

# Advanced Sign Language Video Generation with Compressed and Quantized Multi-Condition Tokenization

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## Introduction

**Fig. (1) Sign Language Video Generation (SLVG):**

generate **sign language video** from **signer image** and **spoken language text**.

**Fig. (2a) Naïve Solution #1:**

Skeleton (single coarse condition) to bridge Translator and Generator. (**poor performance, Fig. (3)**)

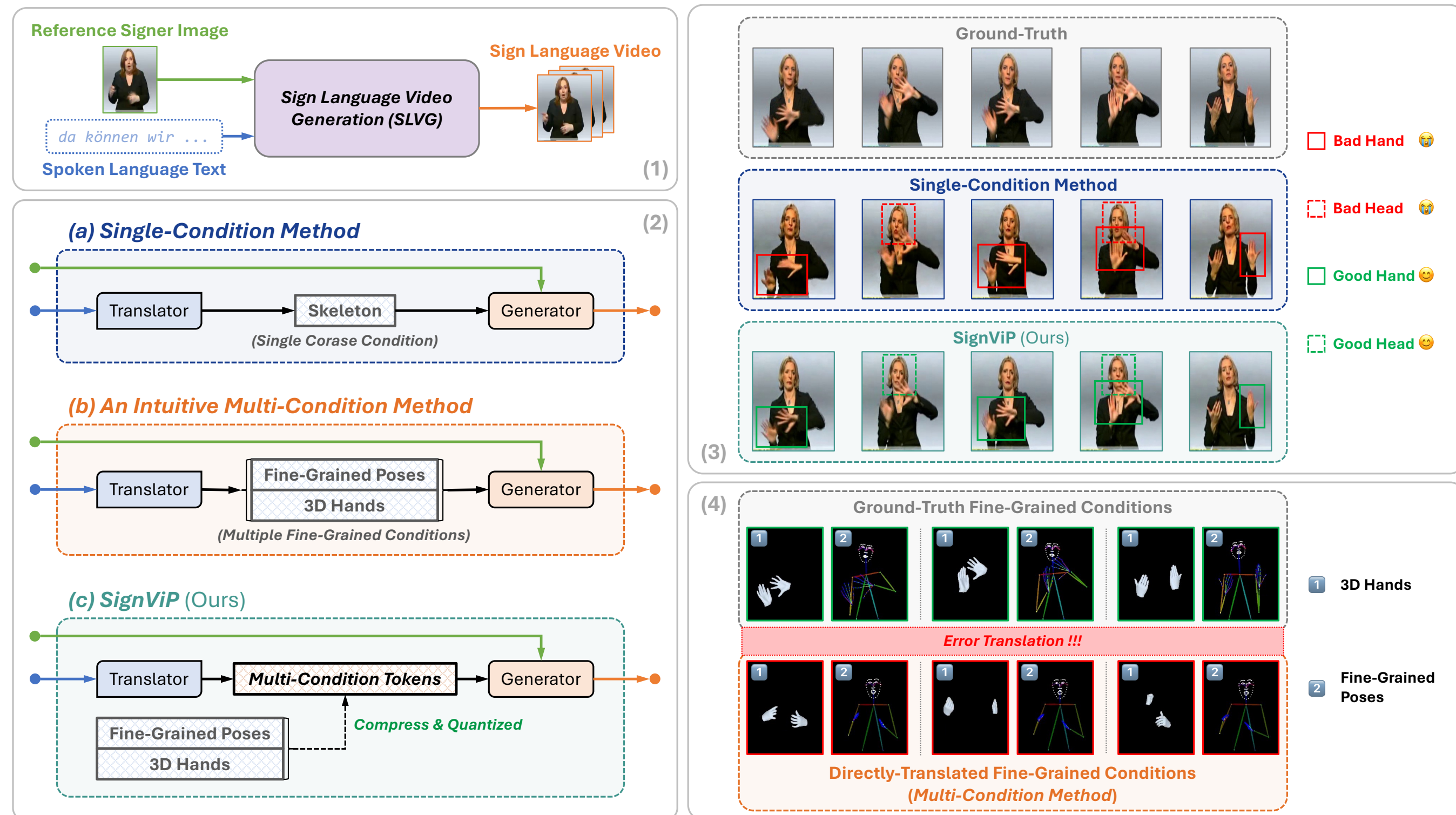
**Fig. (2b) Naïve Solution #2:**

Skeleton  $\rightarrow$  Multiple Fine-Grained Conditions. (**difficult to translate, Fig. (4)**)

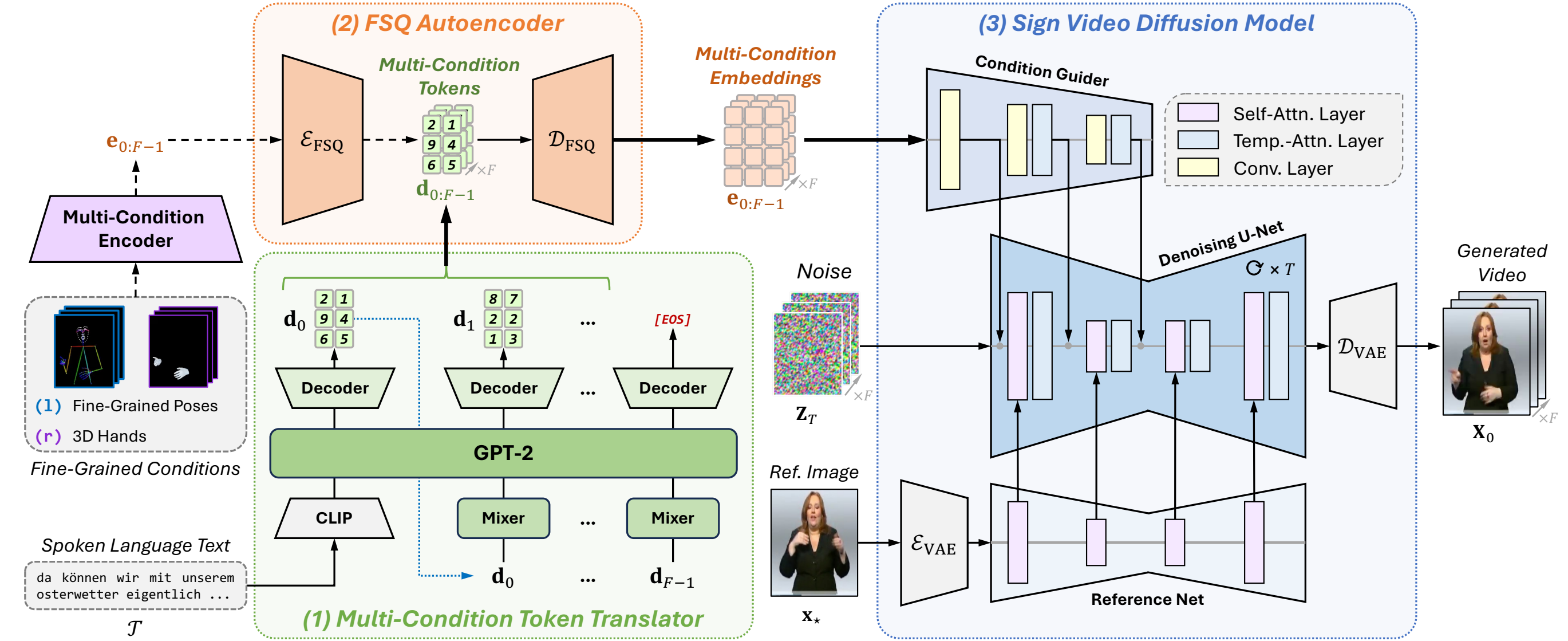
How to overcome the challenges in **the translation of multiple fine-grained conditions?**

**SignViP** adopts **a discrete tokenization paradigm** to integrate and represent **multiple fine-grained conditions**.

(Fig.2(c))



## Method



the construction of **a discrete multi-condition token space**

$$p_{\theta}(\mathbf{X}|\mathbf{x}_{*}, \mathcal{T}) = p_{\theta_{\text{gen}}}(\mathbf{X}|\mathbf{x}_{*}, \mathbf{e}_{0:F-1}) \cdot p_{\theta_{\text{AE}}}(\mathbf{e}_{0:F-1}|\mathbf{d}_{0:F-1}) \cdot p_{\theta_{\text{tran}}}(\mathbf{d}_{0:F-1}|\mathcal{T})$$

Signer Image      Continuous Multi-Condition Embeddings      Discrete Multi-Condition Tokens

## Experiments

Table 1: Comparison of **video back-translation** performance.

	RWTH-2014T					How2Sign				
	BLEU-1	BLEU-2	BLEU-3	BLEU-4	ROUGE	COMET	BLEU-1	BLEU-2	BLEU-3	BLEU-4
Ground-Truth	33.06	20.81	15.00	11.90	34.27	0.6157	20.37	13.11	9.78	7.53
MoMP [59] + ControlNet [88]	19.12	8.95	5.33	3.61	21.54	0.5033	12.53	5.59	3.48	2.31
MoMP [59] + AnimateAnyone [29]	20.05	8.79	5.24	3.72	21.68	0.5091	13.65	5.82	3.39	2.25
SignGAN [61]	17.41	7.93	4.67	3.16	19.64	0.4977	10.66	4.62	2.92	1.97
w/ AnimateAnyone [29]	18.29	7.75	4.59	3.23	19.70	0.4928	11.82	4.85	2.83	1.92
SignGen [47]	13.28	3.05	1.13	0.51	16.13	0.4086	8.21	1.91	0.64	0.41
SignViP (Ours)	26.72	15.65	11.14	8.65	28.85	0.5608	16.21	9.36	6.28	5.04

Table 2: Comparison of **pose back-translation** performance.

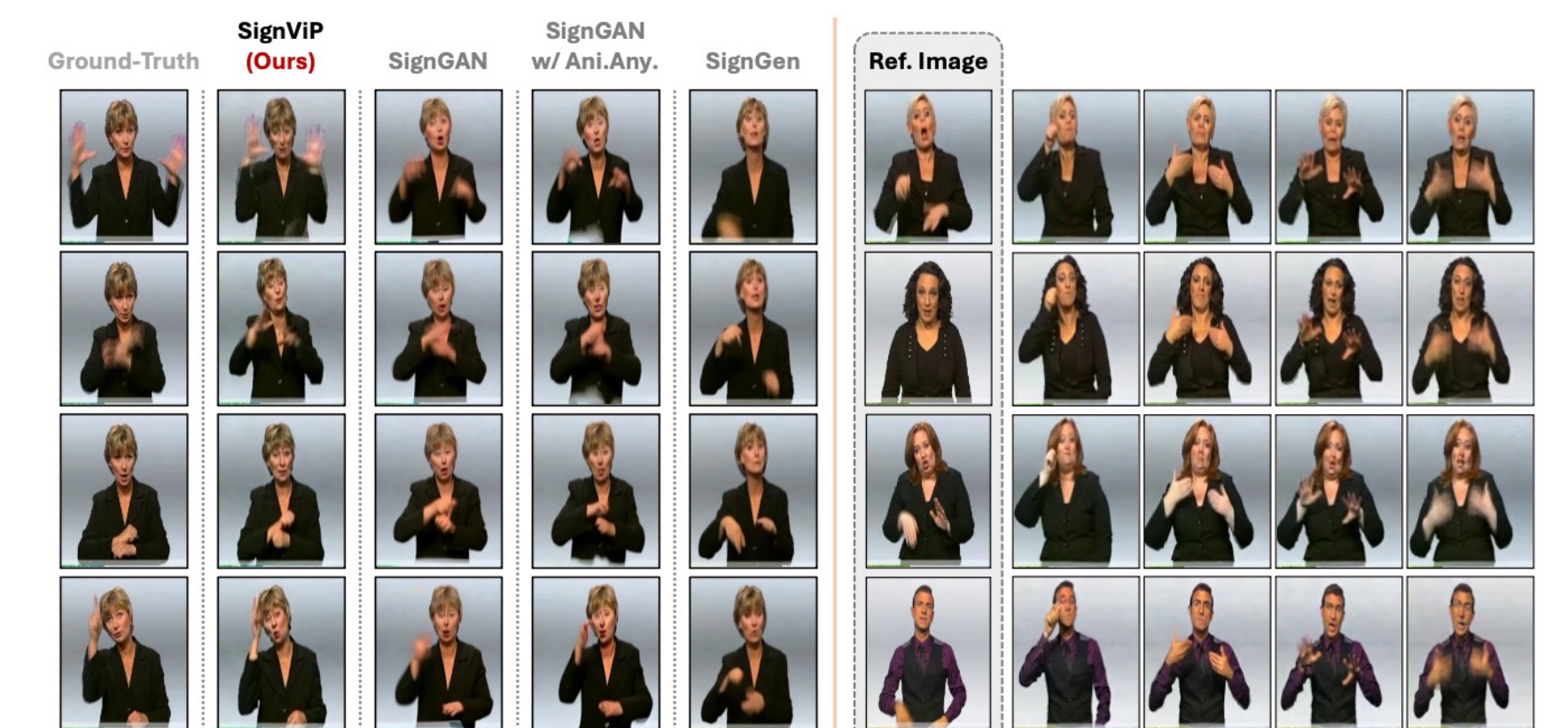
	RWTH-2014T					How2Sign				
	BLEU-1	BLEU-2	BLEU-3	BLEU-4	ROUGE	COMET	BLEU-1	BLEU-2	BLEU-3	BLEU-4
Ground-Truth	30.99	18.36	12.83	9.87	31.02	0.5978	24.56	14.96	10.31	7.91
ProTran [58]	17.96	8.99	5.64	4.07	20.97	0.5091	14.57	7.47	4.59	3.42
Adversarial [56]	17.70	8.96	5.72	4.18	21.15	0.5127	14.76	7.15	4.66	3.48
MDN [60]	18.06	9.30	6.06	4.52	21.44	0.5251	14.94	7.54	5.10	3.67
MoMP [59]	20.55	10.98	7.02	5.14	23.75	0.5466	16.57	8.47	5.38	4.16
SignGAN [61]	12.14	6.10	3.88	2.85	14.79	0.5123	10.86	5.27	3.30	2.62
w/ AnimateAnyone [29]	12.36	6.23	4.01	2.97	14.93	0.5231	10.97	5.36	3.39	2.67
SignGen [47]	10.42	2.42	0.89	0.38	12.68	0.4324	8.67	1.79	2.41	1.36
SignViP (Ours)	21.94	10.06	6.32	4.61	22.67	0.5347	17.35	8.28	5.41	4.42

Table 3: Comparison of **video quality**.

	RWTH-2014T				How2Sign			
	FID ↓	CLIP-FID ↓	FVD ↓	IDS ↑	FID ↓	CLIP-FID ↓	FVD ↓	IDS ↑
SignGAN [61]	547.90	167.70	1431.38	0.463	667.44	210.11	2766.97	0.538
w/ AnimateAnyone [29]	595.99	161.97	1330.54	0.462	679.41	215.05	2484.39	0.533
SignGen [47]	644.06	184.66	1715.32	0.515	815.69	186.32	3538.49	0.539
SignViP (Ours)	508.91	154.10	1025.45	0.571	575.67	109.61	2207.67	0.624

Table 4: Generative capability comparison of video diffusion models.

	RWTH-2014T					How2Sign				
	FVD ↓	SSIM ↑	PSNR ↑	LPIPS ↓	Hand SSIM ↑	FVD ↓	SSIM ↑	PSNR ↑	LPIPS ↓	Hand SSIM ↑
ControlNet [88]	556.63	0.784	19.50	0.137	0.483	427.22	0.826	21.32	0.116	0.657
AnimateAnyone [29]	365.42	0.794	20.06	0.121	0.505	293.18	0.821	21.54	0.103	0.663
Sign Video Diffusion Model (Ours)	275.22	0.829	22.91	0.089	0.614	210.63	0.855	23.11	0.074	0.752



Qualitative comparison

Identity Generalization