

E-3DGS: Event-Based Novel View Rendering of Large-Scale Scenes Using 3D Gaussian Splatting

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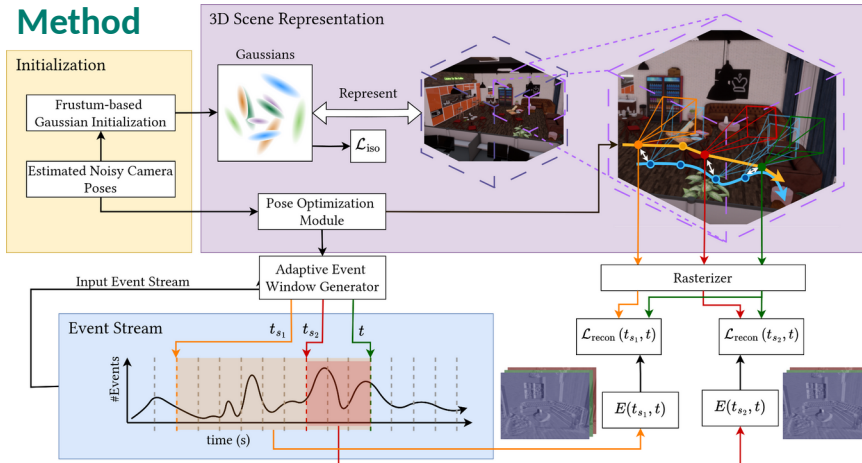


Project Page, Code & Data
 4dqv.mpi-inf.mpg.de/E3DGS/

Contributions

- E-3DGS, the first approach for novel view synthesis from a color event camera that combines 3D Gaussians with event-based supervision
- Frustum-based initialization, adaptive event windows, isotropic 3D Gaussian regularization and 3D camera pose refinement
- New synthetic and real event datasets for large scenes to study novel view synthesis in the new problem setting

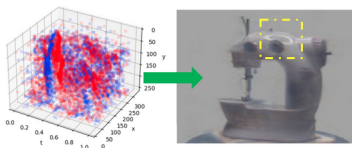
Method



$$\mathcal{L}_{recon}(t_{s1}, t) = \frac{\alpha}{|\mathcal{X}_{noevs}|} \cdot \left(\sum_{\mathbf{x} \in \mathcal{X}_{noevs}} \mathcal{L}_{\mathbf{x}}(t_{s1}, t) \right) + \frac{1-\alpha}{|\mathcal{X}_{evs}|} \cdot \left(\sum_{\mathbf{x} \in \mathcal{X}_{evs}} \mathcal{L}_{\mathbf{x}}(t_{s1}, t) \right)$$

$$\mathcal{L} = \lambda_1 \mathcal{L}_{recon}(t_{s1}, t) + \lambda_2 \mathcal{L}_{recon}(t_{s2}, t) + \lambda_{iso} \mathcal{L}_{iso} + \lambda_{pose} \mathcal{L}_{pose}$$

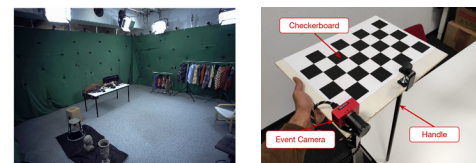
Related Works



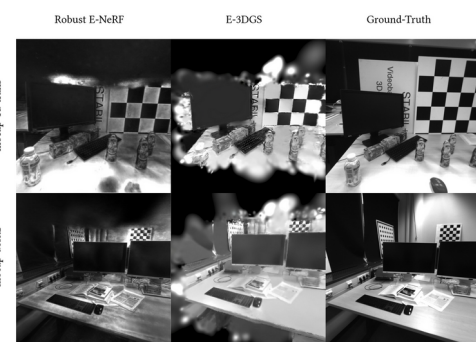
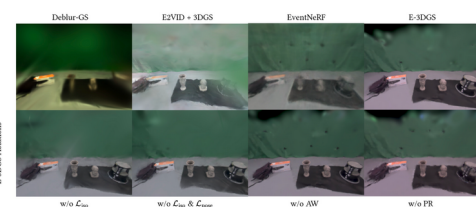
EventNeRF [Rudnev et al., CVPR 2023]

- Supports single object only/Scene extent has to be known in advance
- Comparably slow at training and rendering
- Visual accuracy is behind the RGB-based alternatives

New Real Dataset



Results on Real Data



Results on Synthetic Data



Method	Company			ScienceLab			Subway			Average		
	↑PSNR	↓LPIPS	↑SSIM	↑PSNR	↓LPIPS	↑SSIM	↑PSNR	↓LPIPS	↑SSIM	↑PSNR	↓LPIPS	↑SSIM
EventNeRF [25]	19.59	0.41	0.65	17.22	0.46	0.60	18.71	0.34	0.67	16.80	0.50	0.61
E2VID [23] + 3DGS [9]	9.79	0.37	0.48	11.86	0.38	0.54	9.79	0.40	0.43	10.48	0.38	0.49
E-3DGS (ours)	20.78	0.29	0.72	18.41	0.28	0.73	19.92	0.20	0.74	19.70	0.26	0.73

Scene	EventNeRF [25]			E-3DGS (ours)		
	↑PSNR	↓LPIPS	↑SSIM	↑PSNR	↓LPIPS	↑SSIM
Chair	30.62	0.05	0.94	30.42	0.03	0.95
Drums	27.43	0.07	0.91	31.07	0.03	0.95
Ficus	31.94	0.05	0.94	34.08	0.02	0.96
Hooding	30.26	0.04	0.94	30.73	0.03	0.96
Lego	25.84	0.13	0.89	30.74	0.04	0.94
Materials	24.10	0.07	0.94	33.73	0.02	0.97
Mic	31.78	0.03	0.96	35.87	0.02	0.98
Average	28.85	0.06	0.93	32.39	0.03	0.96

Comparisons on the synthetic EventNeRF dataset.

References:

Rebecq et al. High Speed and High Dynamic Range Video with an Event Camera. TPAMI, 2019.
 Rudnev et al. EventNeRF: Neural Radiance Fields from a Single Colour Event Camera. CVPR 2023.
 Low and Lee. Robust e-NeRF: NeRF from Sparse & Noisy Events under Non-uniform Motion. ICCV, 2023.
 Kerbl et al. 3D Gaussian Splatting for Real-Time Radiance Field Rendering. ACM TOG, 2023.