Embracing Dynamics:

Dynamics-aware 4D Gaussian Splatting SLAM

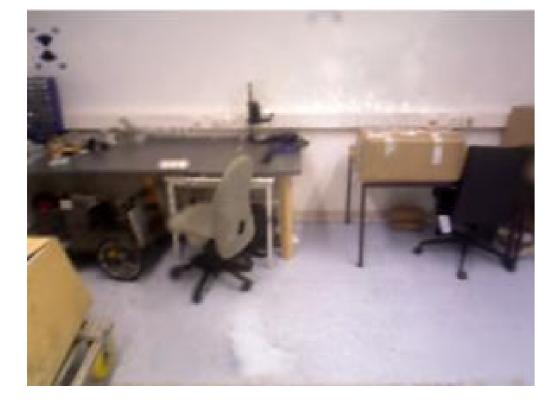
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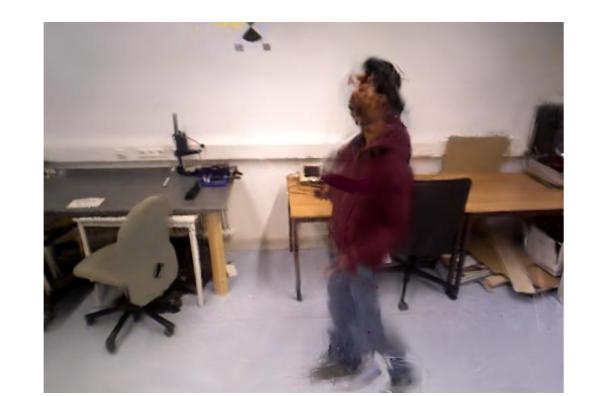
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1. Motivations

- Gaussian Splatting (3DGS): Static map representation
- **Limitations** of 3DGS-based SLAM:



a) Incomplete map (Source: Xu, Y., Jiang, H., Xiao, Z., Feng, J., & Zhang, L. (2024). DG-SLAM: Robust Dynamic Gaussian Splatting SLAM with Hybrid Pose Optimization. NeurIPS 2024)



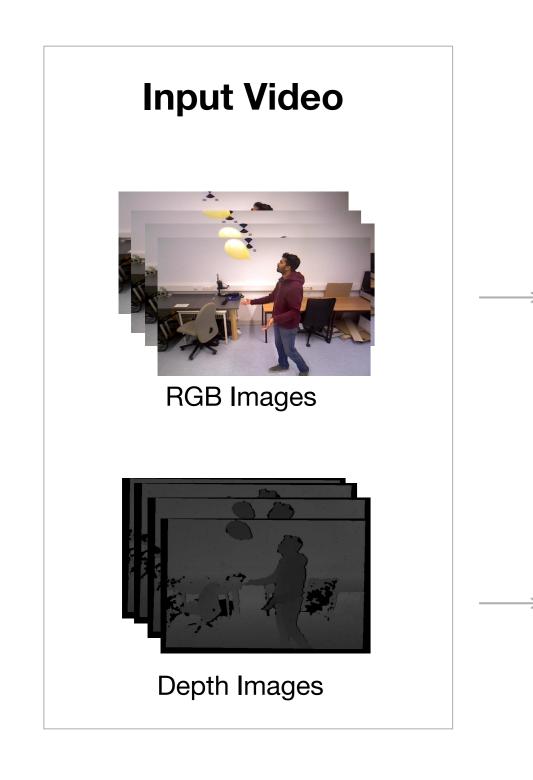
b) Map with Artifacts

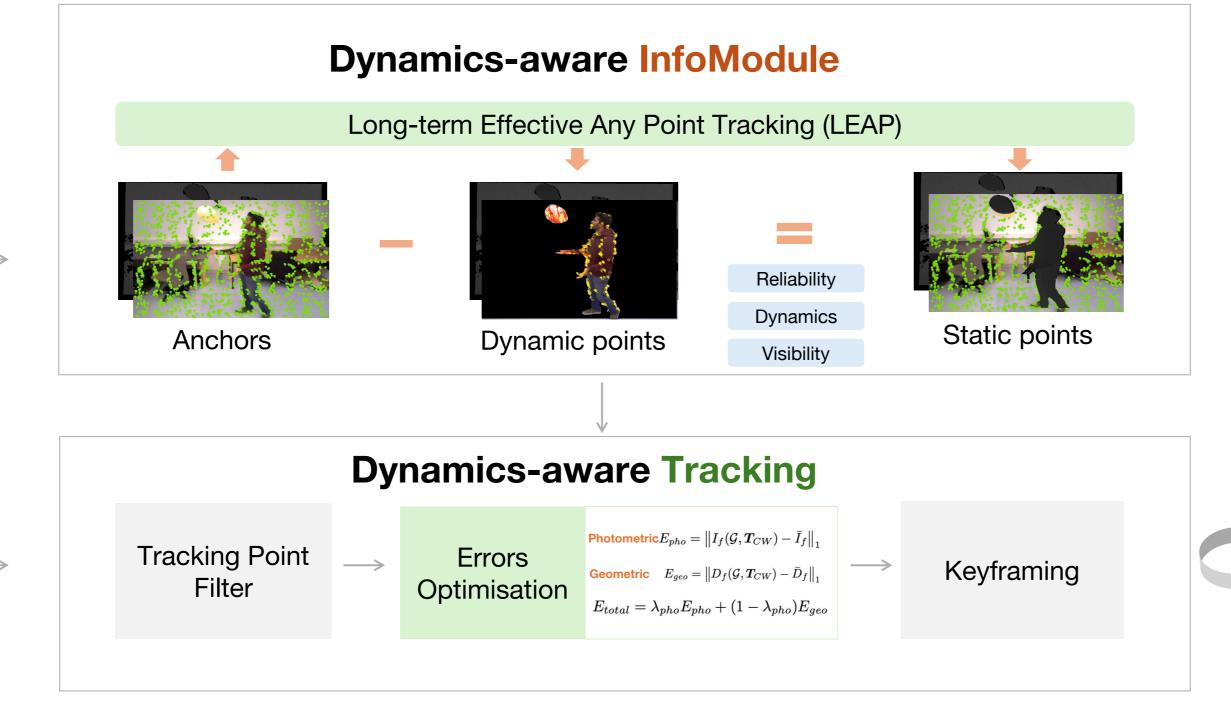
2. Contributions

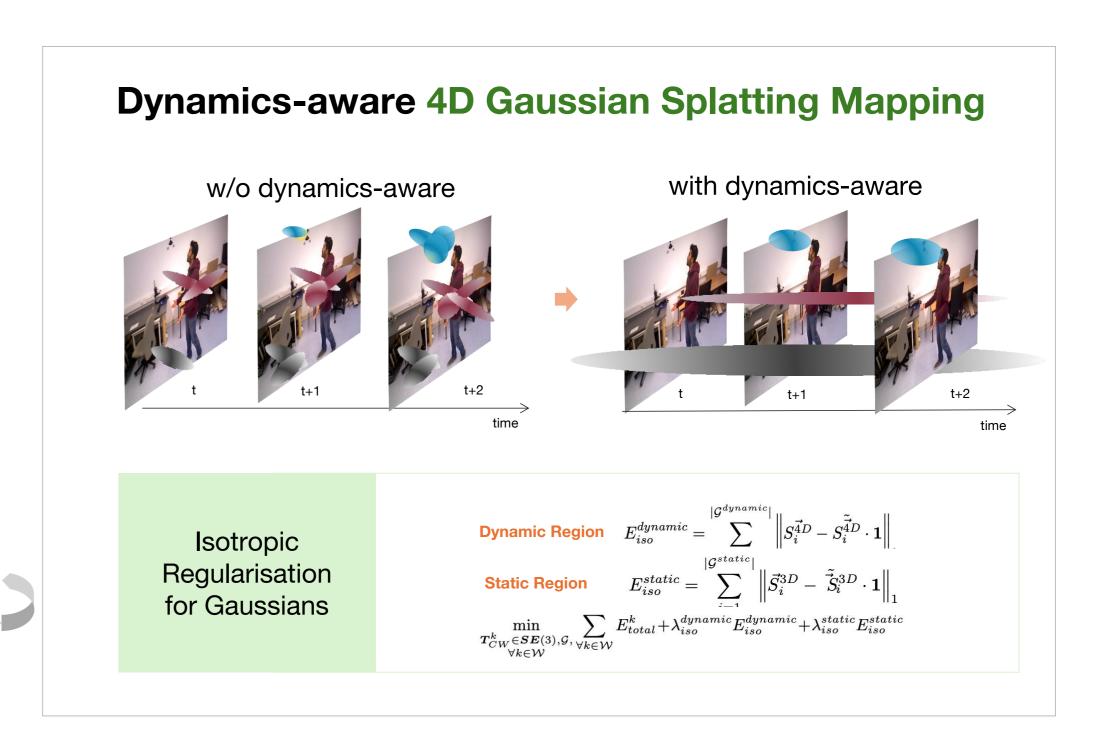
- Proposed D4DGS-SLAM, the 4DGS-based SLAM system for dynamic environments
- Enhanced the SLAM system of with **dynamics-aware** InfoModule
- **SOTA** tracking and mapping performance on multiple dynamic SLAM benchmarks



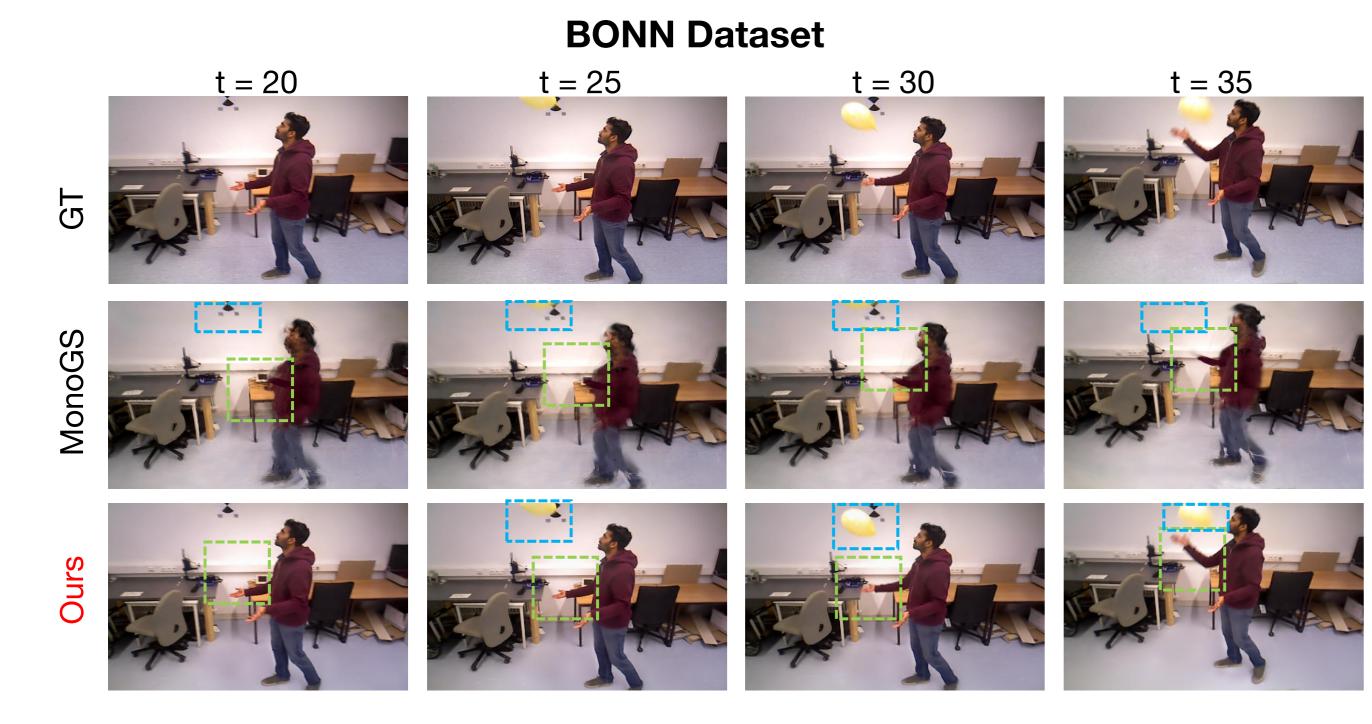
3. Pipeline







4. Results



Results of metric ATE RMSE on several dynamic scene sequences in BONN dataset. *" denotes the version reproduced by NICE-SLAM. "-" denotes the tracking failures. The metric unit is [cm].

Method	ball	ball2	ps_tk	ps_tk2	ball_tk	Avg
ORB-SLAM3 [2]	5.8	17.7	70.7	77.9	3.1	29.
ReFusion [12]	17.5	25.4	28.9	46.3	30.2	27.
DROID-VO[19]	5.4	4.6	21.4	46.0	8.9	15.4
iMAP*[17]	14.9	67.0	28.3	52.8	24.8	36.
NICE-SLAM[30]	-	66.8	54.9	45.3	21.2	44.
Vox-Fusion[25]	65.7	82.1	128.6	162.2	43.9	88.
Co-SLAM[20]	28.8	20.6	61.0	59.1	38.3	46.
ESLAM[5]	22.6	36.2	48.0	51.4	12.4	31.
Rodyn-SLAM[4]	7.9	11.5	14.5	13.8	13.3	12.
SplaTAM[6]	35.5	36.1	149.7	91.2	12.5	57.
GS-SLAM[24]	37.5	26.8	46.8	50.4	31.9	33.
DG-SLAM [22]	3.7	4.1	4.5	6.9	10.0	5.5
Ours	3.6	3.9	4.5	5.2	8.5	5.1

Ablation Study on the Ball Scene in the Bonn Dataset.

	ATE RMSE↓	PSNR↑	SSIM↑	LPIPS↓
w/o D-aware	27.9	20.23	0.790	0.371
w/o 4DGS	7.2	18.23	0.645	0.327
Ours	3.6	27.89	0.857	0.236



Camera tracking and mapping quality on several dynamic sequences in the Tartanair-Shibuya dataset. $\mathsf{ATE}\downarrow \mathsf{PSNR}\uparrow \mathsf{SSIM}\uparrow \mathsf{LPIPS}\downarrow \mathsf{ATE}\downarrow \mathsf{PSNR}\uparrow \mathsf{CPIPS}\downarrow \mathsf{ATE}\downarrow \mathsf{CPIPS}\downarrow \mathsf{ATE}\downarrow \mathsf{CPIPS}\downarrow \mathsf{ATE}\downarrow \mathsf{CPIPS}\downarrow \mathsf{CPI$
 53.0
 13.91
 0.303
 0.576
 58.8
 13.21
 0.241
 0.638
 82.8
 16.63
 0.516
 0.404
 85.6
 14.74
 0.489
 0.534
 70.1
 14.62
 0.387
 0.538

 3.2
 23.48
 0.673
 0.268
 4.1
 24.38
 0.686
 0.267
 2.1
 21.39
 0.793
 0.243
 5.1
 21.73
 0.669
 0.387
 3.6
 22.75
 0.705
 0.291

Map quality on several dynamic sequences in the BONN dataset. | PSNR↑ SSIM↑ LPIPS↓
 SplaTAM [6]
 17.59
 0.766
 0.244
 16.81
 0.650
 0.332
 18.90
 0.655
 0.270
 17.25
 0.721
 0.263
 15.55
 0.633
 0.413
 17.22
 0.685
 0.304

 MonoGS [9]
 17.72
 0.712
 0.478
 19.44
 0.747
 0.367
 18.8
 0.736
 0.399
 20.01
 0.755
 0.375
 18.89
 0.623
 0.272
 18.97
 0.715
 0.378
 27.89 0.857 0.236 29.65 0.839 0.272 27.66 0.832 0.265 31.18 0.876 0.259 27.19 0.865 0.264 28.71 0.854 0.259







