

3DGS-Enhancer: Enhancing Unbounded 3D Gaussian Splatting with View-consistent 2D Diffusion Priors

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Motivation and Key Hypothesis

Method

Problem: Sparse-view 3D reconstructions often result in significant artifacts and inconsistencies.

Hypothesis: Multi-view Consistency in 3D space can be learned as Temporal Consistency in 2D space.

Solution: Fine-tune a **multi-view consistent** video diffusion model using 3D reconstruction data to enhance unbounded 3D Gaussian splatting (3DGS)

DL3DV-10K In-Domain

Ours (PSNR: 22.26)

DL3DV (130 training scenes, 20 test scenes)

3DGS-Enhancer (ours) 14.33 0.424 0.464 16.94 0.565 0.356 18.50 0.630 0.305

3 views PSNR↑ SSIM↑ LPIPS↓ PSNR↑ SSIM↑ LPIPS↓ PSNR↑ SSIM↑ LPIP

0.618 11.56 0.199 0.608

Experiments

3DGS (PSNR: 12.83)

Mip-NeRF

RegNeRF [27 FreeNeRF [4

3DGS [18] DNGaussian [19]

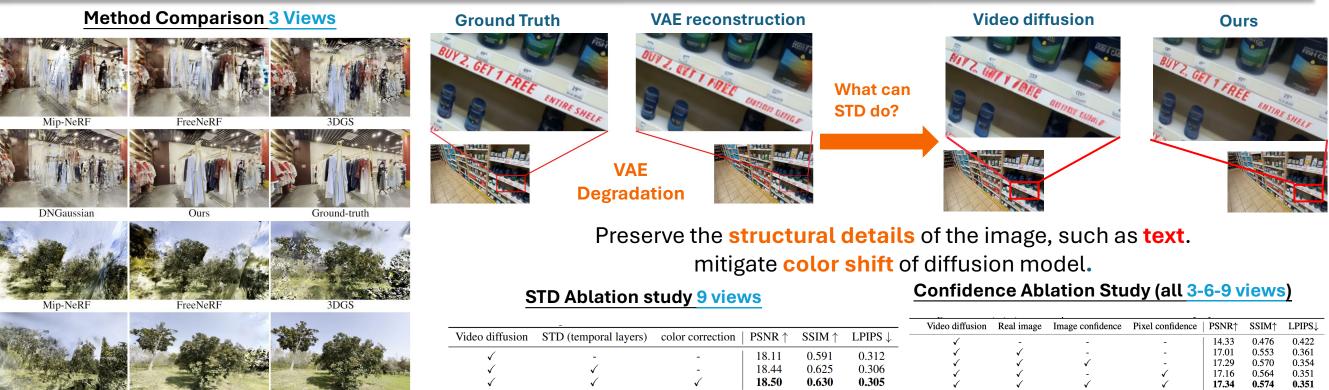




















Ground-truth
9 views

PSNR ↑	SSIM \uparrow	LPIPS ↓	PSNR↑	SSIM \uparrow	LPIPS
Mip-N	le RF360 (a	all test scene	s)		
13.08	0.159	0.637	13.73	0.189	0.628
12.69	0.175	0.660	13.73	0.193	0.629
12.56	0.182	0.646	13.20	0.198	0.635
11.53	0.144	0.651	12.65	0.187	0.607
11.81	0.208	0.689	12.51	0.228	0.683
13.96	0.260	0.570	16.22	0.399	0.454





	PSNR↑	6 views SSIM ↑	LPIPS ↓	PSNR↑	9 views SSIM ↑	
	Mip-N	leRF360 (a	all test scene	s)		
	13.08	0.159	0.637	13.73	0.189	
	12.69	0.175	0.660	13.73	0.193	
	12.56	0.182	0.646	13.20	0.198	
	11.53	0.144	0.651	12.65	0.187	
	11.81	0.208	0.689	12.51	0.228	
ncer (ours)	13.96	0.260	0.570	16.22	0.399	



)		Ours (PSNF	R: 22
	DEND +	6 views	

NR ↑	6 views SSIM ↑	LPIPS ↓	PS
Mip-N	leRF360 (a	all test scene	s)

OGS-Enhancer (ours)	13.96	0.260	0.5
NGaussian	11.81	0.208	0.6
DGS	11.53	0.144	0.6
eenekr	12.50	0.182	0.0

3DGS (PSNR: 13.9

Method

Mip-NeRF

RegNeRF

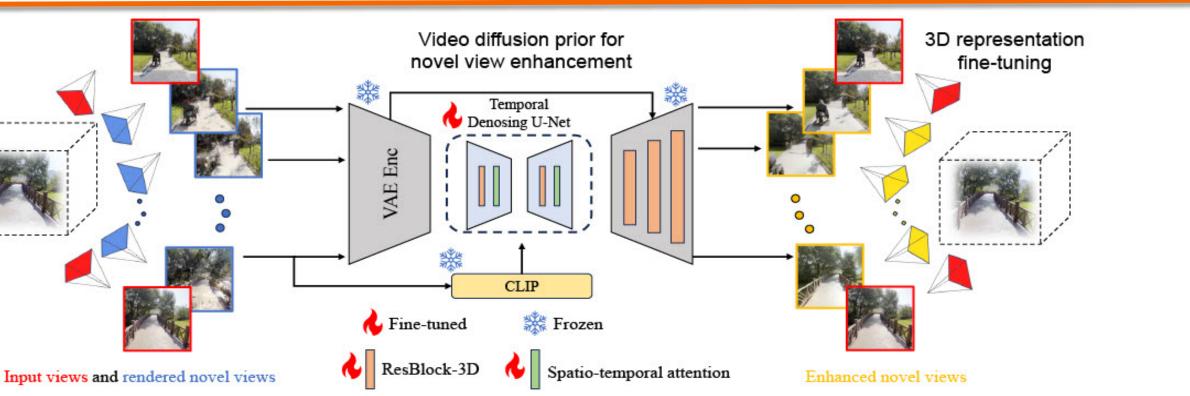
🗡 We achieved SOTA performance on both in-domain and out-of-domain data.7

Ground-truth





Clemson Vision Learning Lab



1. Video Diffusion Prior: Finetune an view-consistency video diffusion model for enhancing novel view synthesis 2. Spatial-Temporal Decoder: Boost the and mitigate the VAE degradation

3. Confidence-aware: Reduce the impact of non-photorealistic and inconsistent regions generated by diffusion models.

Confidence Ablation Study (all 3-6-9 views)

	-					Video diffusion	Real image	Image confidence	Pixel confidence	PSNR↑	SSIM↑	LPIPS
Video diffusion	STD (temporal layers)	color correction	PSNR ↑	SSIM \uparrow	LPIPS \downarrow				_	14.33	0.476	0.42
(10 11	0.591	0.312	v	 ✓	-	-	17.01	0.553	0.36
V	-	-	18.11			\checkmark	\checkmark	\checkmark	-	17.29	0.570	0.35
V	V	-	18.44	0.625	0.306	\checkmark	\checkmark	-	\checkmark	17.16	0.564	0.35
\checkmark	\checkmark	\checkmark	18.50	0.630	0.305	\checkmark	\checkmark	\checkmark	\checkmark	17.34	0.574	0.3

PSNR: 18.11 -> 18.44

PSNR: 17.01 -> 17.34