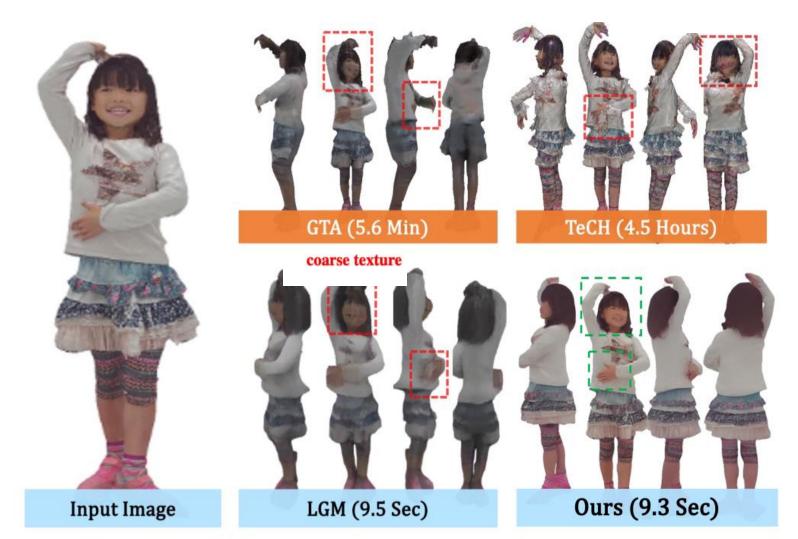


Panwang Pan<sup>\*1</sup>, Zhuo Su<sup>\*†1</sup>, Chenguo Lin<sup>\*1,2</sup>, Zhen Fan<sup>1</sup>, Yongjie Zhang<sup>1</sup>, Zeming Li<sup>1</sup>, Tingting Shen<sup>3</sup>, Yadong Mu<sup>2</sup>, Yebin Liu<sup>‡4</sup>





**The Drawbacks of Related Works:** 

- **\***Lack human priors as inductive biases.
- **\*** The requirements for dense input images.
- **\*** Time-consuming per-instance optimization.

## Contribution

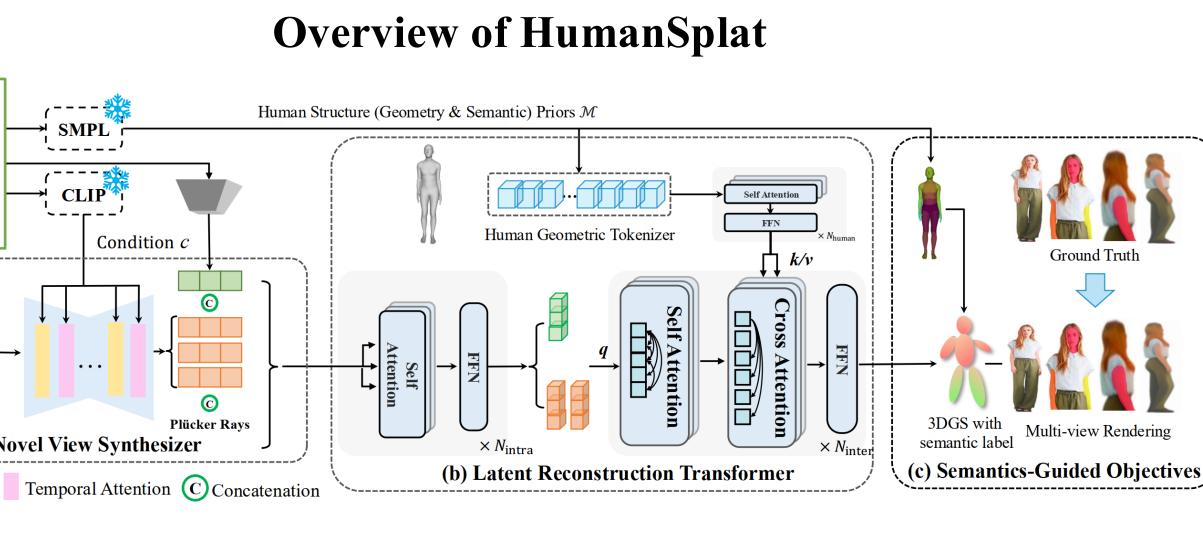
- ✤ It is the first work to propose a novel generalizable human Gaussian Splatting network for high-fidelity human reconstruction from a single image.
- **We integrate a universal Transformer framework by** leveraging human geometry priors and appearance priors from the 2D generative diffusion model.
- **\*** We enhance the fidelity of reconstructed human models by introducing semantic cues, hierarchical supervision, and tailored loss functions.
- **\*** Extensive experiments demonstrate that our method achieves SOTA performance.

(Prompt)						
		(a) No				
Sp	atial Atte	ention				
	Input Image (Prompt)					

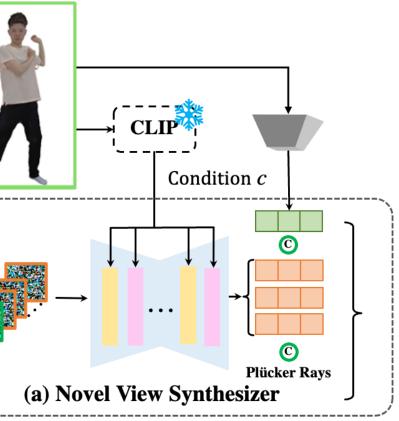
# HumanSplat: Generalizable Single-Image Human **Gaussian Splatting with Structure Priors**

<sup>1</sup>ByteDance, <sup>2</sup>Peking University, <sup>3</sup>Xiamen University, <sup>4</sup>Tsinghua University

## Methodology

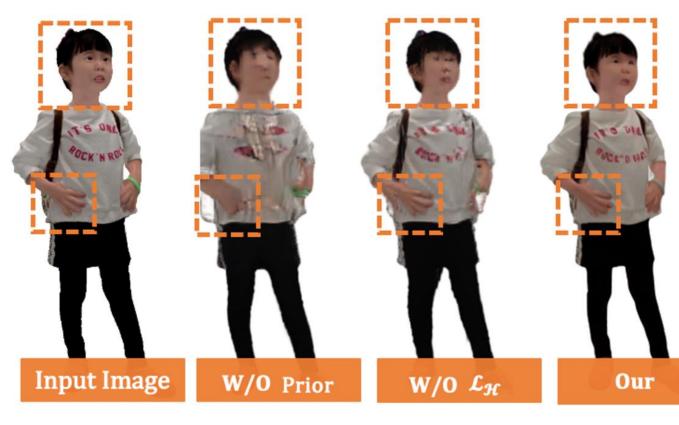


### **Novel-view Synthesizer**



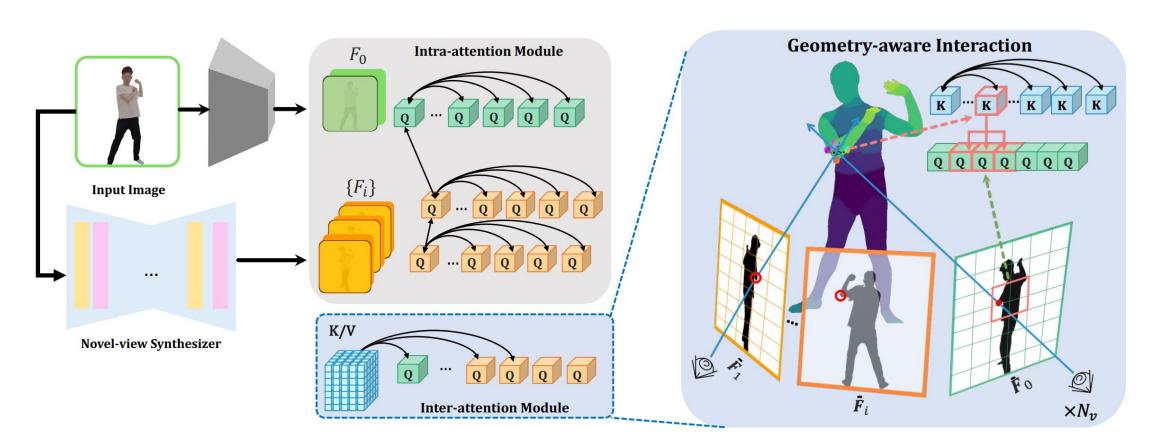
 $\mathbb{E}_{\epsilon \sim p(\epsilon)} \left[ \lambda(\epsilon) \right] \left[ \left( D_{\theta} \{ F_i^{\epsilon} \}_{i=1}^N; \{ e_i, a_i \}_{i=1}^N, c, F_0, \epsilon \right) - \{ F_i \}_{i=1}^N \right]$ 

### **Semantics-guided Objectives**

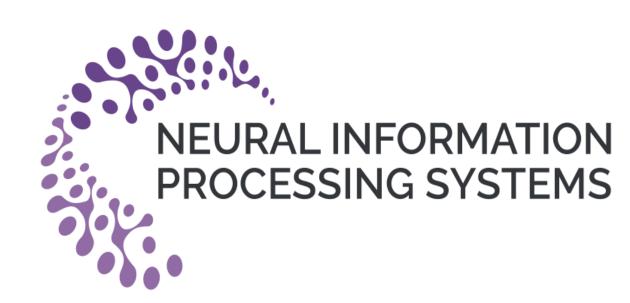


# $\mathcal{L}_{H} = \frac{1}{n} \frac{1}{m} \sum_{i=1}^{m} \sum_{j=1}^{m} \lambda_{i} \lambda_{j} (\mathcal{L}_{\text{MSE}}(\mathbf{I}_{i,j}, \hat{\mathbf{I}}_{i,j}) + \lambda_{p} \mathcal{L}_{p}(\mathbf{I}_{i,j}, \hat{\mathbf{I}}_{i,j})),$

### **Latent Reconstruction Transformer**









## **Experiments**

### **Quantitative Comparison** •

Quantitative comparison against SOTA methods.

Method	Category		THuman2.0 [105]			Twindom [106]			Time
	Diffusion	Human Prior	<b>PSNR</b> ↑	<b>SSIM</b> ↑	LPIPS↓	<b>PSNR</b> ↑	<b>SSIM</b> ↑	LPIPS↓	Inne
PIFu [107]	×	×	18.093	0.911	0.137	-	-	-	30.0s
LGM* [29]	$\checkmark$	×	20.013	0.893	0.116	19.840	0.851	0.292	<u>9.5s</u>
GTA [30]	×	$\checkmark$	18.050	-	-	17.669	0.741	0.418	43s
SIFU [102]	×	$\checkmark$	22.025	0.921	0.084	19.714	0.832	0.312	44s
SIFU <sup>†</sup> [102]	$\checkmark$	$\checkmark$	22.102	0.923	0.079	-	-	-	6min
Magic123 <sup>†</sup> [108]	$\checkmark$	×	14.501	0.874	0.145	-	-	-	1h
HumanSGD <sup>†</sup> [22]	$\checkmark$	$\checkmark$	17.365	0.895	0.130	-	-	-	7min
TeCH <sup>†</sup> [21]	$\checkmark$	$\checkmark$	25.211	0.936	0.083	<u>21.192</u>	<u>0.884</u>	<u>0.188</u>	4.5h
HumanSplat (Ours)	$\checkmark$	$\checkmark$	24.033	<u>0.918</u>	0.055	23.346	0.913	0.125	9.3s

### **Qualitative Comparison** •••

Qualitative comparison against SOTA methods.

