



中国科学技术大学
University of Science and Technology of China



MotionGS: Exploring Explicit Motion Guidance for Deformable 3D Gaussian Splatting

Ruijie Zhu*, Yanzhe Liang*, Hanzhi Chang, Jiacheng Deng, Jiahao Lu,
Wenfei Yang, Tianzhu Zhang†, Yongdong Zhang

*Equal contribution

†Corresponding author



Problem Statement

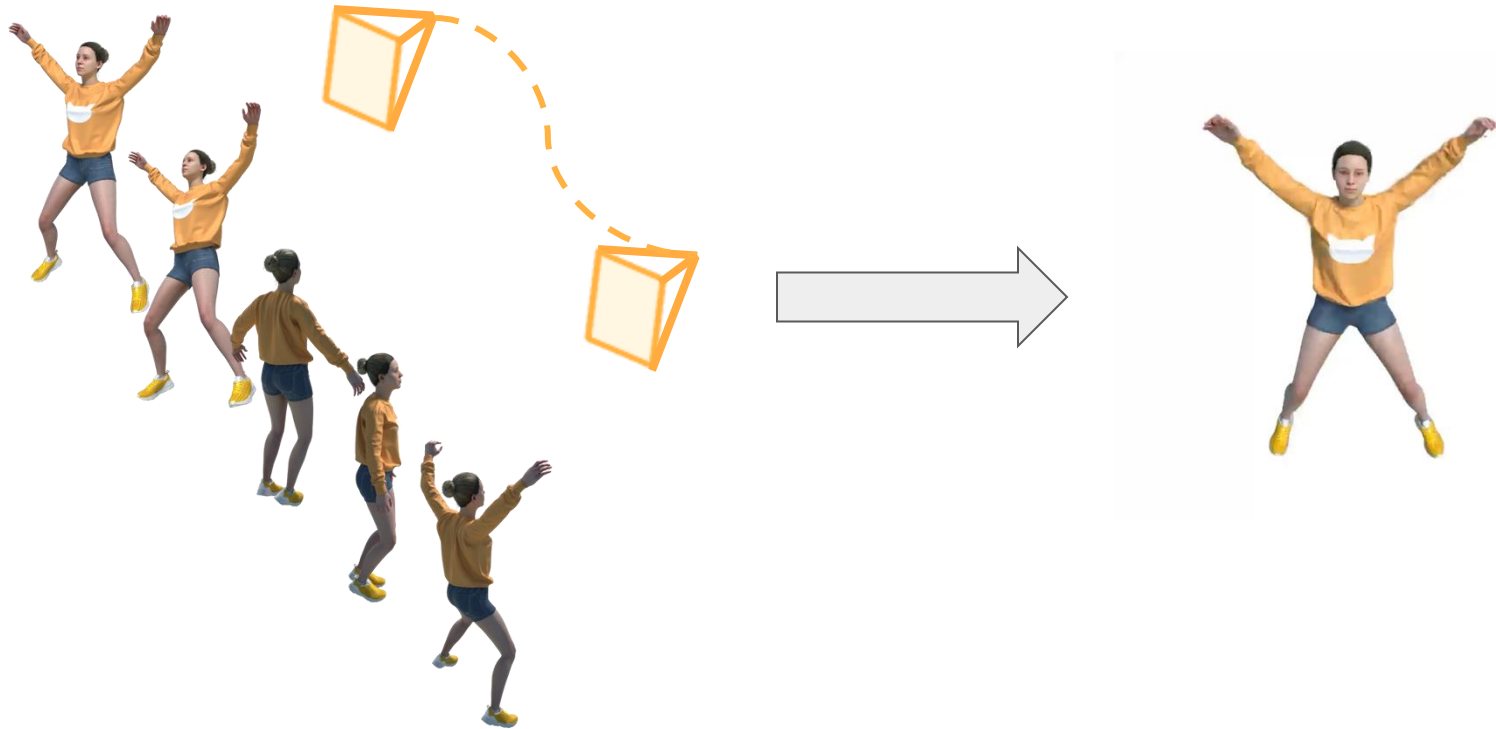
Input: Multi-view images captured in a **Dynamic** scene (usually by **a moving camera**) and their poses



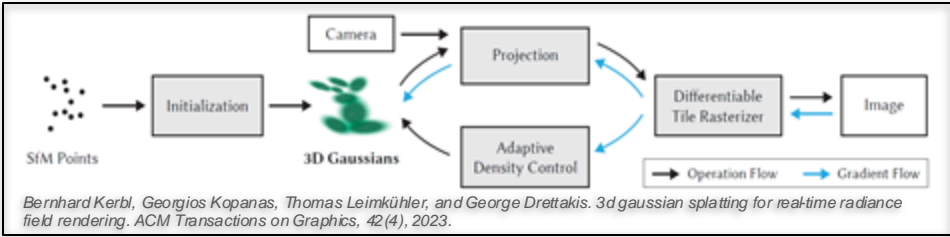
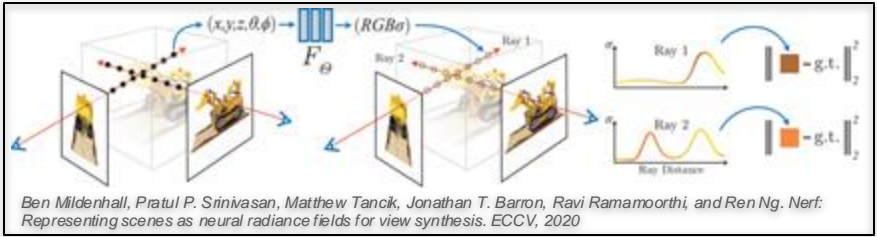
Problem Statement

Input: Multi-view images captured in a **Dynamic** scene (usually by **a moving camera**) and their poses

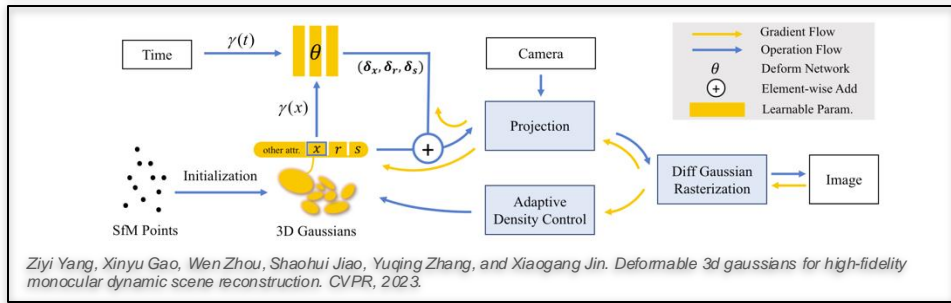
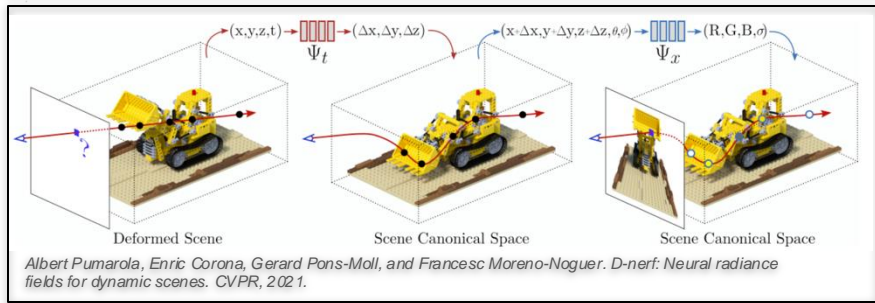
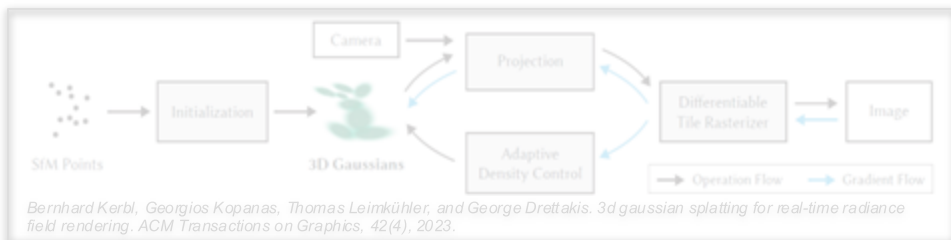
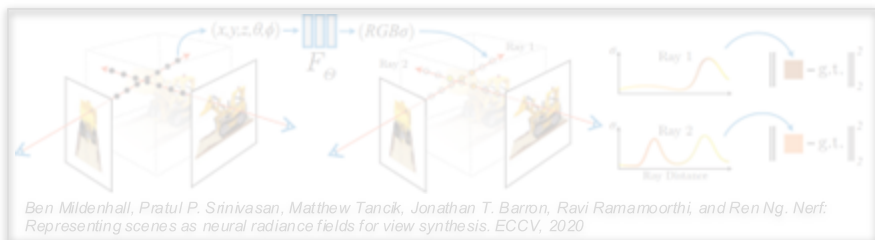
Output: A reconstructed 3D model with **realistic appearance** and **natural motion**, continuously capturing dynamic changes over time.



Related Work



Related Work



Related Work



Ben Mildenhall, Pratul P. Srinivasan, Matthew Tanck, Jonathan T. Barron, Ravi Ramamoorthi, and Ren Ng. Nerf: Representing scenes as neural radiance fields for view synthesis. ECCV, 2020.



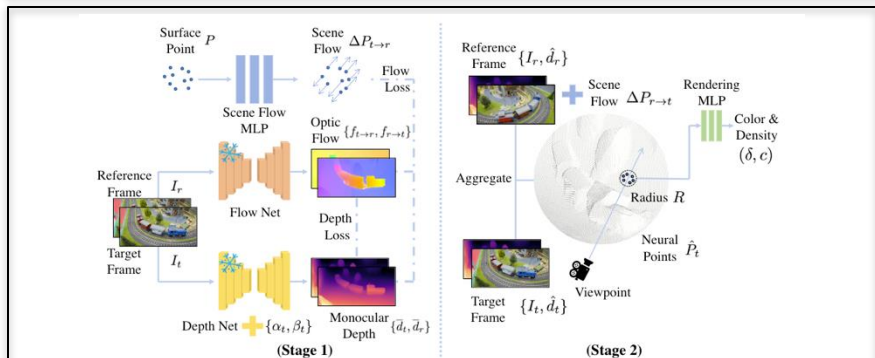
Bernhard Kerbl, Georgios Kopanas, Thomas Leimkühler, and George Drettakis. 3d gaussian splatting for real-time radiance field rendering. ACM Transactions on Graphics, 42(4), 2023.



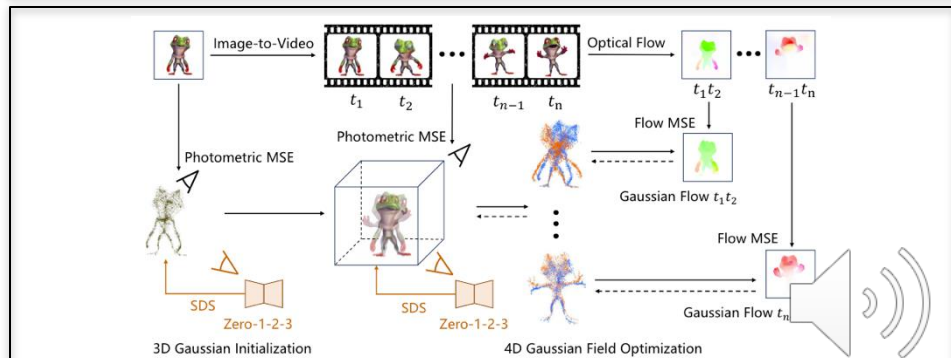
Albert Pumarola, Enric Corona, Gerard Pons-Moll, and Francesc Moreno-Noguer. D-nerf: Neural radiance fields for dynamic scenes. CVPR, 2021.



Ziyi Yang, Xinyu Gao, Wen Zhou, Shaohui Jiao, Yuying Zhang, and Xiaogang Jin. Deformable 3d gaussians for high-fidelity monocular dynamic scene reconstruction. CVPR, 2023.



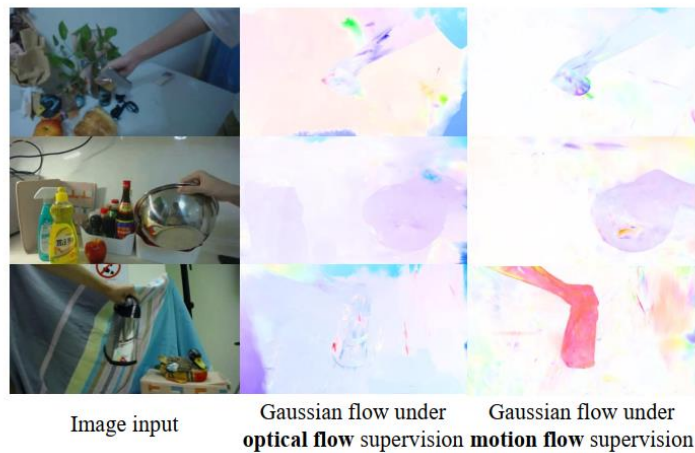
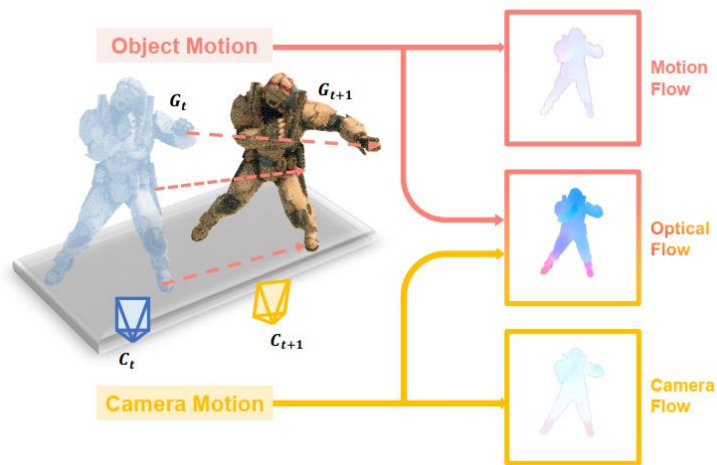
Kaichen Zhou, Jia-Xing Zhong, Sangyun Shin, Kai Lu, Yiyuan Yang, Andrew Markham, and Niki Trigoni. Dynpoint: Dynamic neural point for view synthesis. Advances in Neural Information Processing Systems, 36, 2023.



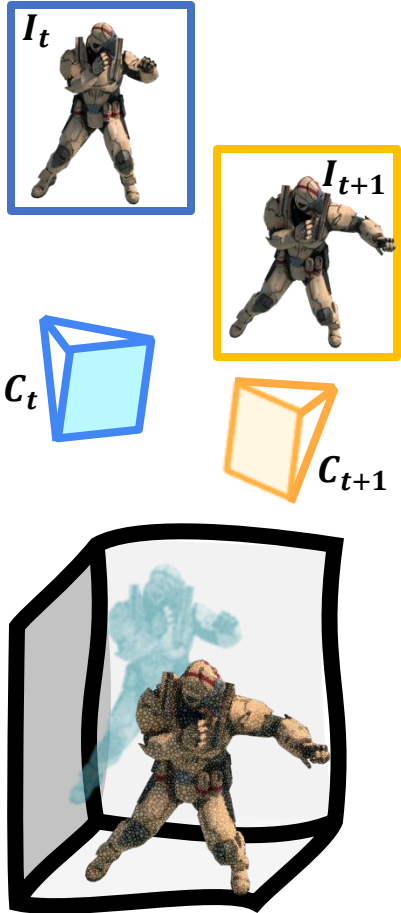
Quankai Gao, Qiangeng Xu, Zhe Cao, Ben Mildenhall, Wenchao Ma, Le Chen, Danhang Tang, and Ullrich Neumann. Gaussianflow: Splatting gaussian dynamics for 4d content creation. arXiv preprint arXiv:2403.12365, 2024.

Our Motivation

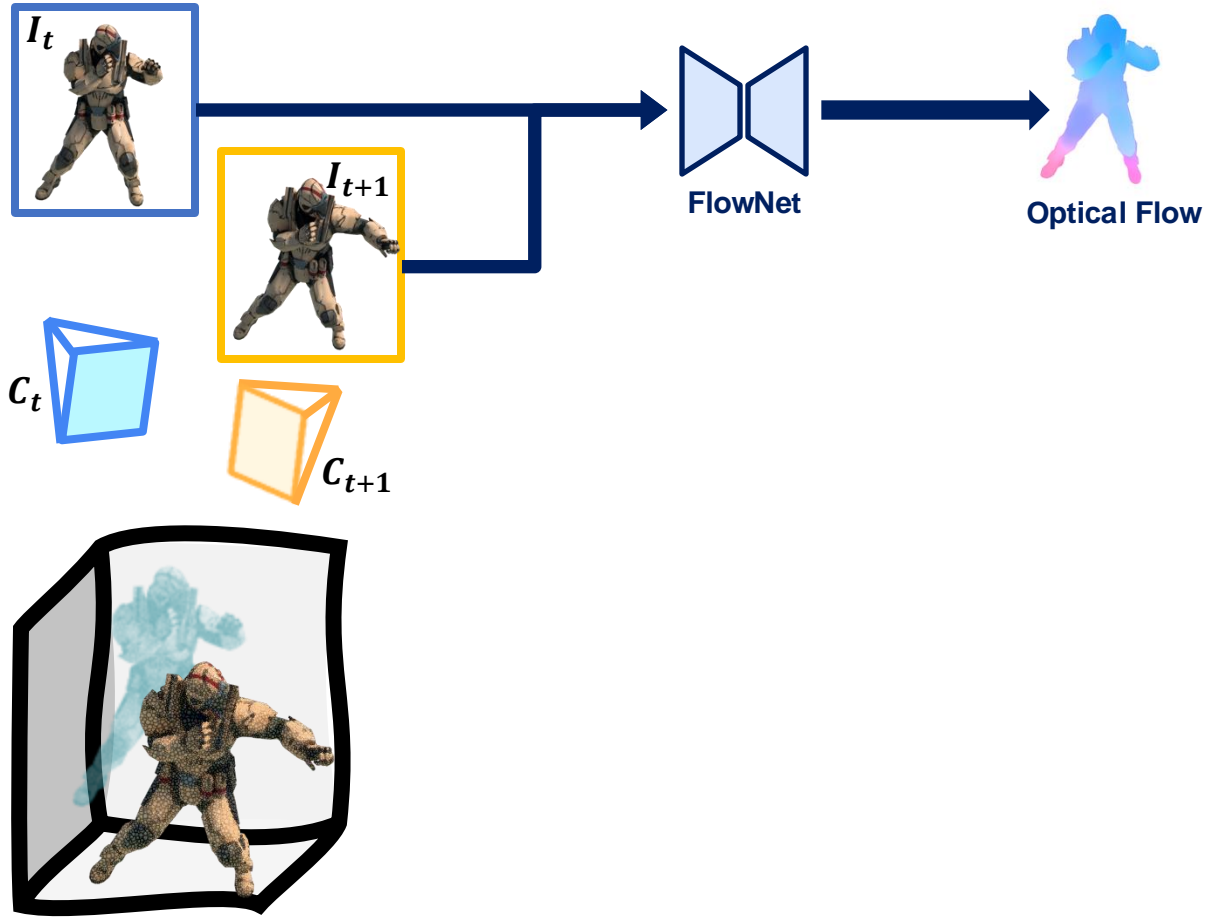
1. Most existing dynamic 3DGS methods simply use deformation fields to model Gaussian motion, which lacks explicit motion constraints.
2. Compared with using optical flow to supervise Gaussian deformation, we decouple motion flow from optical flow to directly provide explicit motion priors, thus making motion constraints more effective.



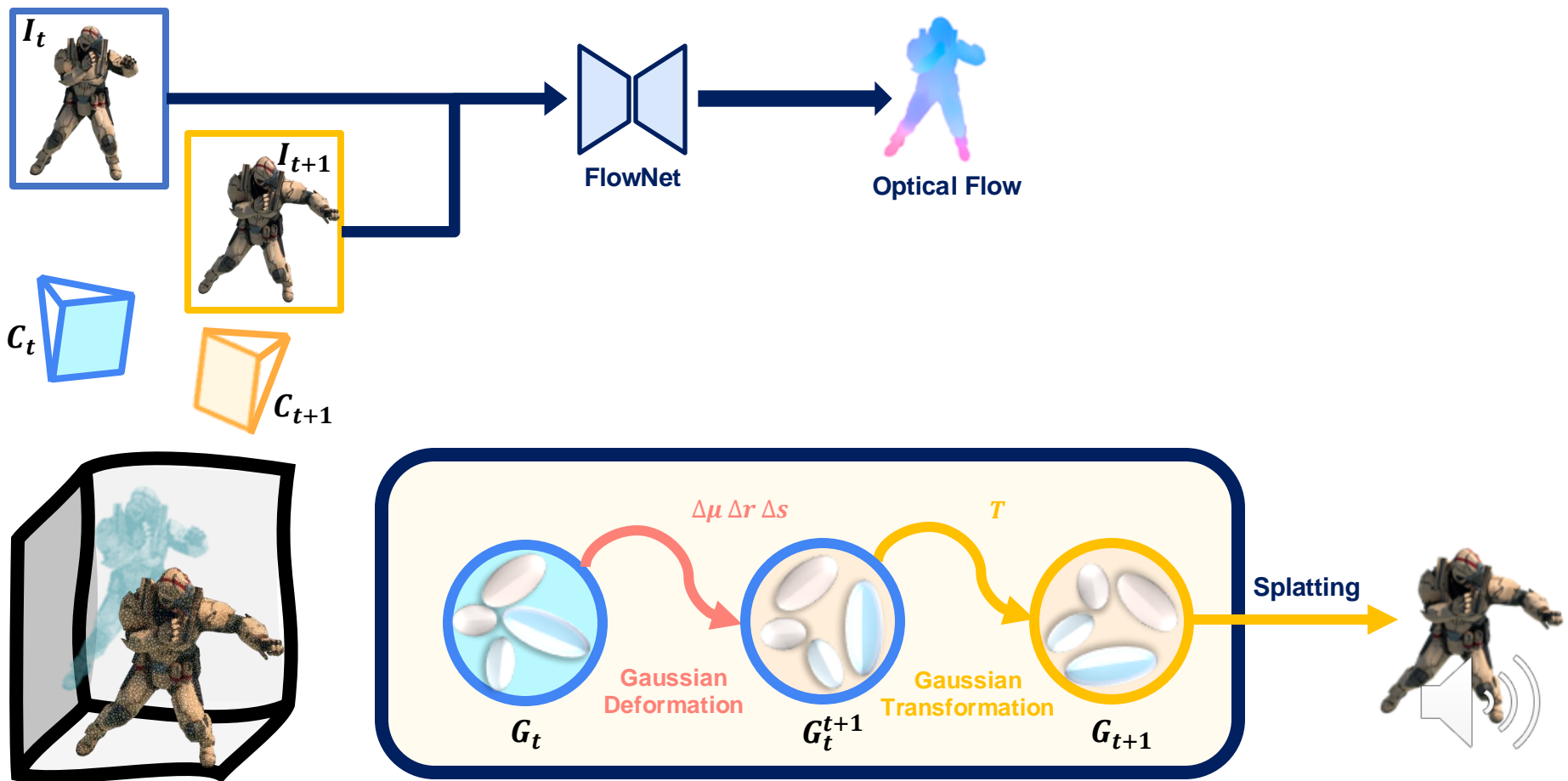
Framework



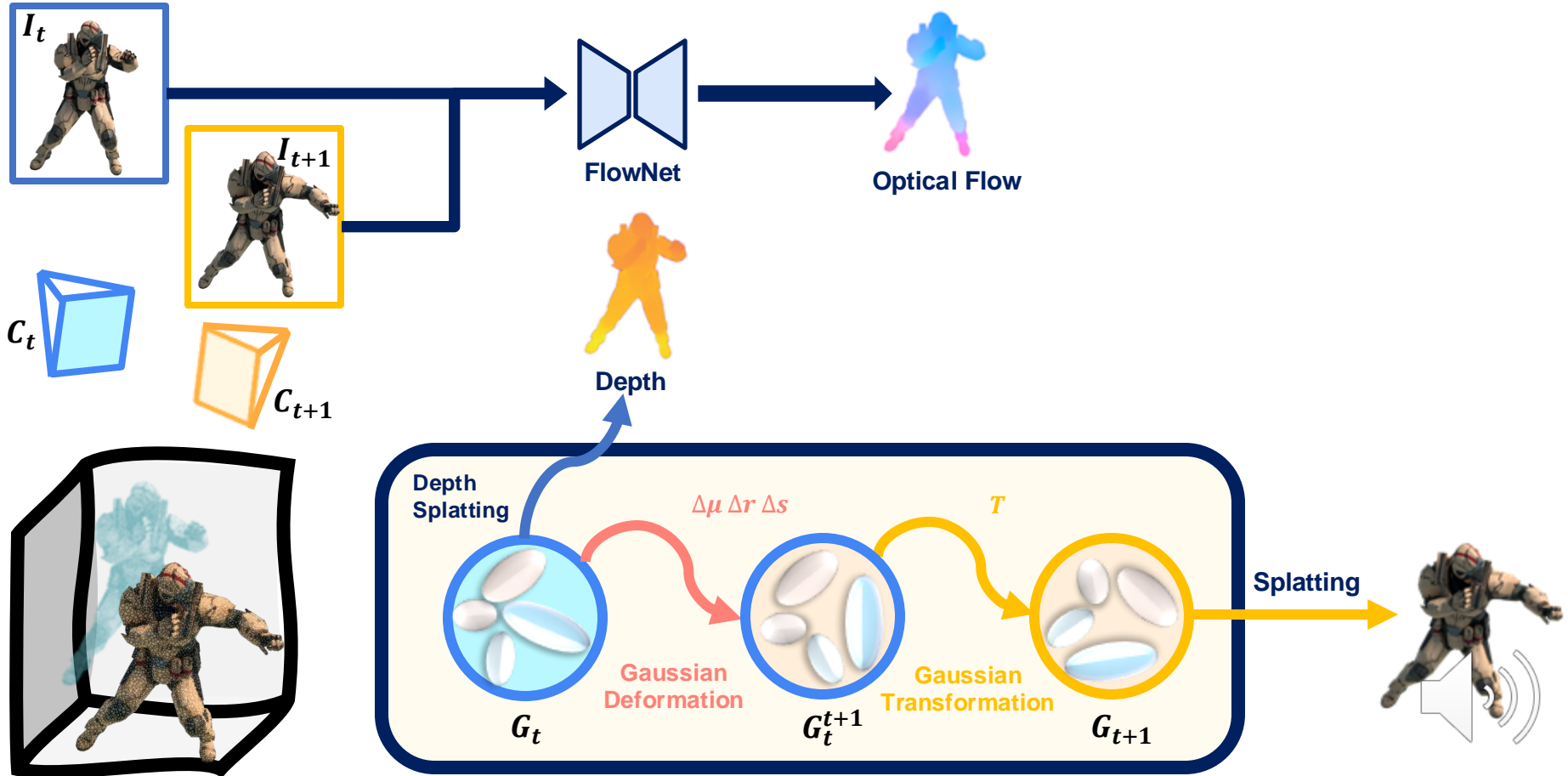
Framework



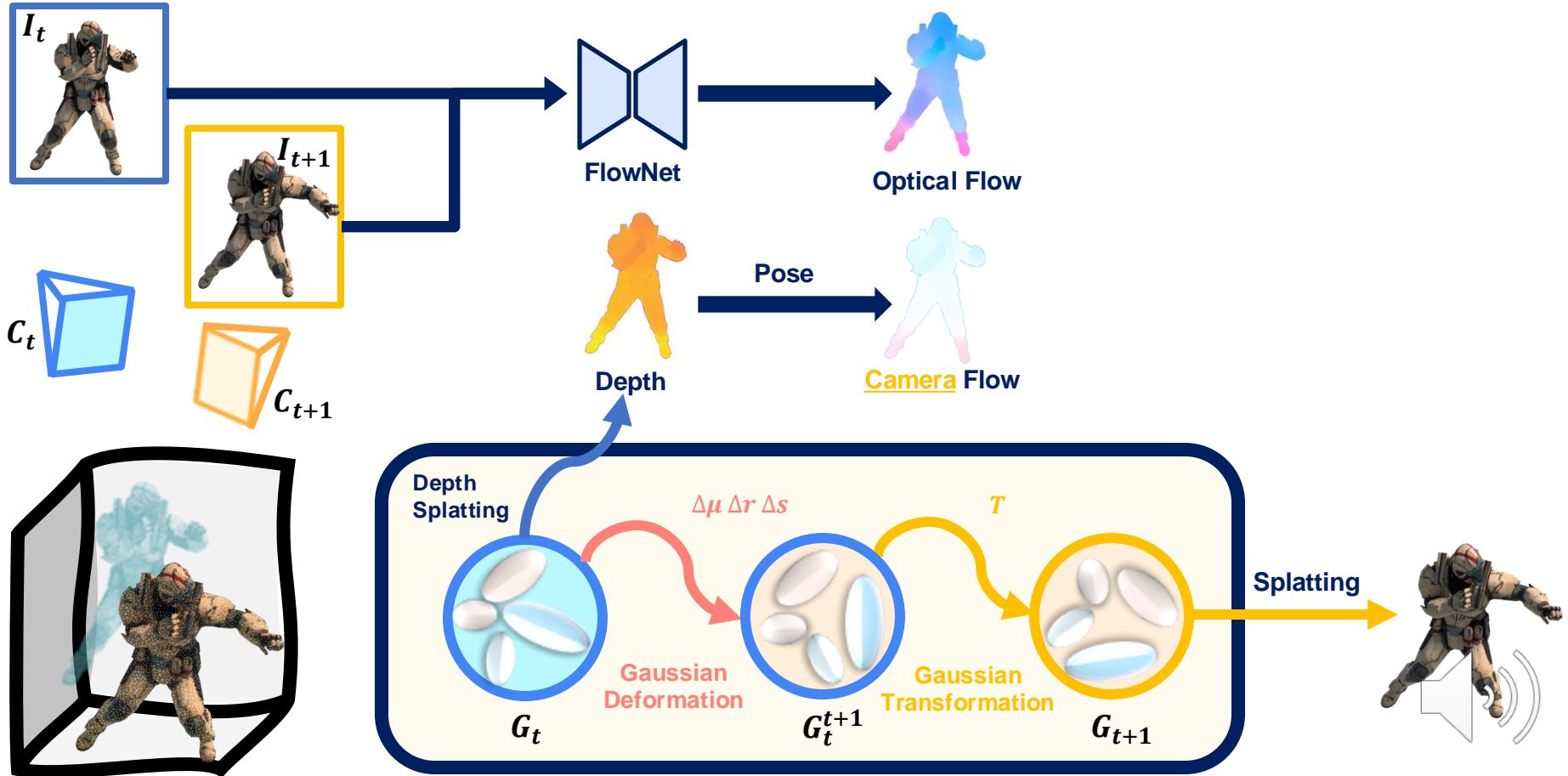
Framework



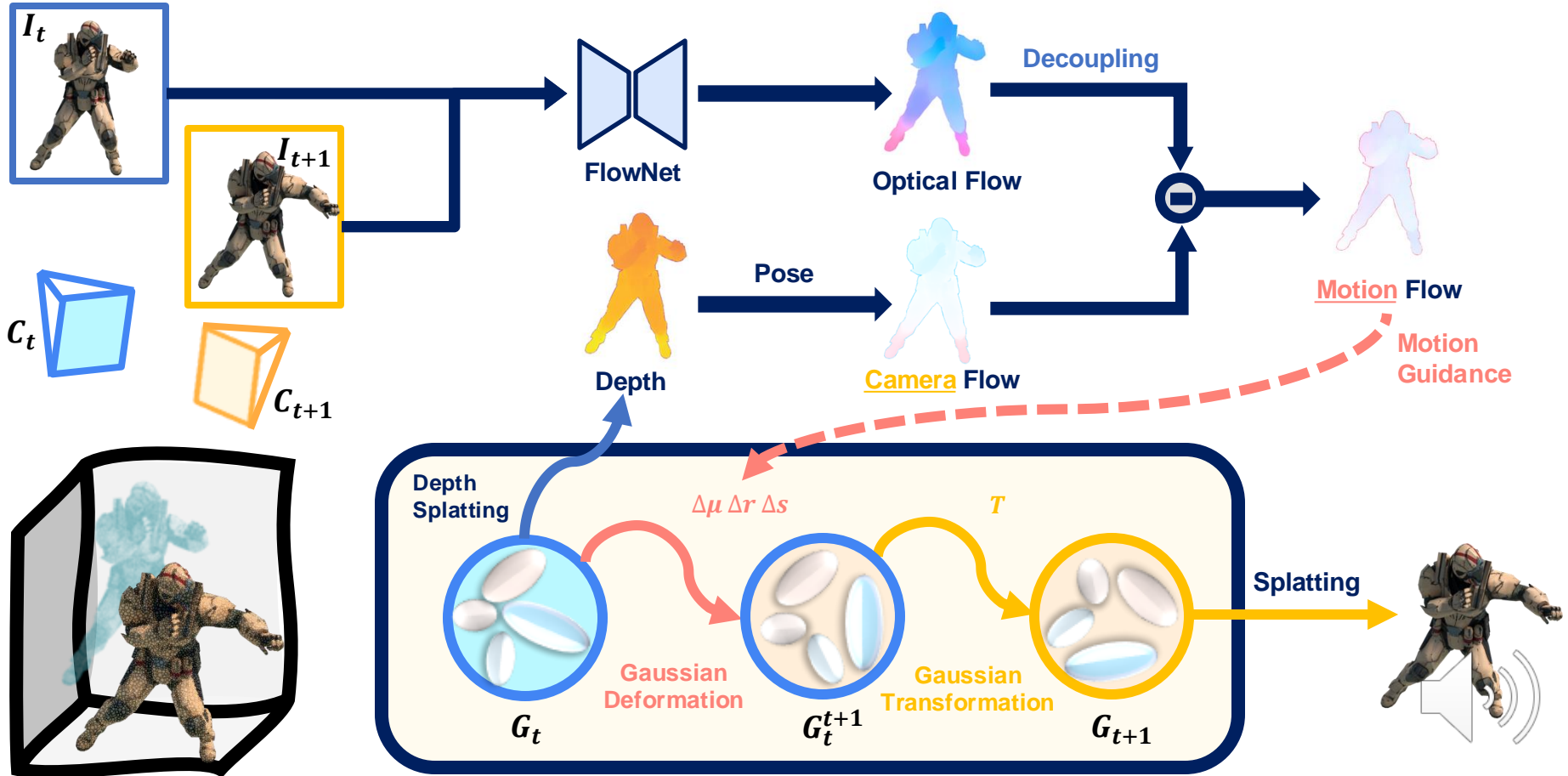
Framework



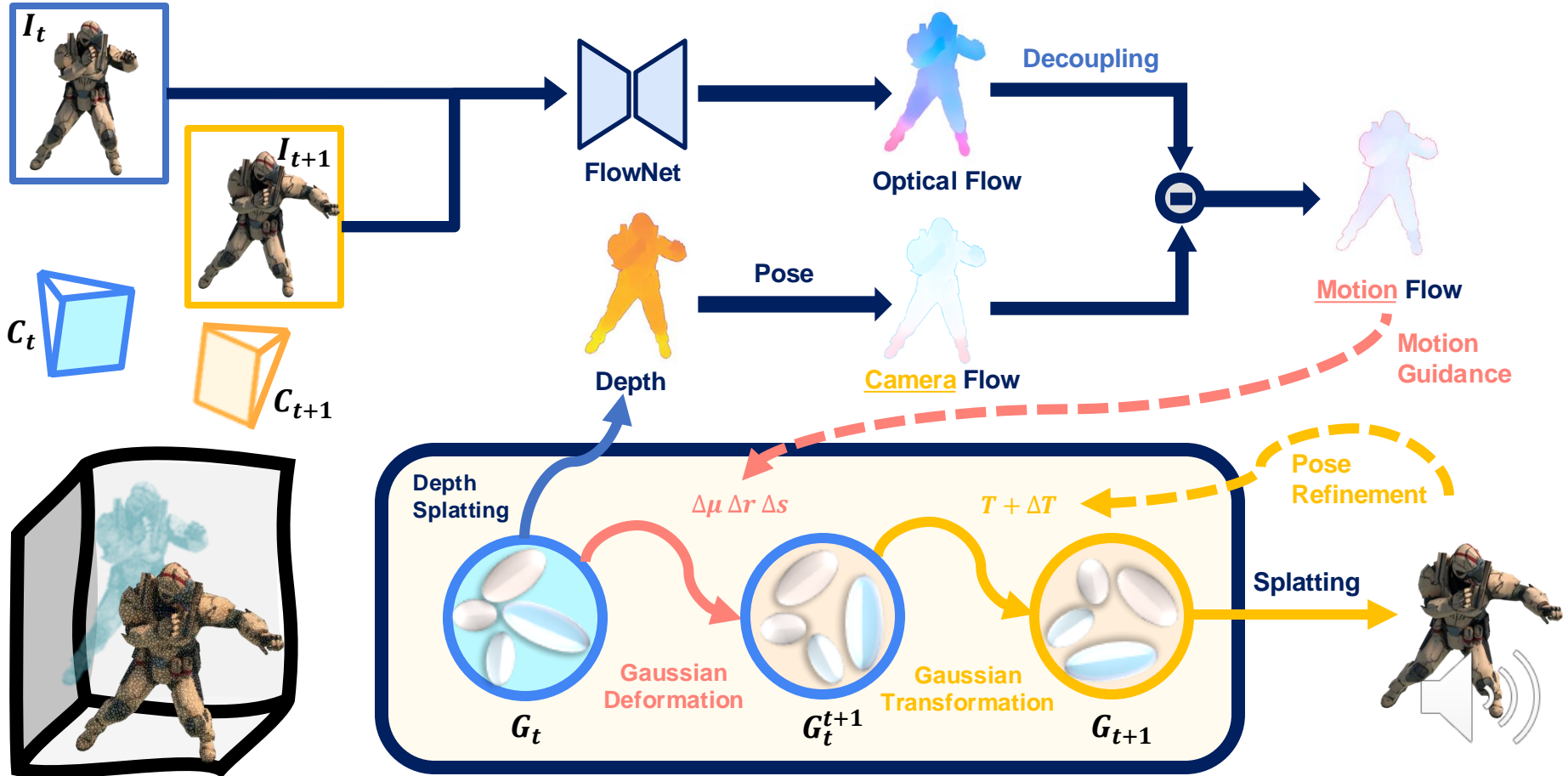
Framework



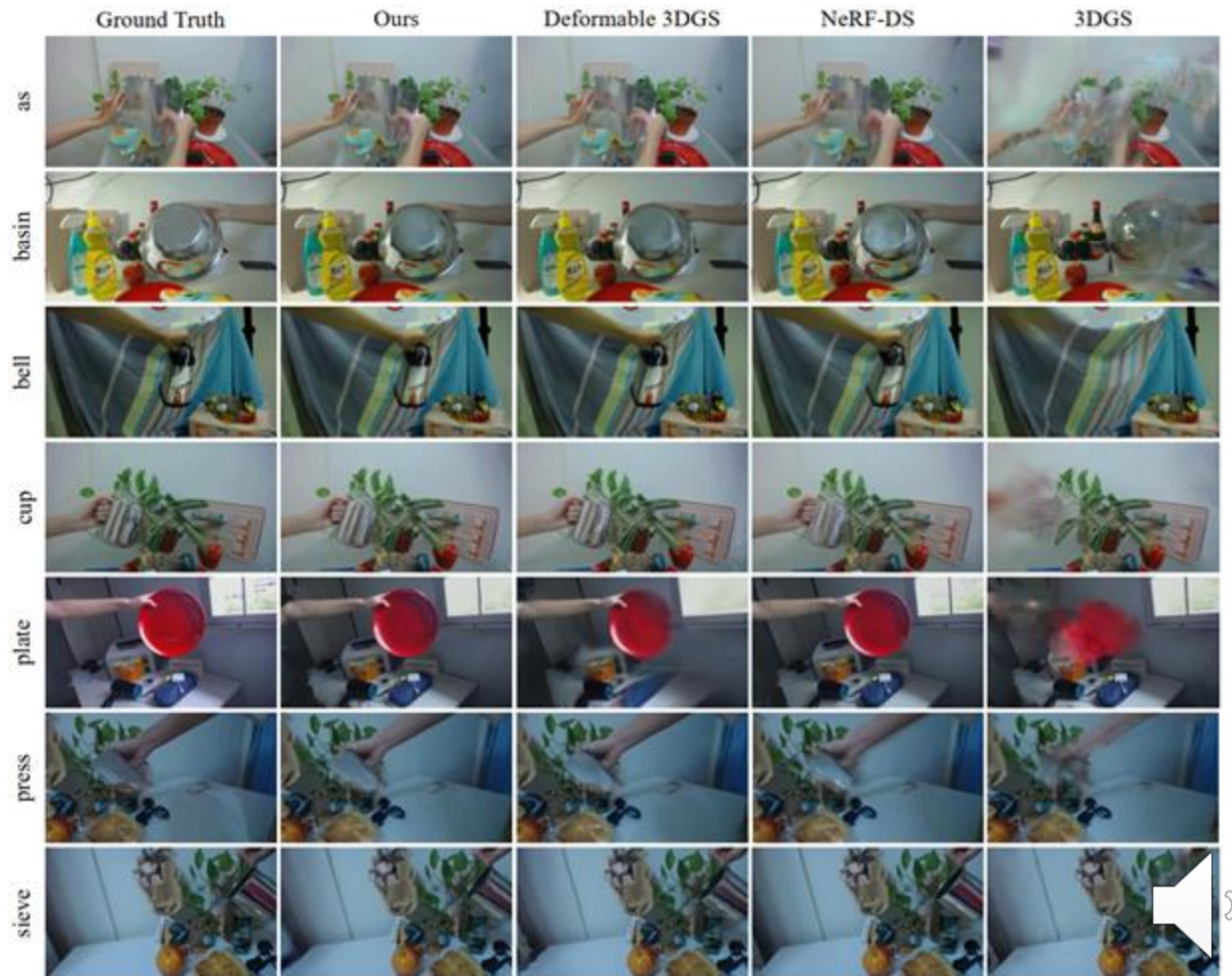
Framework



Framework



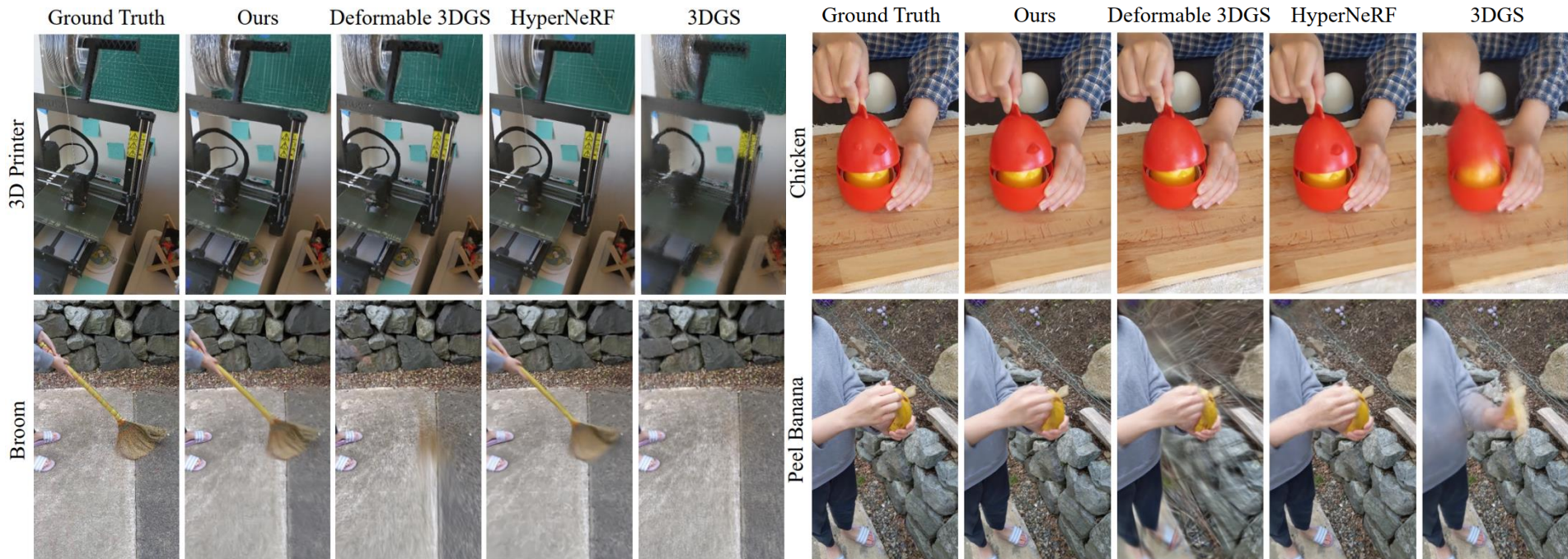
Comparison



NeRF-DS dataset



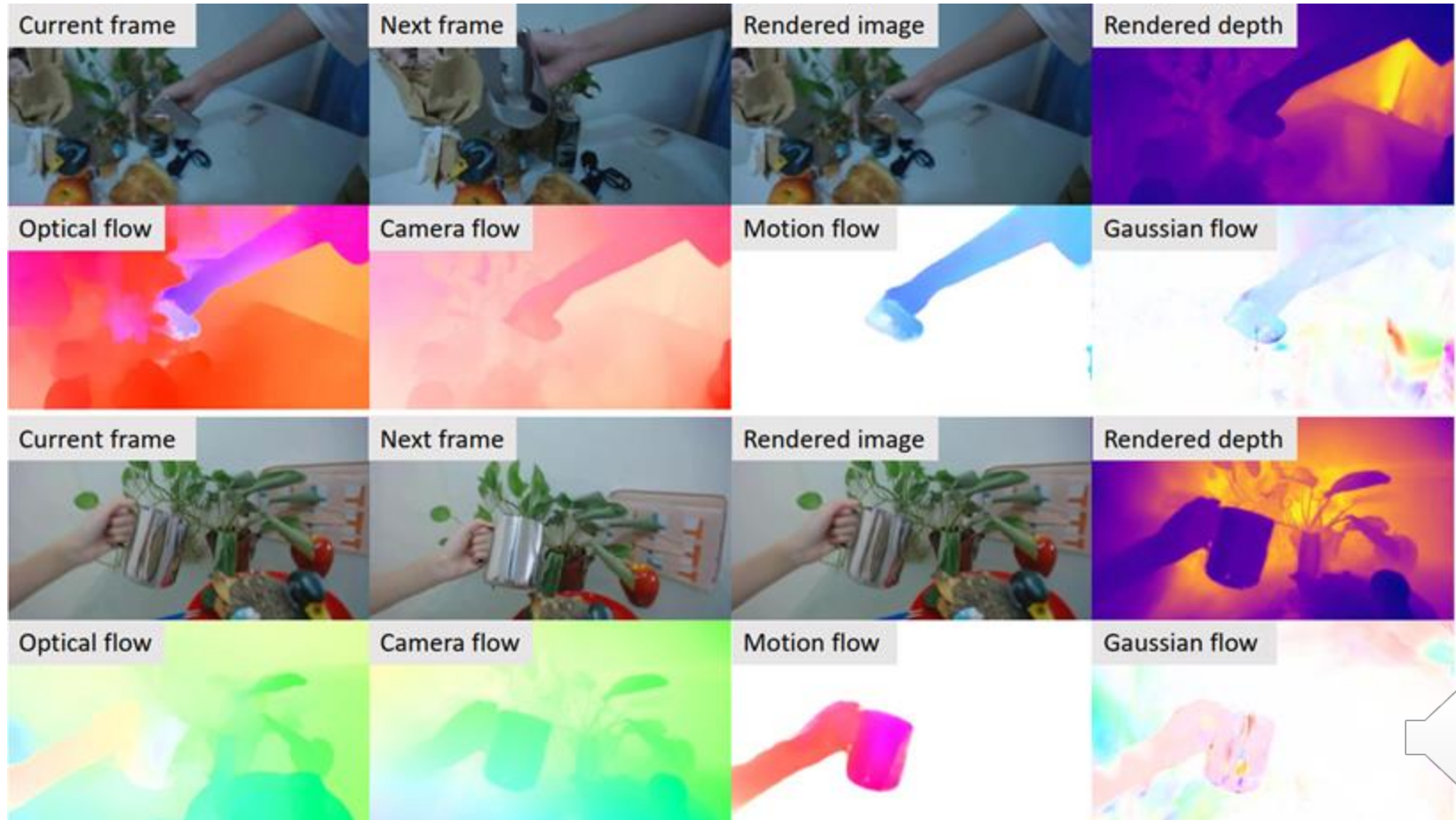
Comparison



HyperNeRF dataset



Visualization



Visualization



Basin



As



Cup



Sieve



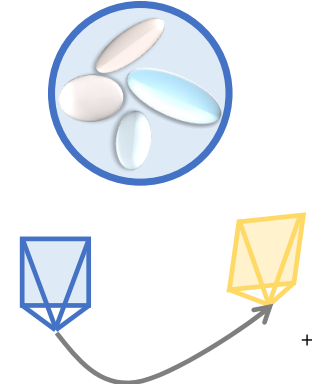
Press



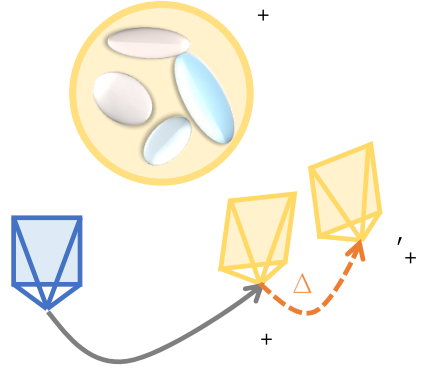
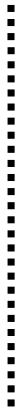
Bell



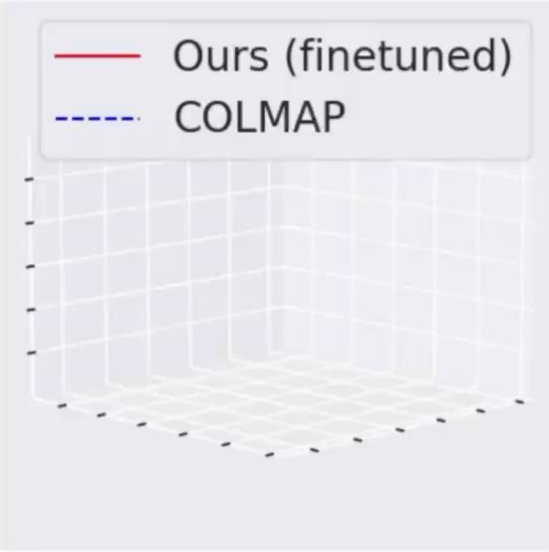
Visualization



Training Gaussians
(Camera Pose Fixed)



Finetuning Camera Pose
(Gaussians Fixed)



Thanks!

https://ruijiezhu94.github.io/MotionGS_page/

