

CyberHost: A One-stage Diffusion Framework for Audio-driven Talking Body Generation

Gaojie Lin*¹, Jianwen Jiang*¹ (equal contributions)
Chao Liang¹, Tianyun Zhong², Jiaqi Yang¹, Zerong Zheng¹, Yanbo Zheng¹

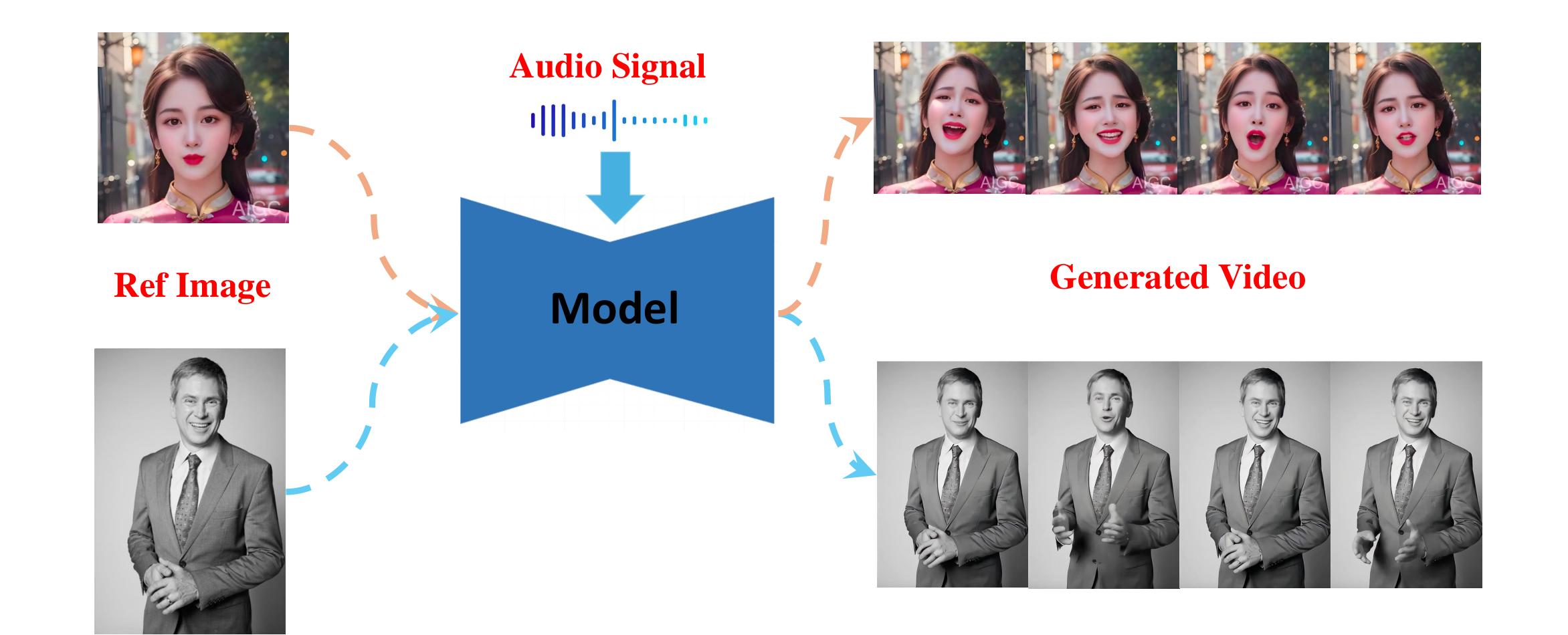
¹ByteDance ²ZheJiang University





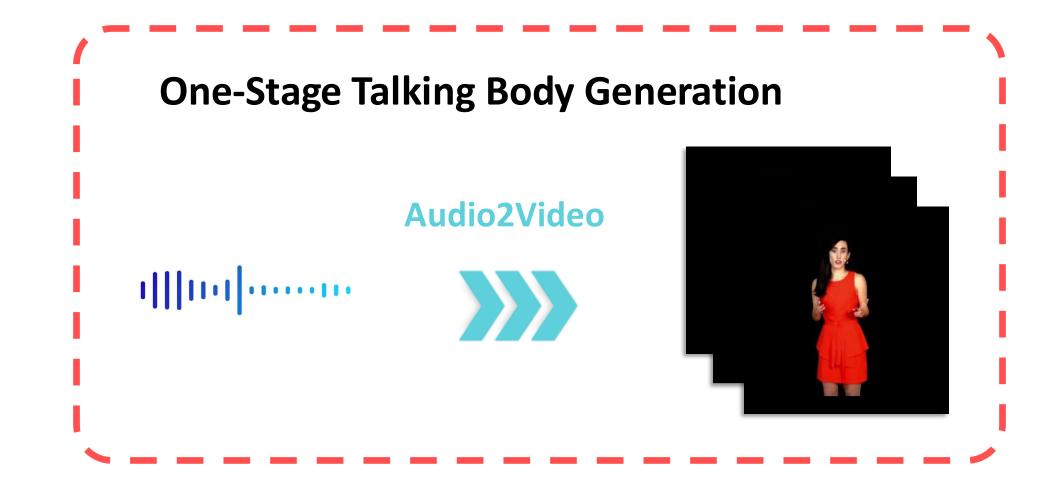
Problem Definition

Audio-driven Human Animation



Background & Motivation

- ➤ Most previous work focused on Talking Head Generation.
- ➤ Two-stage Talking Body Generation methods suffer from:
 - > Increased system complexity and reduced learning efficiency.
 - > Limited by the capability of intermediate representations.
 - > Effectiveness relies on pose/mesh annotation accuracy.

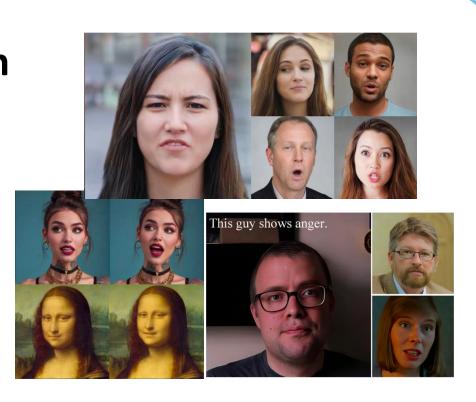


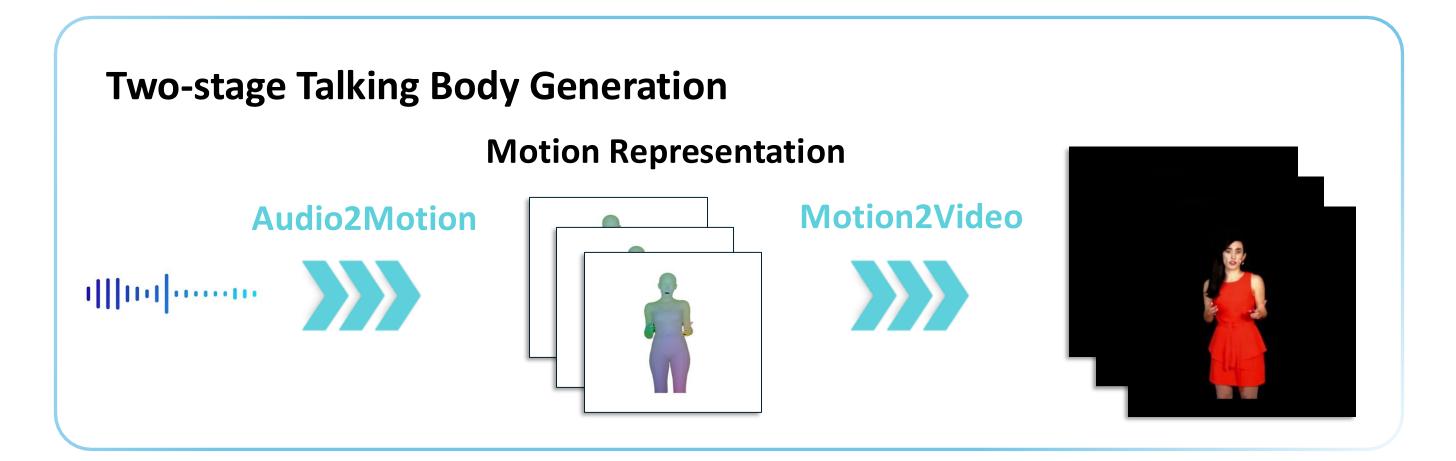


We aim to explore the upper limits of one-stage framework for Talking Body Generation.

Talking Head Generation

- VASA, NIPS 2024
- EMO, ECCV 2024
- Hallo2, ICLR 2025
- EchoMimic, AAAI 2025
- SadTalker, CVPR 2024
-





Comparsion with Two-Stage Method

Speech2Gesture



MoGlow



SDT



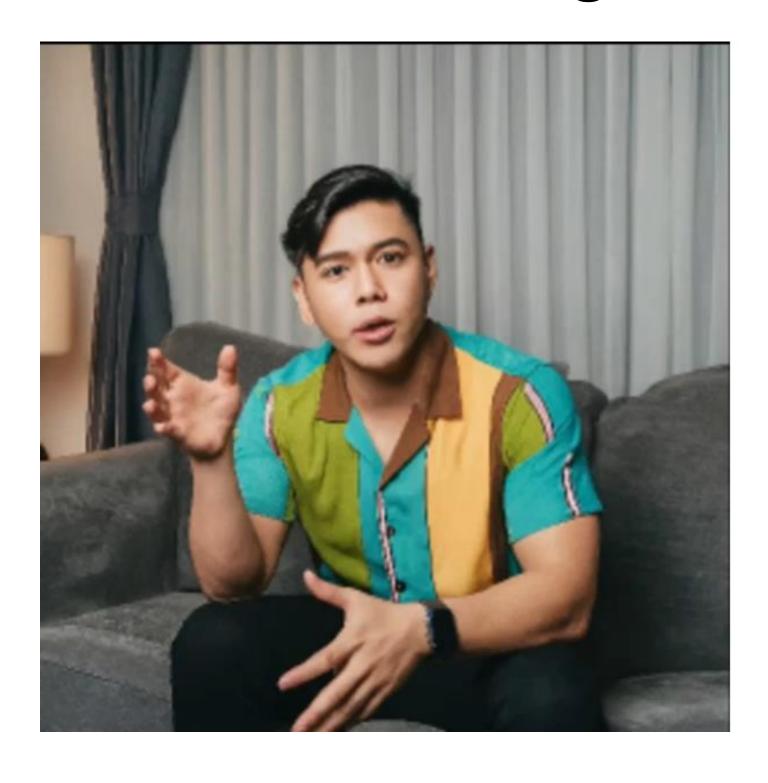
CyberHost *



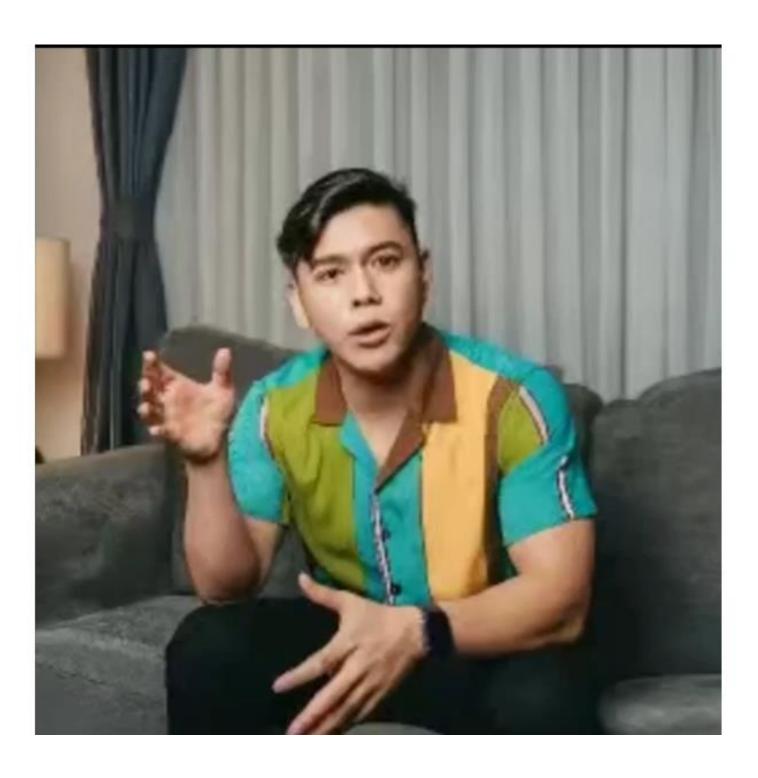
Speech2Gesture: Learning Individual Styles of Conversational Gestures, CVPR 2019 MoGlow: Style-controllable speech-driven gesture synthesis using normalising flows, TOG 2020 SDT, Speech Drives Templates: Co-Speech Gesture Synthesis with Learned Templates, ICCV 2021

Comparsion with Two-Stage Method

Reference Image



VLOGGER



CyberHost



VLOGGER: Multimodal Diffusion for Embodied Avatar Synthesis, Arxiv 2024

Zero-shot Testing

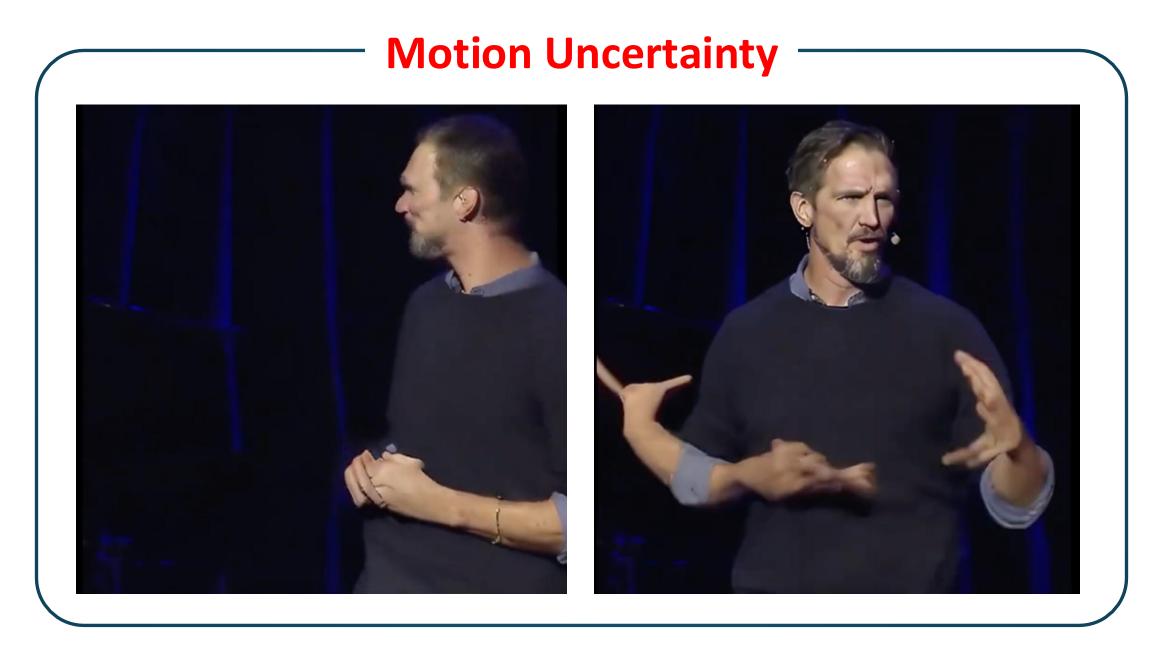




Challenge

- > One-stage Talking Body Generation faces two challanges:
 - > Details Underfitting: Local structural priors missing & Small critical region coverage.
 - > Motion Uncertainty: Higher motion freedom & Weak audio-limb correlation.

Details Underfitting | Image | Figure | Figure

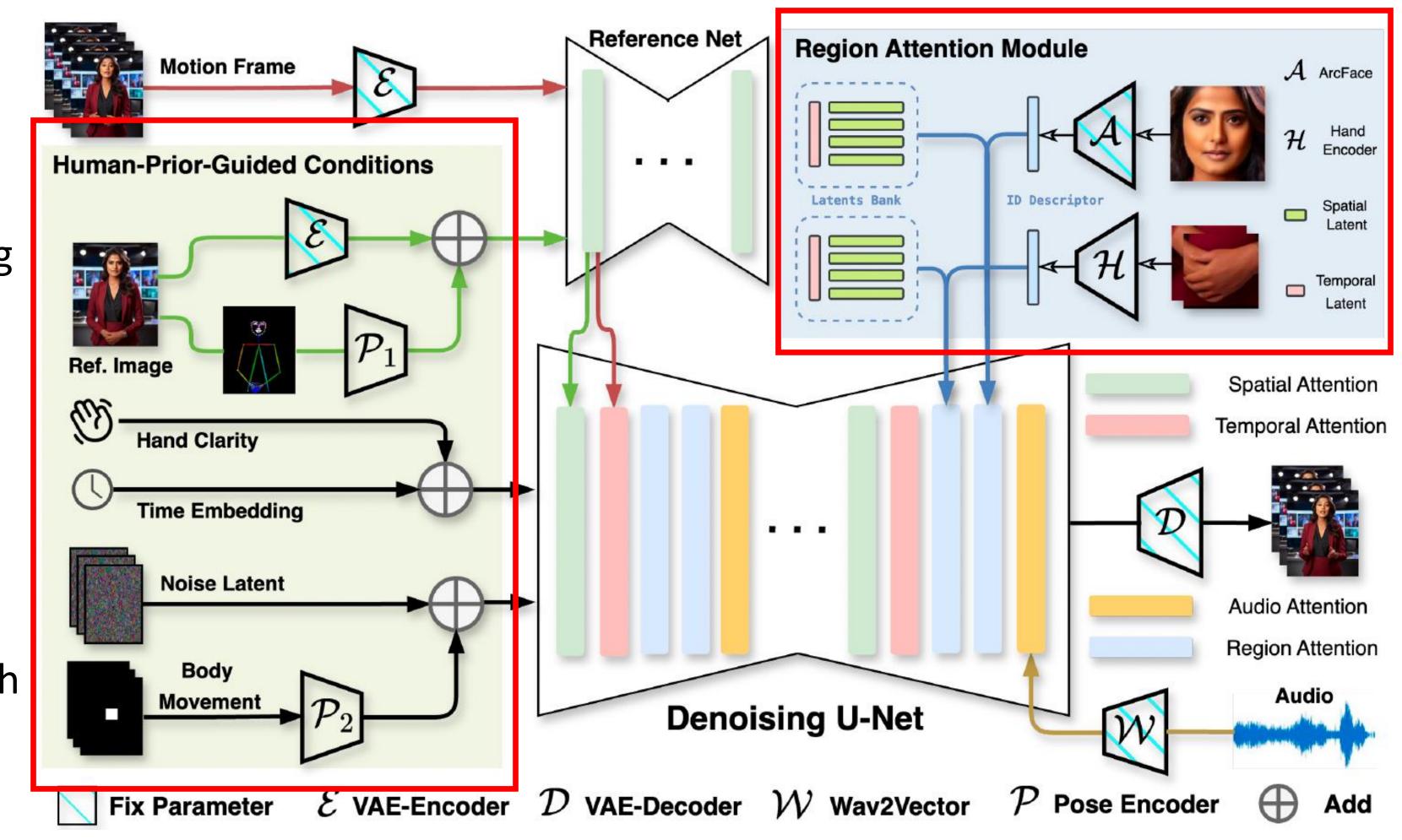


Overall Framework

To tackle Details Underfitting, we desigin Region Attention Module

(RAM) to enhance the model's fitting ability for critical local regions.

To tackle Motion Uncertaincy, we introduce Human-Prior-Guided
 Conditions to provide the model with prior knowledge of motion patterns and human structure.

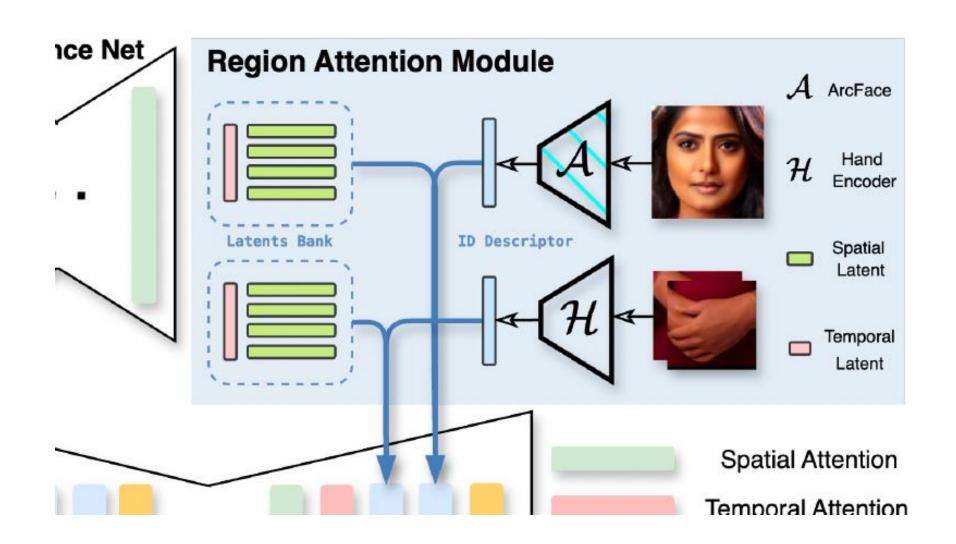


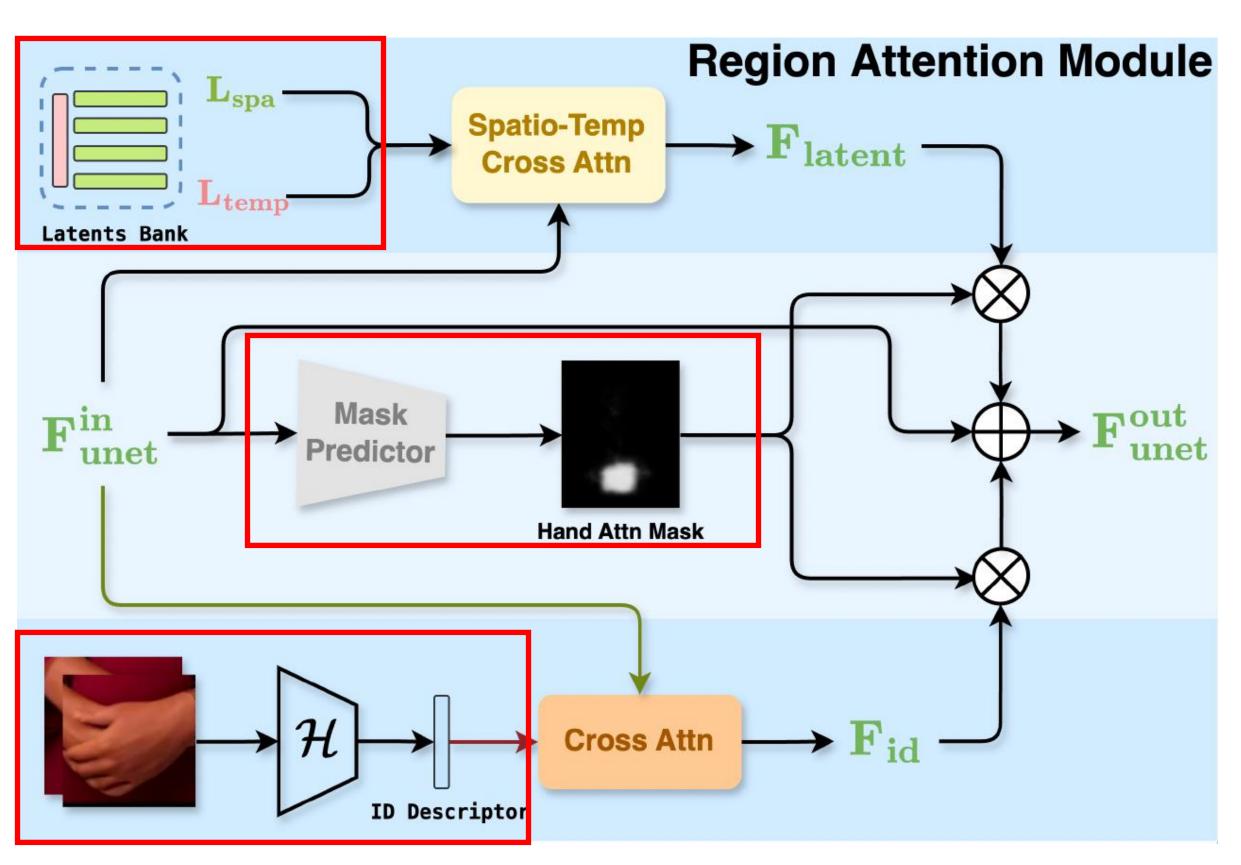
Region Attention Module

➤ Spatio-temporal Latents Bank: Additional learnable parameters, to learn shared priors of local details like structural topology and motion patterns.

➤ **Identity Descriptor:** Appearance features extracted from cropped images, to improve ID consistency.

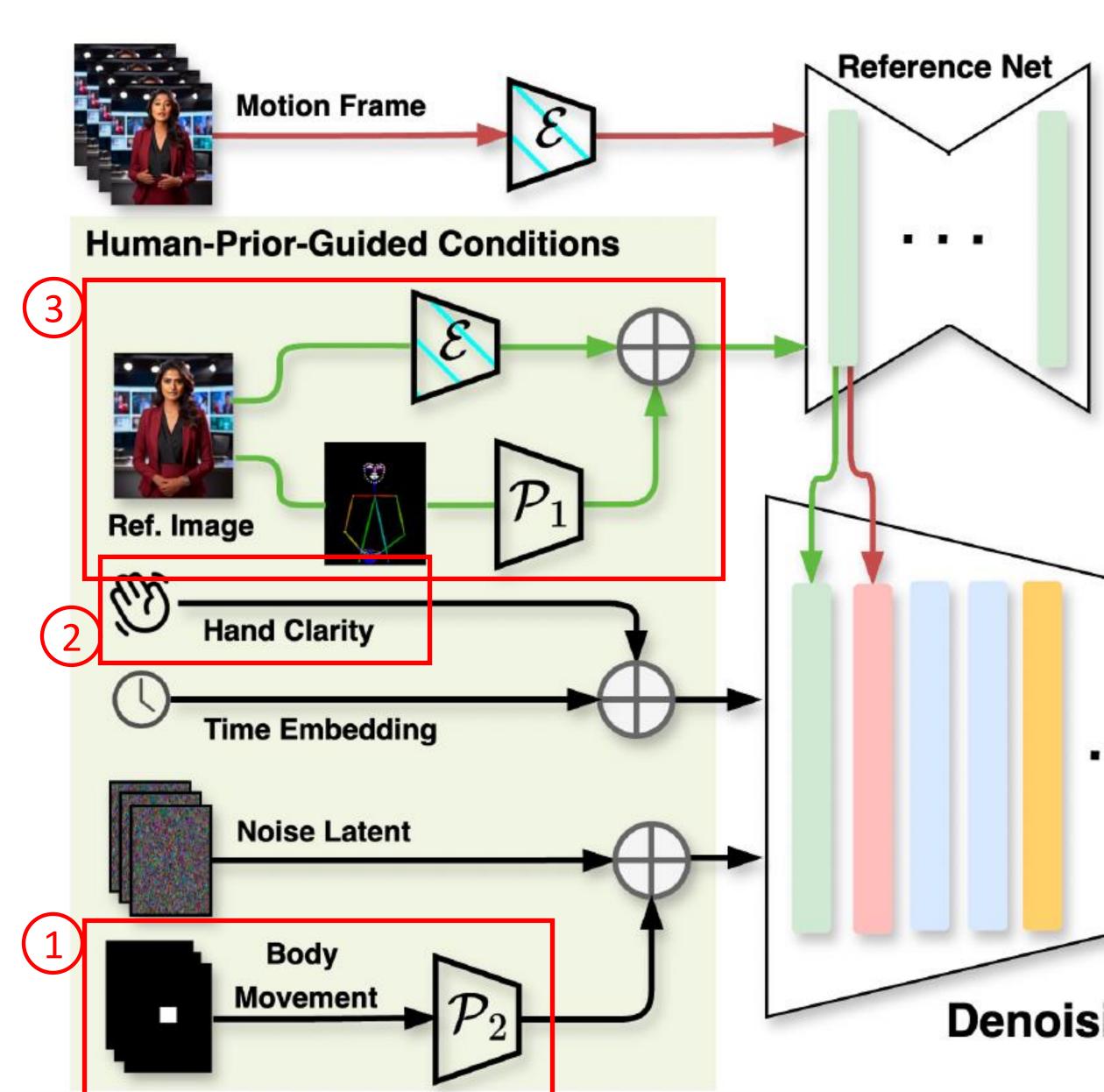
Regional Mask Predictor: Guide the learning process to focus on targeted local regions





Human-Prior-Guided Condition

- 1 Body Movement Map: A control signal for the movement amplitude of the body root
- 2 Hand Clarity Score: Indicate the clarity of hand regions in the training video frames
- Ose-aligned Reference Feature: Encode skeleton map along with reference image to incorporate its topological structure information.



Experiment Results

Hand Keypoint Confidence

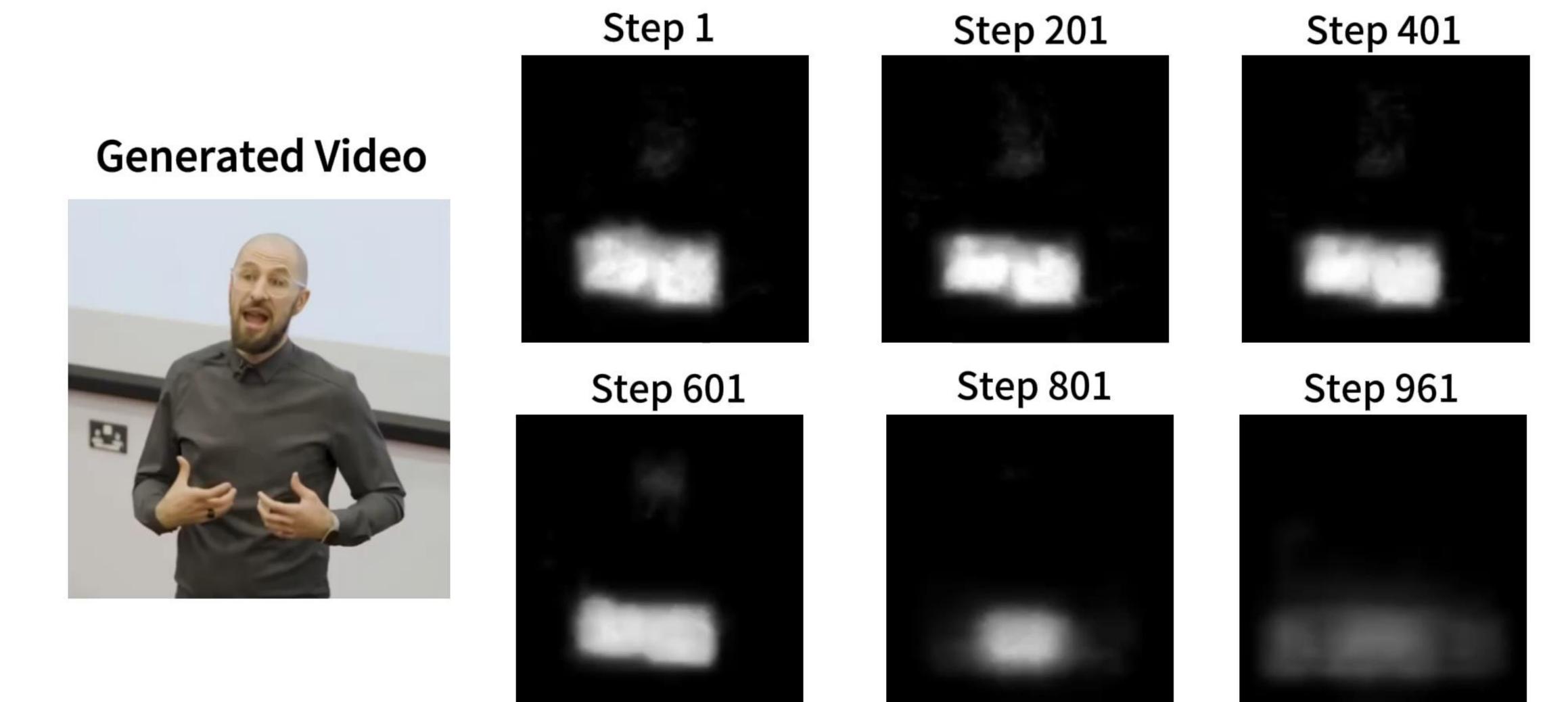
Hand Keypoint Variance

Table 1: Quantitative comparison of audio-driven talking body. * denotes evaluate on vlogger test set.

Methods	SSIM↑	PSNR ↑	FID↓	FVD↓	CSIM↑	SyncC [↑]	SyncD↓	HKC↑	HKV↑
DiffTED	0.667	15.48	95.45	1185.8	0.185	0.925	12.543	0.769	-
DiffGest.+MimicMo.	0.656	14.97	58.95	1515.9	0.377	0.496	13.427	0.833	23.40
CyberHost (A2V-B)	0.691	16.96	32.97	555.8	0.514	6.627	7.506	0.884	24.73
Vlogger *	_	_	-	_	0.470	0.601	11.132	0.923	9.84
CyberHost (A2V-B) *	_	-	-	-	0.522	7.897	7.532	0.907	18.75

- > Superior visual quality (SSIM, PSNR, FID, FVD)
- > Enhanced ID consistency (CSIM)
- > Improved lip-sync accuracy (SyncC, SyncD)
- > High hand generation quality (HKC) and richer hand gesture diversity (HKV)

Predicted Hand Mask at different timesteps within the Region Attention Module



Visual ablation studies on the Latents Bank and ID Descriptor:

> w/o Latents Bank: decrease in the local structural stability.

> w/o ID Descriptor: decrease in ID consistency.

Ref. Image



w/o Latents Bank



w/o ID Descriptor



CyberHost (Full)



Hand Clarity Score: sacrifice some hand movement richness, greatly reduces hand blur occurrences.

Medium High Low

Pose-Align Referece Feature: Preventing the generation of results with evident limb ambiguities.

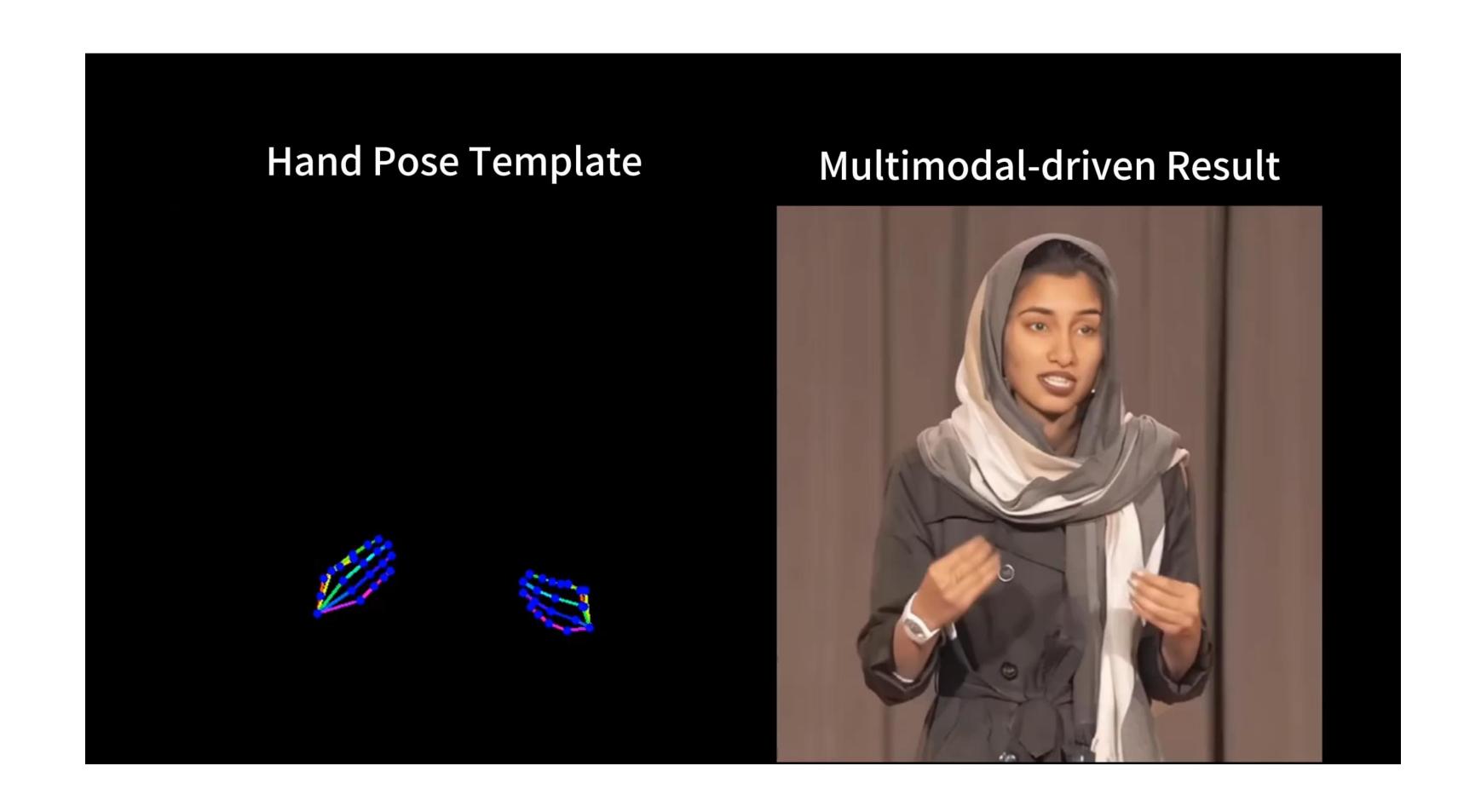
Ref. Image





Multimodal-driven Results

Combining hand pose templates with audio signals to support Multimodal-driven setting.

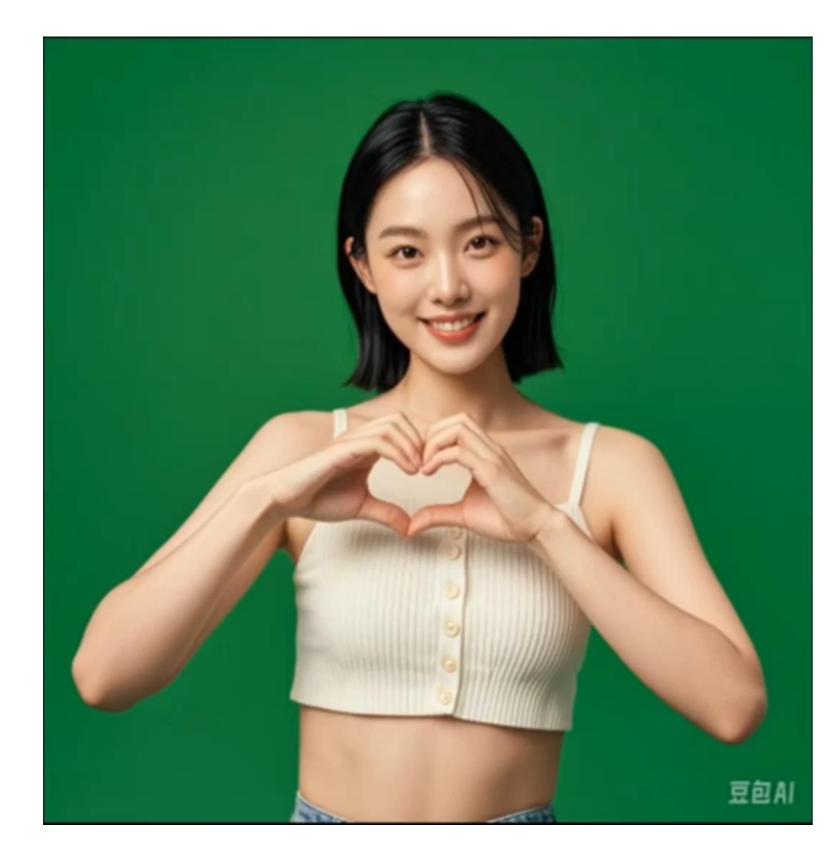


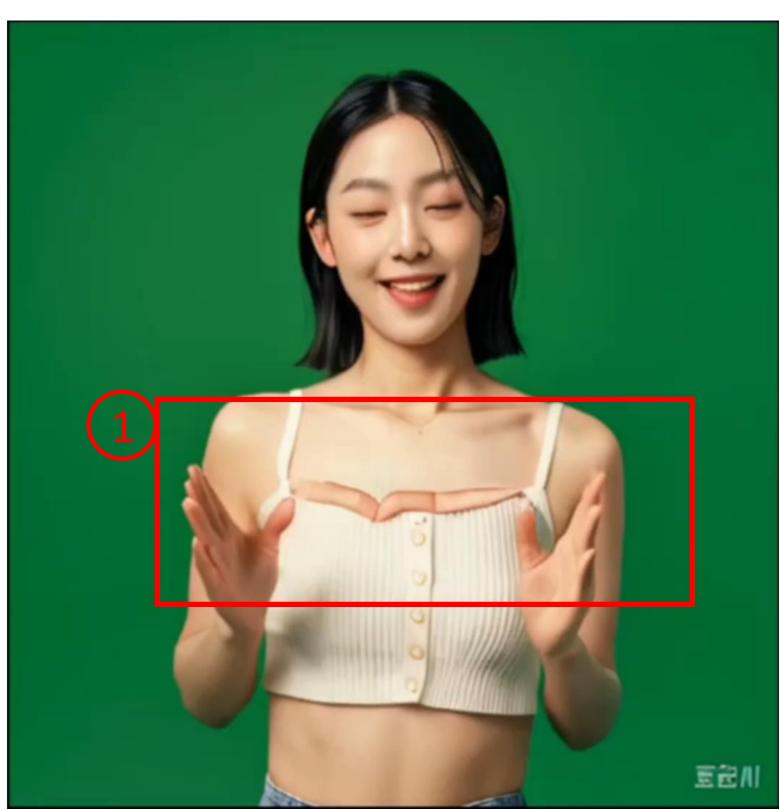
Failure Cases & Limitations

> Challenging reference images

> Exaggerated limb proportions

> Challenging audio signals





Failure Cases & Limitations

Challenging reference images

> Exaggerated limb proportions

> Challenging audio signals



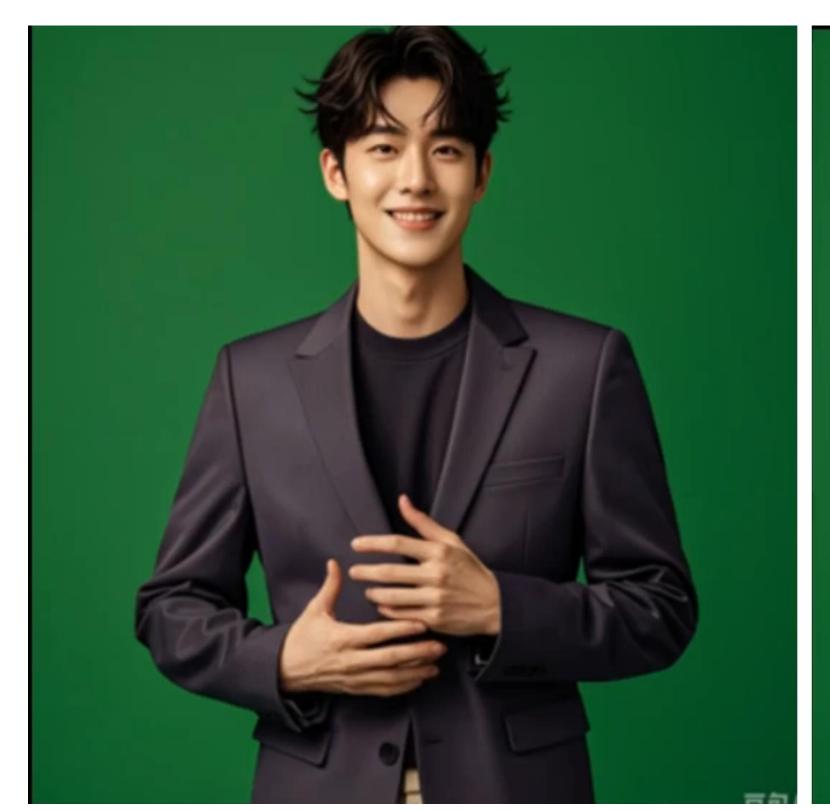


Failure Cases & Limitations

Challenging reference images

> Exaggerated limb proportions

> Challenging audio signals





Conclusion

- The first One-stage audio-driven talking body framework without relying any intermediate representations.
- A Regional Attention Module to address the issue of *Details Underfitting* and enhance the generation quality of critical local regions.
- A suite of Human-Prior-Guided Conditions to mitigate the *Motion Uncertainty* in audio-driven settings.
- > Comprehensive visualization results to validate its effectiveness and superiority.



THANK YOU

Poster Location: Hall 3 + Hall 2B #71



Paper



Proj Page