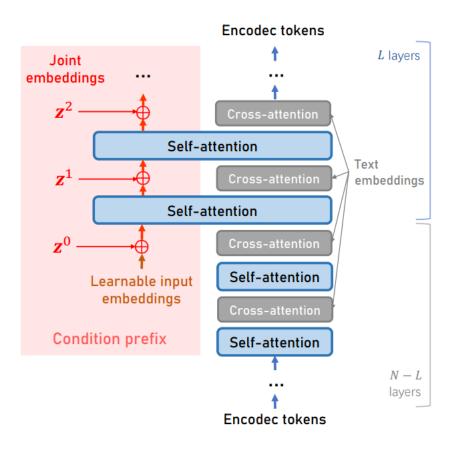
方法(1): 多个条件的编码嵌入



方法(2): 条件生成结构

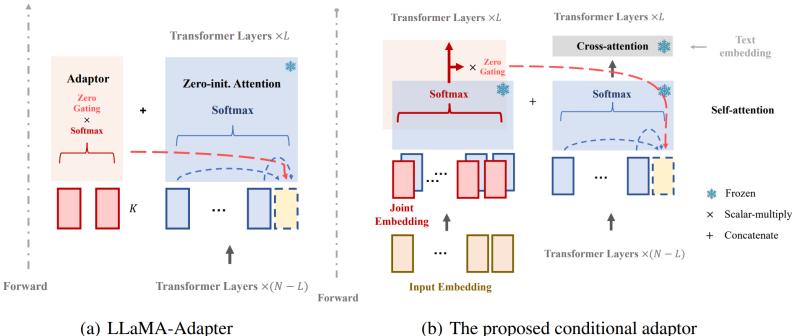


Transformer模型

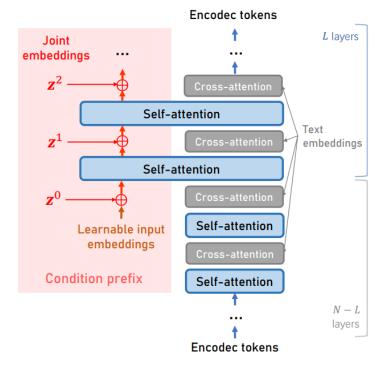


Coco-mulla模型

方法(2): 条件生成结构

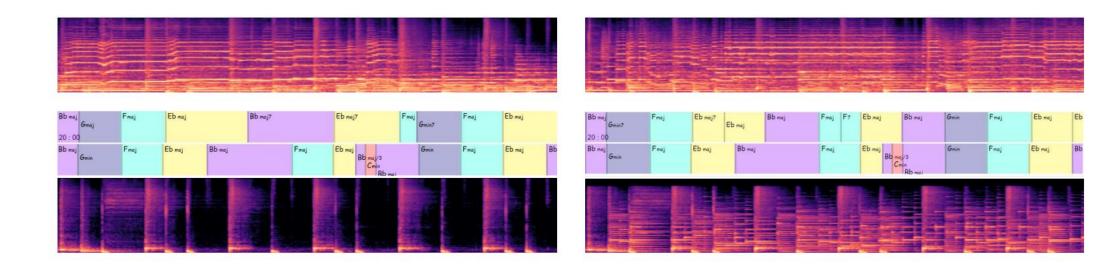


(b) The proposed conditional adaptor



实验评估

	$Chord_{\mathrm{rec}} \uparrow$	$\mathbf{Chord}^*_{\mathrm{rec}} \uparrow$	$\mathbf{Beat}_{F_1} \uparrow$	$CLAP_{scr} \uparrow$	$FAD_{ ext{vgg}}\downarrow$	$ extsf{FAD}^*_{ ext{vgg}}\downarrow$
Chord-only	0.412	0.195	-	0.401	6.209	6.695
MIDI-only	0.649	0.406	-	0.381	7.105	7.094
Drums-only	0.530	0.267	0.856	0.360	3.845	4.933
Full	0.791	0.524	0.864	0.351	3.697	4.370
MusicGen	-	-	-	0.441	6.434	6.847
Oracle	0.885	0.695	0.898	-	-	



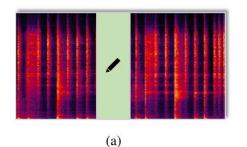
编配、内绘和润色:通过基于内容的控制来指导音乐生成和编辑

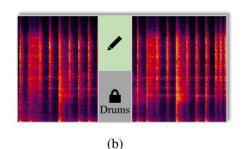
Arrange, Inpaint, and Refine: Steerable Long-term Music Audio Generation and Editing via Content-based Controls, IJCAI 2024 Liwei Lin, Gus Xia, Yixiao Zhang, Junyan Jiang

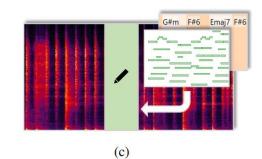
论文主要贡献

• 在Coco-mulla的基础上,新增: 音乐内绘、编配任务

• 为此改进了模型结构,实现了更好的控制

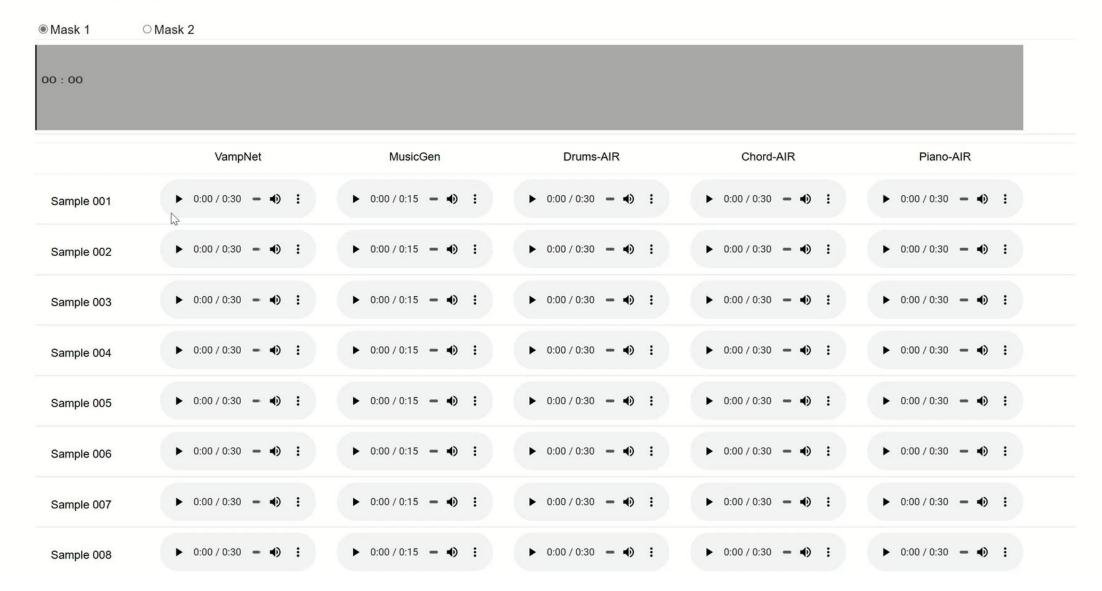






Inpianting

The gray blocks denote regions designated for inpainting. Note that musicgen operates as an autoregressive model, for which only continuation samples are placed here, rather than being subjected to inpainting.



方法

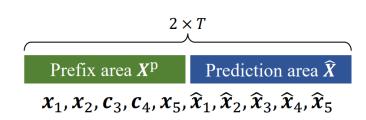
• 对于一个这样的内绘任务:

$$x_1, x_2, c_3, c_4, x_5 \rightarrow \widehat{x}_3, \widehat{x}_4$$

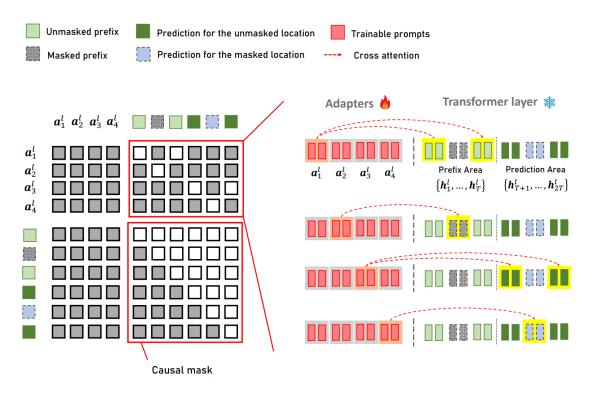
• MusicGen是自回归模型,不支持内绘

$$x_1, x_2$$
 可以输入, x_5 不可以

• 我们通过这样的方式提供信息:



方法



$$r(t) \begin{cases} 1, & \text{if } t \leq T \text{ and } t\text{-th frame is unmasked,} \\ 2, & \text{if } t \leq T \text{ and } t\text{-th frame is masked,} \\ 3, & \text{if } t > T \text{ and } (t-T)\text{-th frame is unmasked,} \\ 4, & \text{otherwise.} \end{cases}$$

输入的音乐表示:

$$m{H}^l = \{m{h}_1^l, m{h}_2^l, ..., m{h}_{2T}^l\}$$

Transformer的self-attention:

$$S^l = \text{Self-Attention}(H^l).$$

$$m{S}^l = \{m{s}_1^l, m{s}_2^l, ... m{s}_{2T}^l\}$$

进行cross-attention计算:

$$oldsymbol{u}_t^l = ext{Cross-Attention}(oldsymbol{h}_t^l, oldsymbol{a}_{r(t)}^l).$$

Adapter和Transformer信息融合:

$$\boldsymbol{s}_t^{l,*} = \boldsymbol{s}_t^l + g_{r(t)}^l \cdot \boldsymbol{u}_t^l,$$

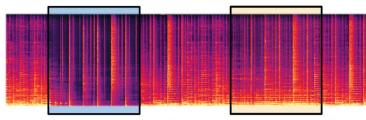
实验

(a) Slakh2100 Test Set

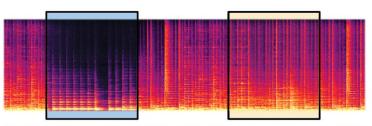
	CLA	$\mathbf{P}_{\mathrm{src}} \uparrow$	FAD	$_{ m vgg}\downarrow$
	Full	Prefix	Full	Prefix
Drum-AIR	0.749	0.756	1.423	1.422
Chord-AIR	0.753	0.757	1.220	1.222
Piano-AIR	0.755	0.761	1.290	1.282
MusicGen	0.656	0.687	1.251	1.218
VampNet	0.631	0.643	2.910	3.424

(b) RWC-POP-100

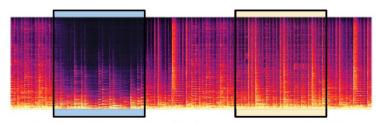
	$CLAP_{src} \uparrow$		FAD	$_{ m vgg}\downarrow$
	Full	Prefix	Full	Prefix
Drum-AIR	0.619	0.627	1.606	1.691
Chord-AIR	0.614	0.625	1.593	1.681
Piano-AIR	0.611	0.621	1.531	1.623
MusicGen	0.373	0.441	2.474	2.276
VampNet	0.613	0.618	3.689	3.910



(a) Drum track controls, where the condition is drum audio.



(b) Chord progression controls, where the condition is block chords audio.



(c) Arrangement and orchestration from piano cover, where the condition is piano cover audio.

指令微调的音乐大模型: 多任务音乐编辑

Instruct-MusicGen: Unlocking Text-to-Music Editing for Music Language Models via Instruction Tuning, arXiv/2405.18386

<u>Yixiao Zhang</u>, Yukara Ikemiya, Naoki Murata, Woosung Choi, Marco Martínez, Liwei Lin, Gus Xia, Wei-Hsiang Liao, Yuki Mitsufuji, Simon Dixon

论文主要贡献

• 在Coco-mulla的基础上,新增:添加、消除、提取乐器任务

• 通过指令微调, 让同一个模型能够适应多任务指令

• 同样只需要训练很短时间: 5000 steps微调

Demo

Adding a stem

Slakh

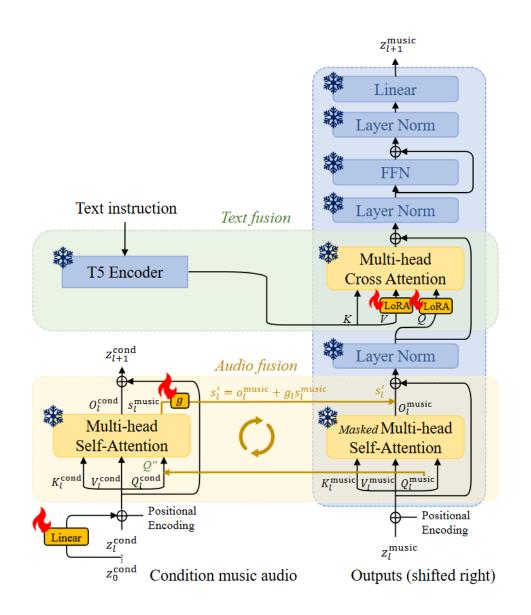
Instruction	Input audio	Output audio	Ground truth
add bass	▶ 0:00 / 0:00 ● •	► 0:00 / 0:00 (i)	► 0:00 / 0:00 (1)
add bass	► 0:00 / 0:00 () :	► 0:00 / 0:00 ———————————————————————————	▶ 0:00 / 0:00
add piano	▶ 0:00 / 0:00	► 0:00 / 0:00 ()	► 0:00 / 0:00 ()
add piano	▶ 0:00 / 0:00	► 0:00 / 0:00 (i)	► 0:00 / 0:00 (i)
add guitar	▶ 0:00 / 0:00	▶ 0:00 / 0:00	▶ 0:00 / 0:00

Extracting a stem

Slakh

Instruction	Input audio	Output audio	Ground truth
only drums	▶ 0:00 / 0:00	► 0:00 / 0:00 ———————————————————————————	▶ 0:00 / 0:00
only drums	▶ 0:00 / 0:00	► 0:00 / 0:00 ———————————————————————————	▶ 0:00 / 0:00
only bass	▶ 0:00 / 0:00	► 0:00 / 0:00 ———————————————————————————	▶ 0:00 / 0:00
only bass	▶ 0:00 / 0:00 4 0 1	▶ 0:00 / 0:00 () !	► 0:00 / 0:00

方法



输入的音乐表示:

$$\begin{split} Z^{\text{cond}} &= \{z_0^{\text{cond}}, z_1^{\text{cond}}, \dots, z_M^{\text{cond}}\}, \\ Z^{\text{music}} &= \{z_0^{\text{music}}, z_1^{\text{music}}, \dots, z_M^{\text{music}}\}, \end{split}$$

Transformer的self-attention:

$$\begin{split} Q_l^{\text{music}}, K_l^{\text{music}}, V_l^{\text{music}} &= \text{QKV-projector}(z_l^{\text{music}}), \\ o_l^{\text{music}} &= \text{SelfAttn}(Q_l^{\text{music}}, K_l^{\text{music}}, V_l^{\text{music}}). \end{split}$$

将条件音轨经过一层变换:

$$h = f_l(z^{\text{cond}}) + e_l$$
.

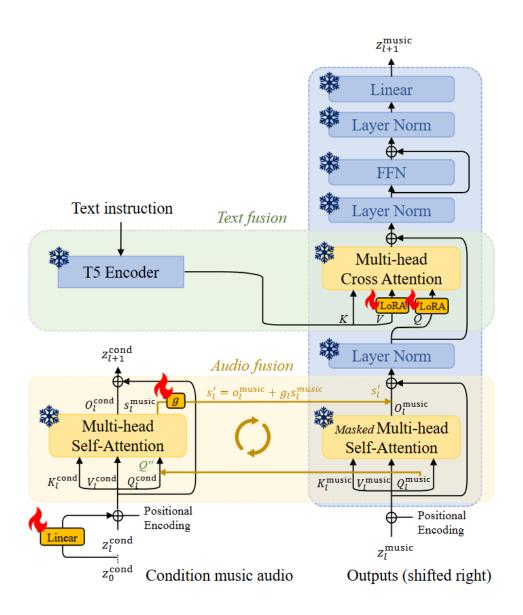
条件Transformer的self-attention:

$$\begin{split} Q_l^{\text{cond}}, K_l^{\text{cond}}, V_l^{\text{cond}} &= \text{QKV-projector}(z_l^{\text{cond}} + h), \\ z_{l+1}^{\text{cond}} &= \text{SelfAttn}(Q_l^{\text{cond}}, K_l^{\text{cond}}, V_l^{\text{cond}}). \end{split}$$

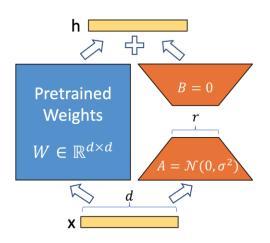
两个self-attention信息进行融合:

$$s_l^{\text{music}} = \text{CrossAttn}(Q_l^{\text{music}} + Q_l^{\text{cond}}, K_l^{\text{cond}}, V_l^{\text{cond}}).$$
$$s_l' = o_l^{\text{music}} + g_l \cdot s_l^{\text{music}},$$

方法



使用LoRA微调文本编码器的cross-attention:



得到最终的音乐表示:

$$z_{l+1}^{\text{music}} = \text{TextFusion}(s_l', X^{\text{instruct}}),$$

实验

Task	Models	FAD↓	CLAP↑	KL↓	SSIM↑	P-Demucs↑	SI-SDR↑	SI-SDRi↑
	AUDIT	6.88	0.12	1.02	0.21	0.53	-	-
	M ² UGen	7.24	0.22	0.99	0.20	0.43	-	-
Add	Ours	3.75	0.23	0.67	0.26	0.80	-	-
	AUDIT	15.48	0.07	2.75	0.35	0.33	-45.60	-47.28
Remove	M ² UGen	8.26	0.09	1.59	0.23	0.70	-44.20	-46.13
	Ours	3.35	0.12	0.66	0.45	0.76	-2.09	-3.77
	AUDIT	15.08	0.06	2.38	0.42	0.61	-52.90	-50.16
Extract	M ² UGen	8.14	0.11	2.15	0.31	0.60	-46.38	-43.53
	Ours	3.24	0.12	0.54	0.52	0.75	-9.00	-6.15

Table 2: Comparison of text-based music editing models on the Slakh dataset (4 stems).

Task	Models	FAD↓	CLAP↑	KL↓	SSIM↑	P-Demucs↑	SI-SDR↑	SI-SDRi↑
	AUDIT	4.06	0.12	0.84	0.21	0.50	-	-
Add	M ² UGen	5.00	0.18	0.83	0.20	0.45	-	-
	Ours	3.79	0.18	0.35	0.35	0.77	-	-
	AUDIT	10.72	0.10	2.46	0.34	0.41	-44.32	-57.10
Remove	M ² UGen	3.75	0.13	1.27	0.19	0.72	-43.94	-56.73
	Ours	5.05	0.10	0.84	0.34	0.78	-13.70	-26.48
	AUDIT	6.67	0.07	1.97	0.45	0.60	-54.53	-56.17
Extract	M ² UGen	5.74	0.08	1.91	0.25	0.52	-42.84	-44.49
	Ours	4.96	0.11	1.36	0.40	0.78	-21.39	-23.03

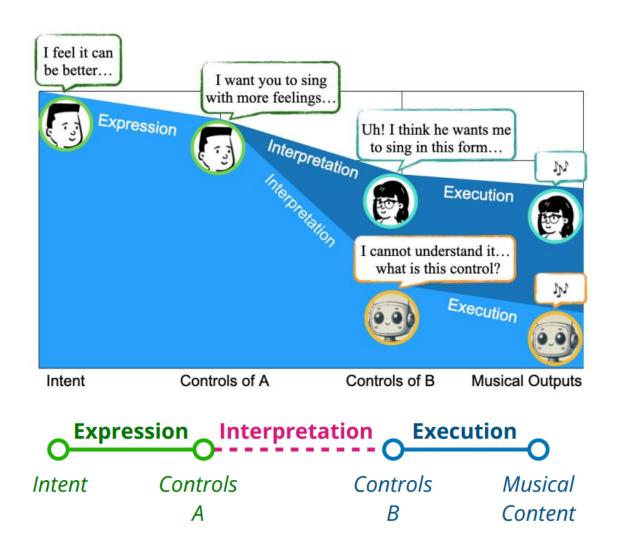
Table 3: Comparison of text-based music editing models on the MoisesDB dataset.

Model	Instruction Adherence↑	Audio Quality ↑
AUDIT	1.54	2.56
M^2UGen	1.70	1.92
Ours	3.85	3.55
Ground truth	4.36	4.21

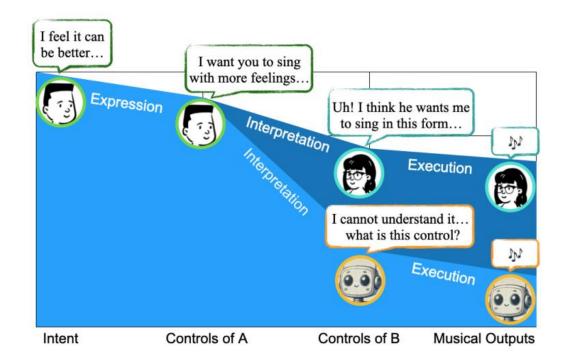
音乐大模型为什么仍然不能像人类一样合作?

The Interpretation Gap in Text-to-Music Generation Models, NLP4MusA Workshop @ ISMIR 2024 Yongyi Zang*, Yixiao Zhang*

音乐协作的三个阶段



音乐协作的三个阶段





- 人工智能模型与人类存在"解释差距"
- 音乐家在创作过程中会使用各种形式的控制信号,包括文字描述、旋律片段、和弦进程…
- 这些信号**可能非常具体**也可能带有**相当程度的 模糊性**
- 例如:
 - "让这个贝斯音轨听起来更加温暖"
 - "让这段旋律听起来更具有表现力"
- 包含很多上下文信息和隐含的意义,需要接收者有一定的音乐素养才能正确理解并执行

人类音乐协作的控制信号

Intent	Controls A Controls B	Outputs
teractions		
Light touch Spacious sound Spacious sound Spacious sound Modulate key Emphasizing a chord Guitarist Emphasizing a chord		 → Piano audio → Natural result → Unnatural result → Score → Clean strum sound → Muffled strum sound
Party Interactions		
er & Experienced Vocalist Emotive singing	→ "More feelings" → More dynamics & articulation	→ Emotional vocal track
er & Novice Vocalist Emotive singing	\rightarrow "More feelings" \rightarrow Sing closer to microphone	→ Unnatural vocal track
enced Rock Band Guitar solo	→ Gesture → Drums and bass play fill; vocalist stop singing	→ Solo section
Rock Band Guitar solo		→ Solo fights with vocal, creating cacophony
ctor & Orchestra Crescendo		→ Balanced crescendo
Crowd Build energy	\rightarrow Throwing hands up in \rightarrow Crowd thinks it's peak the air	→ Early climax

现在的文本到音乐生成模型不具备必要的理解能力

现有的文本到音乐生成控制一览

Model	Semantic controls	Precise controls
Integrated Controls in Foundation Models	7	
Mustango (Melechovsky et al., 2024)	Text description, metadata	-
MusicGen (Copet et al., 2024)	Text description	melody spectrogram
Diff-A-Riff (Nistal et al., 2024)	Text description	Music audio mixture
Jen-1 Composer (Yao et al., 2023)	Text description	Other instrument tracks
GMSDI (Postolache et al., 2024)	Instrument name	Other instrument tracks
Control Enhancement Modules		
Coco-mulla (Lin et al., 2023)	Text description	Drum track, chord, melody
AIRGen (Lin et al., 2024)	Text description	Drum track, chord, melody
JASCO (Tal et al., 2024)	Text description	Drum track, chord, melody
Music ControlNet (Wu et al., 2024)	Text description	Dynamic, melody, rhythm
Jen-1 DreamStyler (Chen et al., 2024)	Text description	Reference music audio
Music Editing Methods		
MusicMagus (Zhang et al., 2024b)	Text swapping	Music audio mixture
InstructME (Han et al., 2023)	Edit instruction	Music audio mixture
Instruct-MusicGen (Zhang et al., 2024a)	Edit instruction	Music audio mixture
Loop Copilot (Zhang et al., 2023)	Edit instruction	Conversational context (music audio, text)
M ² UGen (Hussain et al., 2023)	Edit instruction	Conversational context (music audio, text)
ChatMusician (Yuan et al., 2024a)	Edit instruction	Conversational context (symbolic music, text)

现在的文本到音乐生成模型不具备必要的理解能力

可能的解决办法

- 收集解释数据(文本、视频)进行训练; 或
- •利用大语言模型(LLM)的先验知识。

• 这仍然是一个亟待解决的开放问题



声学大讲堂 音频产业创新技术公益讲座

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