OnlineSplatter: Pose-Free Online 3D Reconstruction for Free-Moving Objects

Mark He Huang

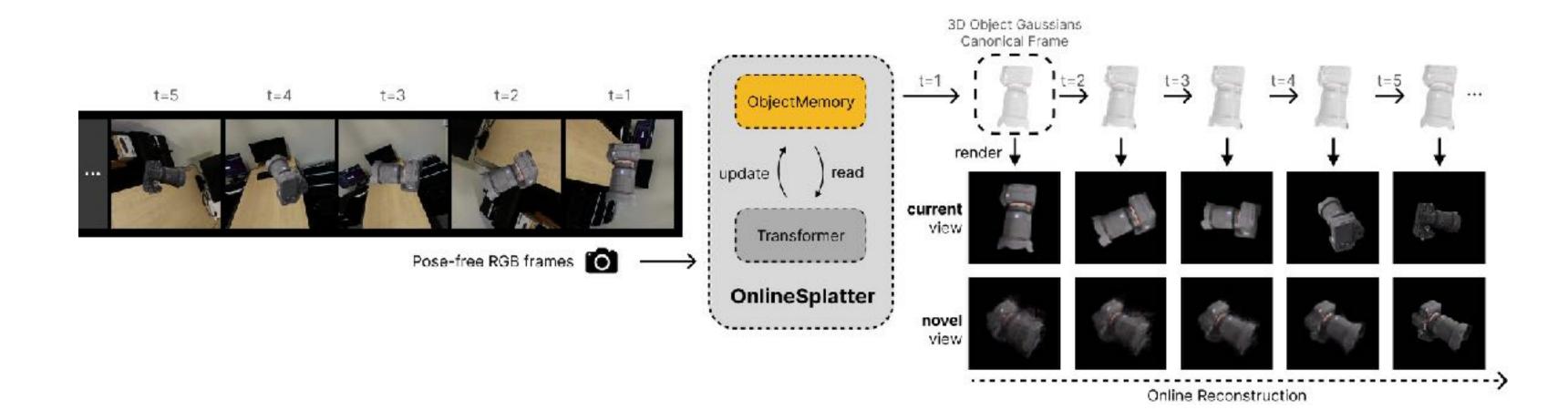
Lin Geng Foo

Christian Theobalt

Ying Sun

De Wen Soh

NeurIPS 2025 (Spotlight)









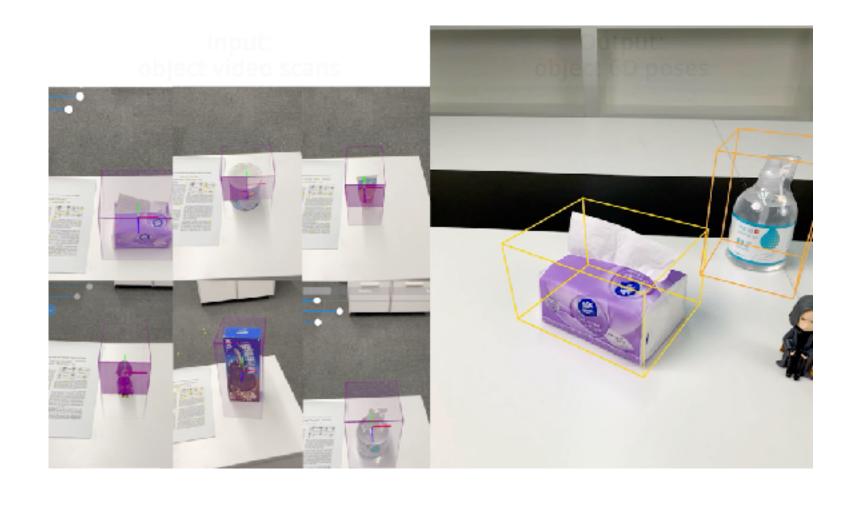


Towards Unconstrained Assumption-Free Object Perception

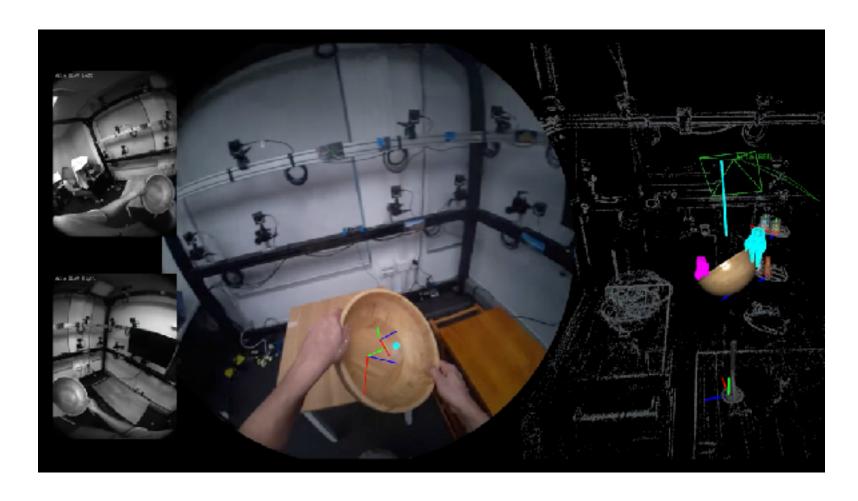
Controlled Laboratory



Constrained Real-world



Unconstrained Real-world



Static Scene

Static Camera, Known Pose

Static Scene

Moving Camera, Known SfM Pose

Dynamic Scene

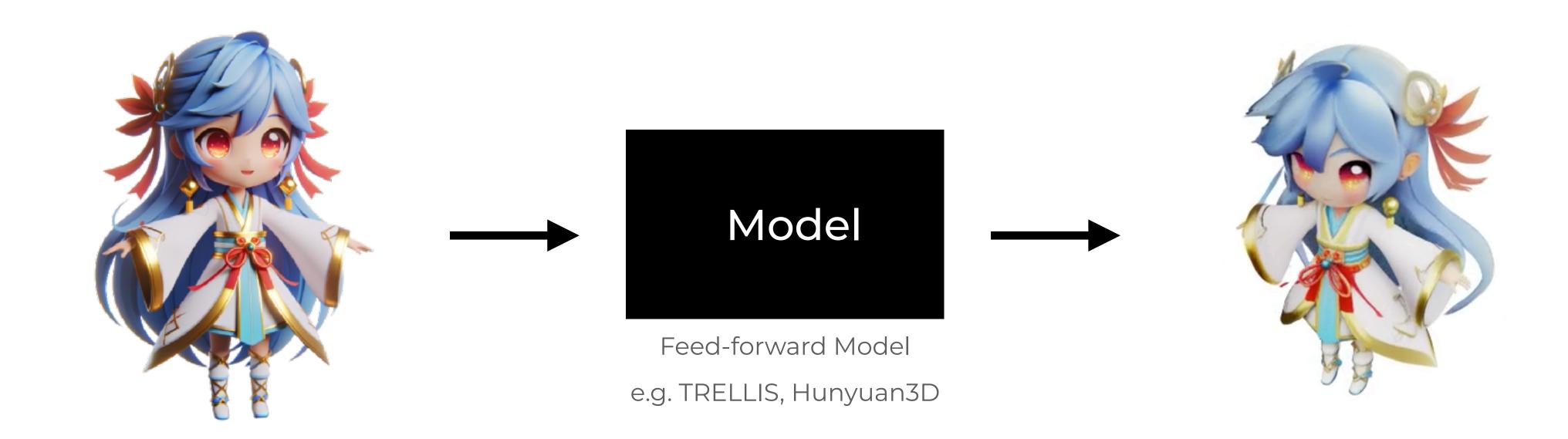
Moving Camera, Unknown Pose

Ego-centric Captures

Autonomous Robots Perception



Is Large Generative Model the Solution to Object Perception?



Single Image

3D Geometry & Texture

Enabled by large-scale object-level priors & use Hallucination as a feature



Is Large Generative Model the Solution to Object Perception?

Enabled by Large-scale trained priors & use Hallucination as a feature

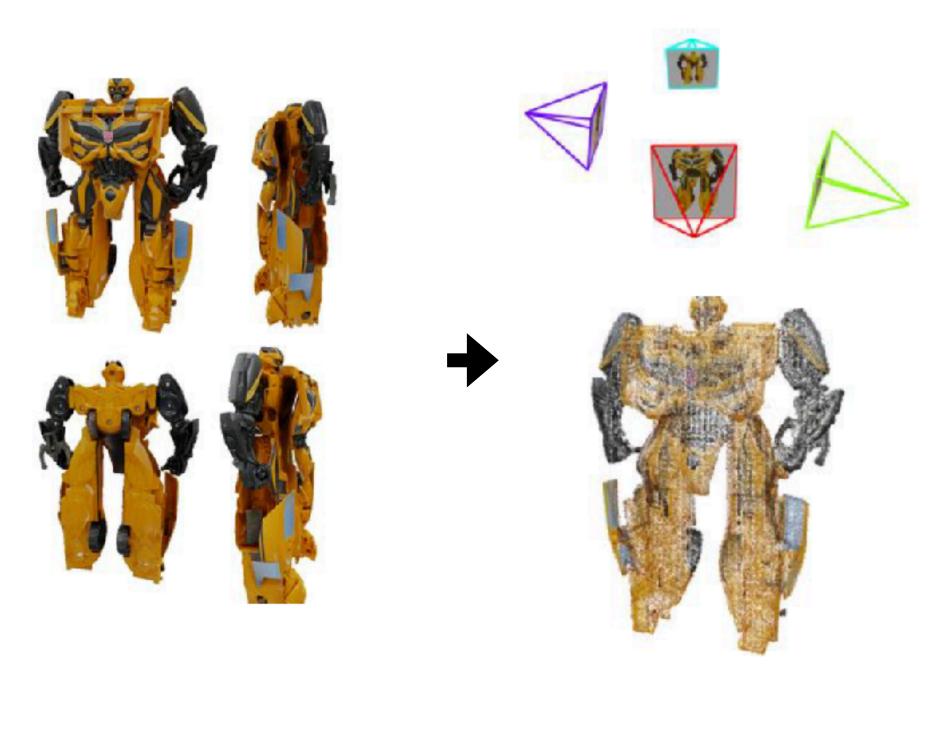
- Great for 3D asset creation, but not for real-time perception
- Performs poorly for out-of-domain objects
- Not applicable to real-time agents in dynamic environments

How might we enable object perception for autonomous agents?

How might we perceive physical objects naturally and continuously?

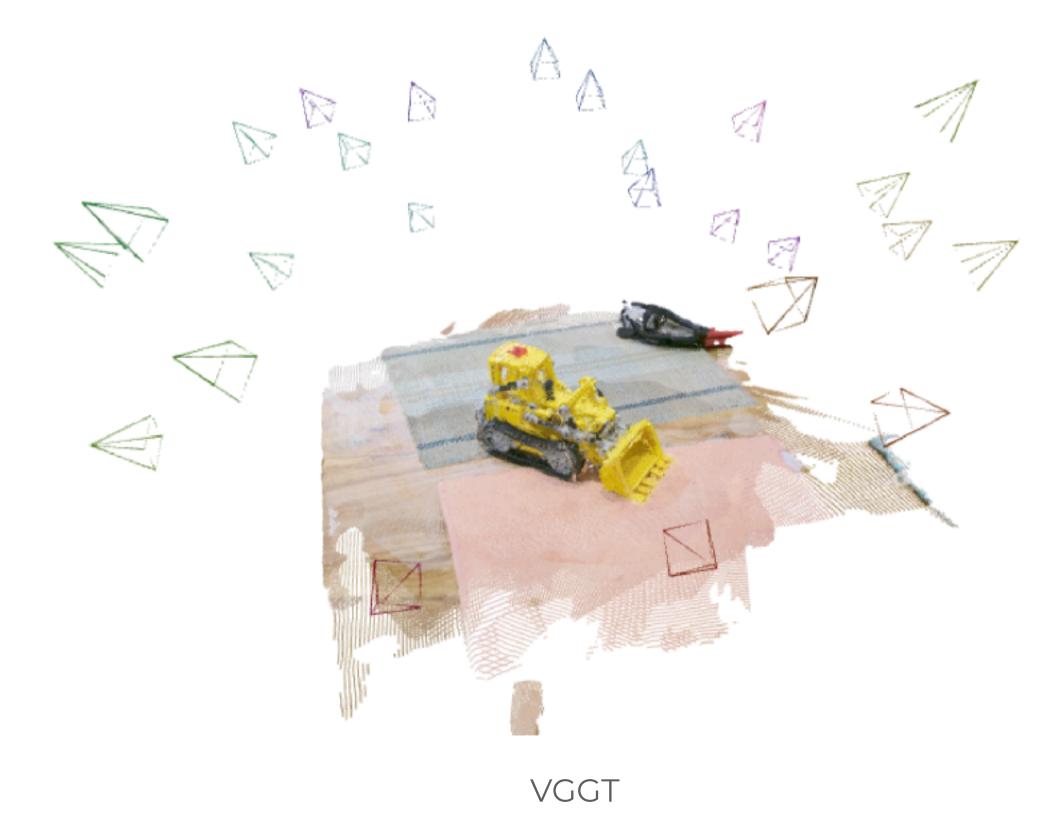


Recent Advances in Feed-forward RGB-only Reconstruction



FreeSplatter

Pixel-align Gaussian Prediction



Pixel-align Points Prediction

Enabled by large-scale learned capability to match and reason in pixel & 3D space



Our Work

Pose-Free RGB Online 3D Reconstruction for Free-Moving Objects

No Camera Pose

Removing the requirement for using camera poses as input

No Depth Sensor

Towards using only monocular RGB frames as inputs

No Pre-capturing

Towards using video stream to reconstruct objects on-the-fly

No Static Assumption | Alleviate the constraints of object placement & movements

Eliminating constraints & assumptions

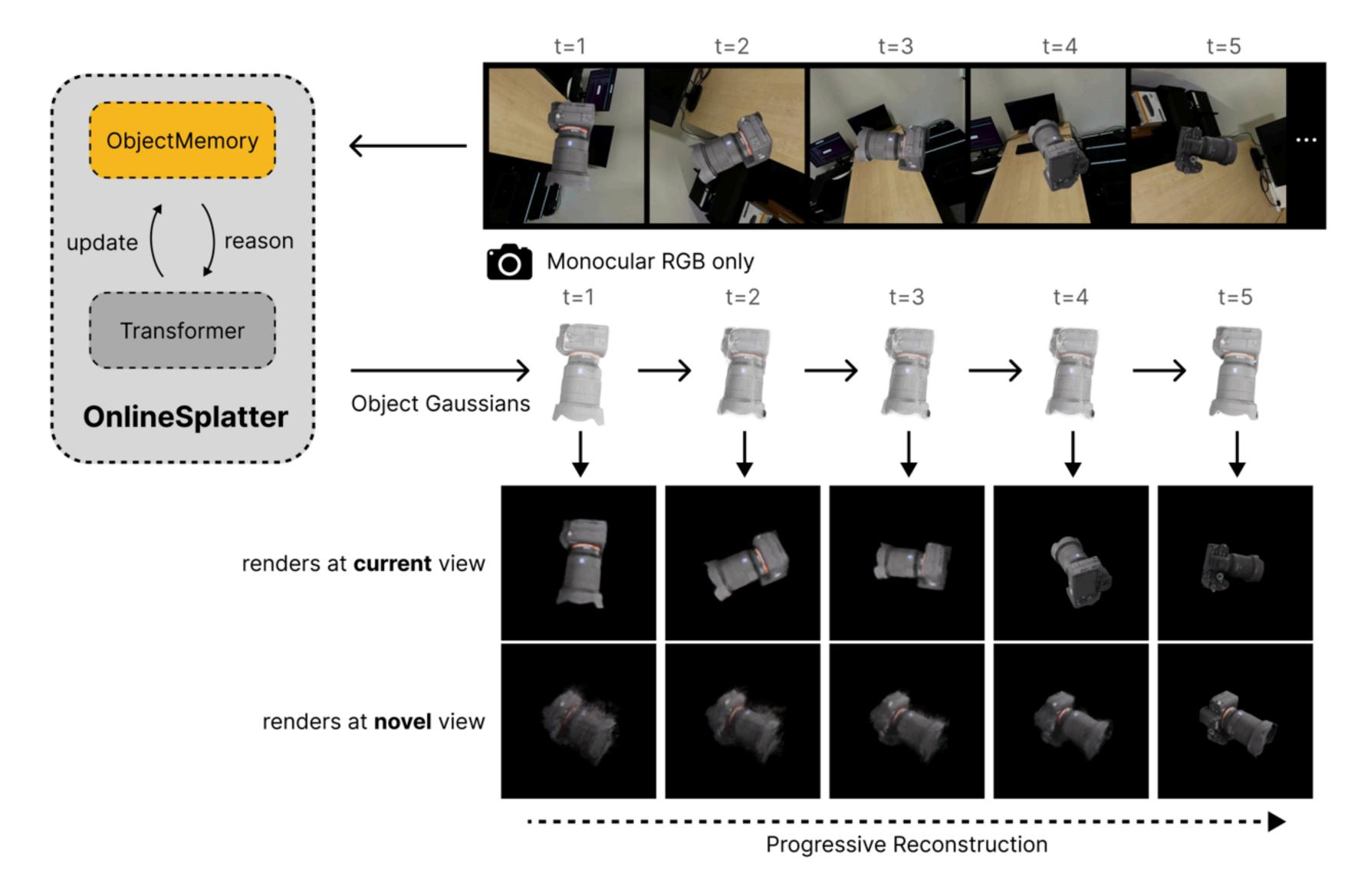
Unlock data in-the-wild

Bring offline reconstruction to online

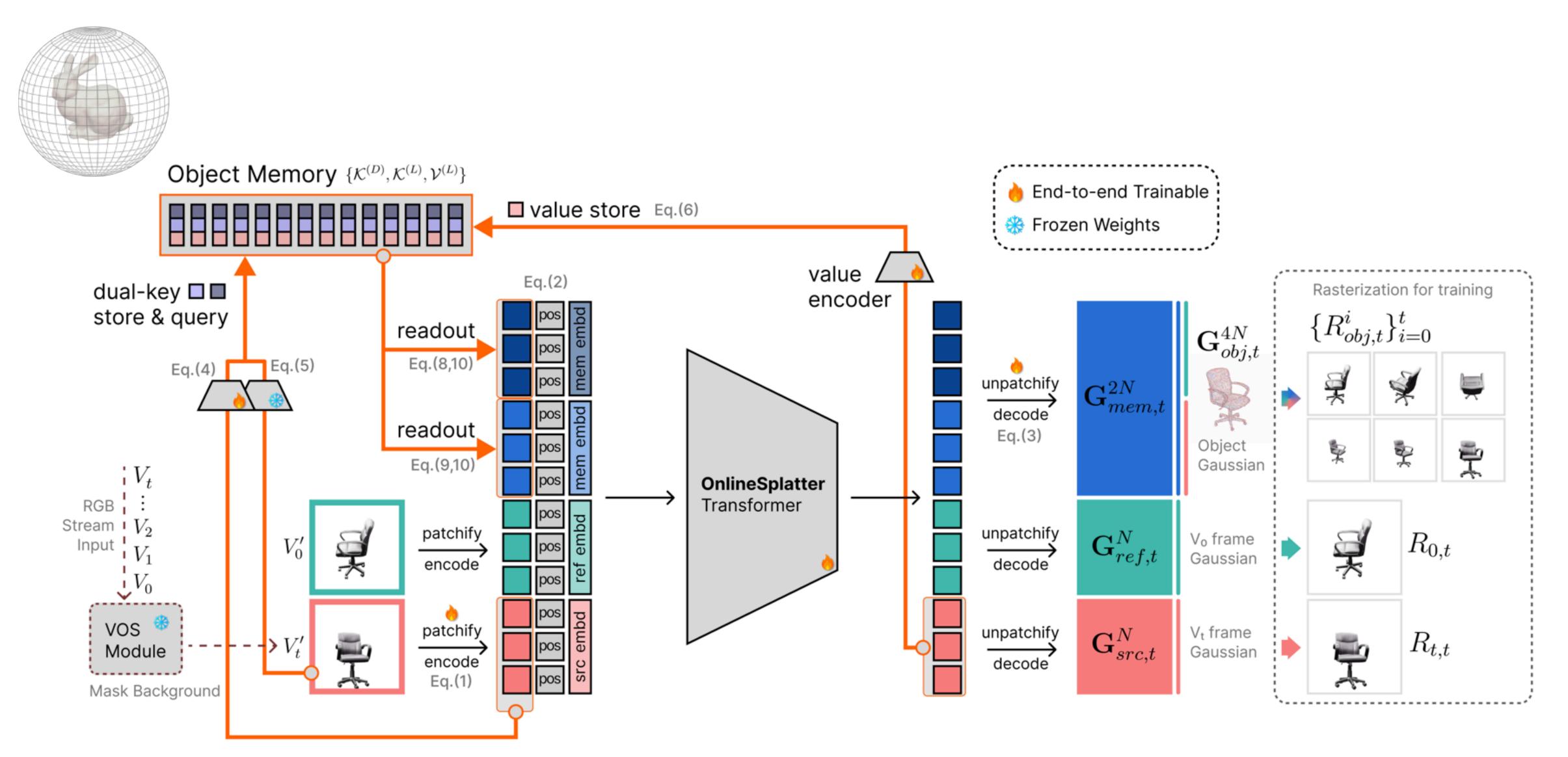
Unlock real-time experience



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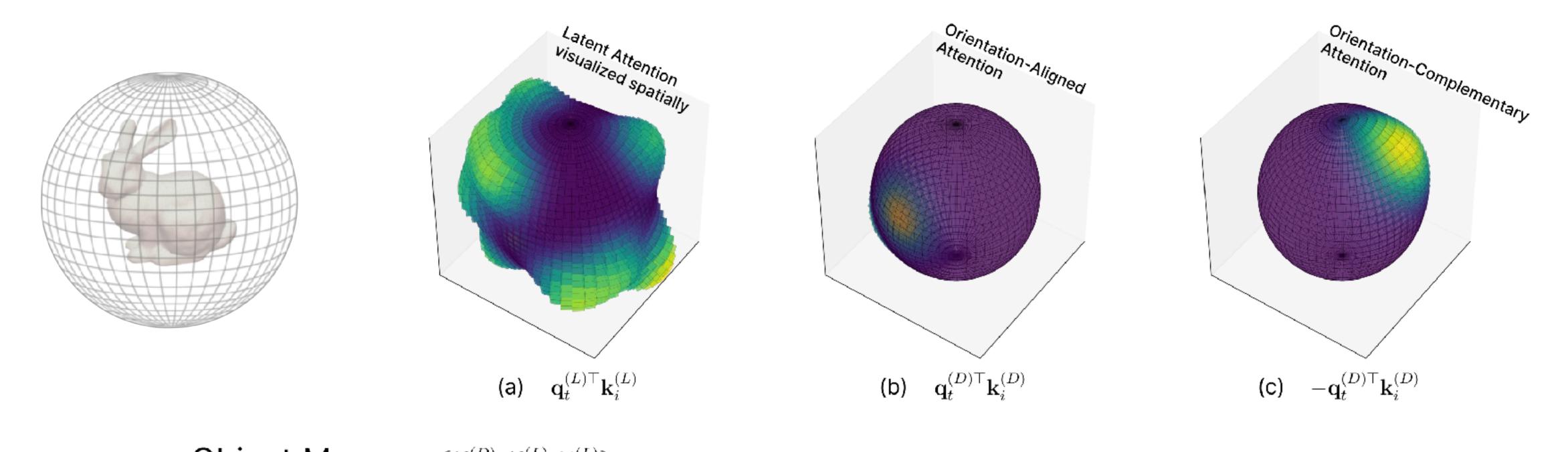
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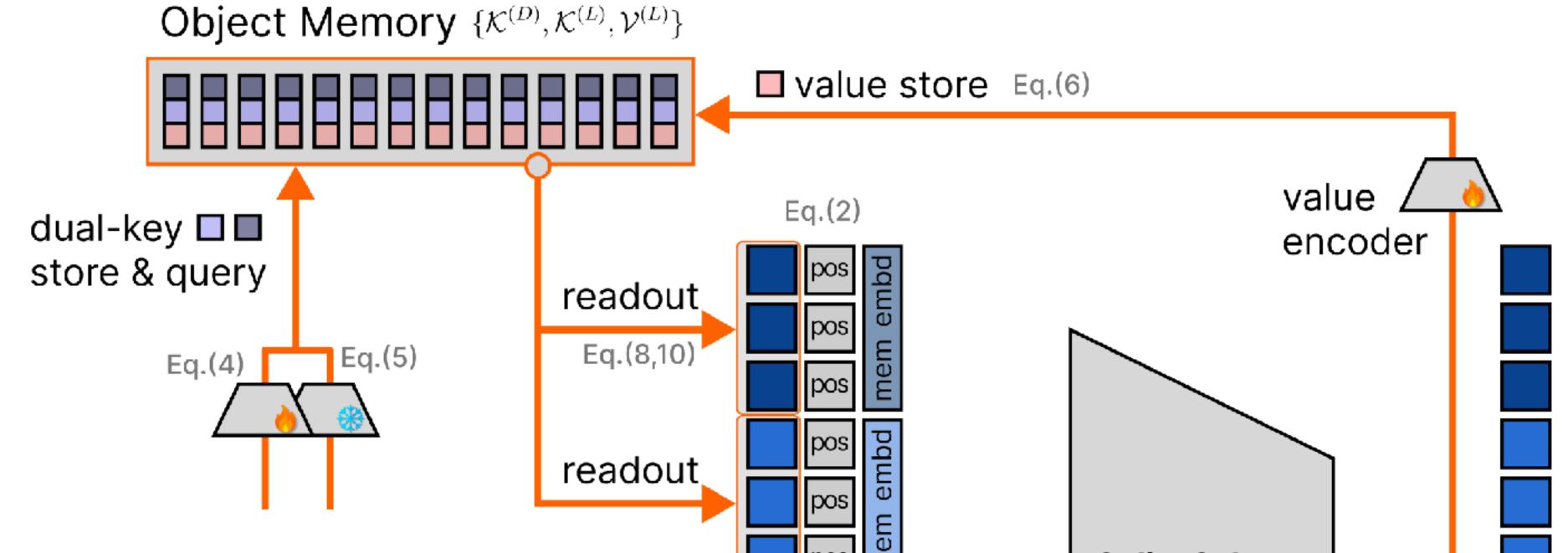






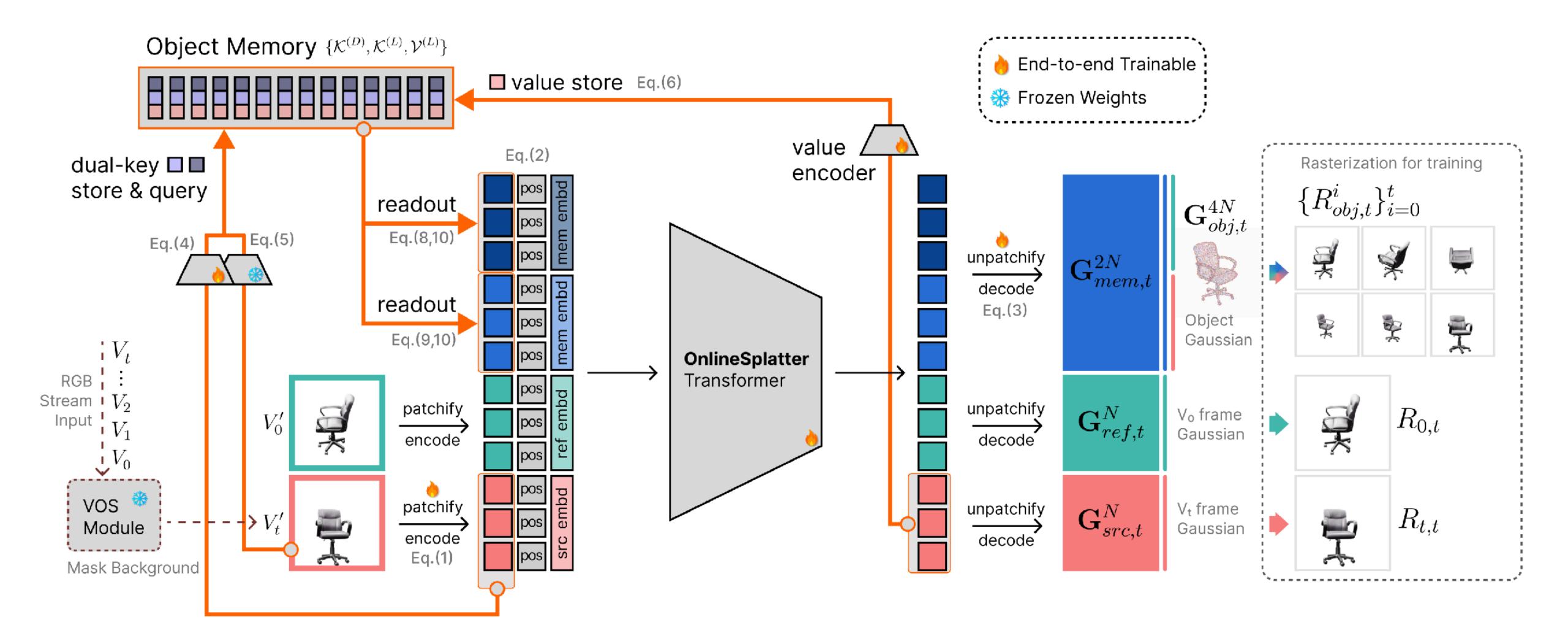
Dual-Key Spatial Object Memory Insertion and Readout







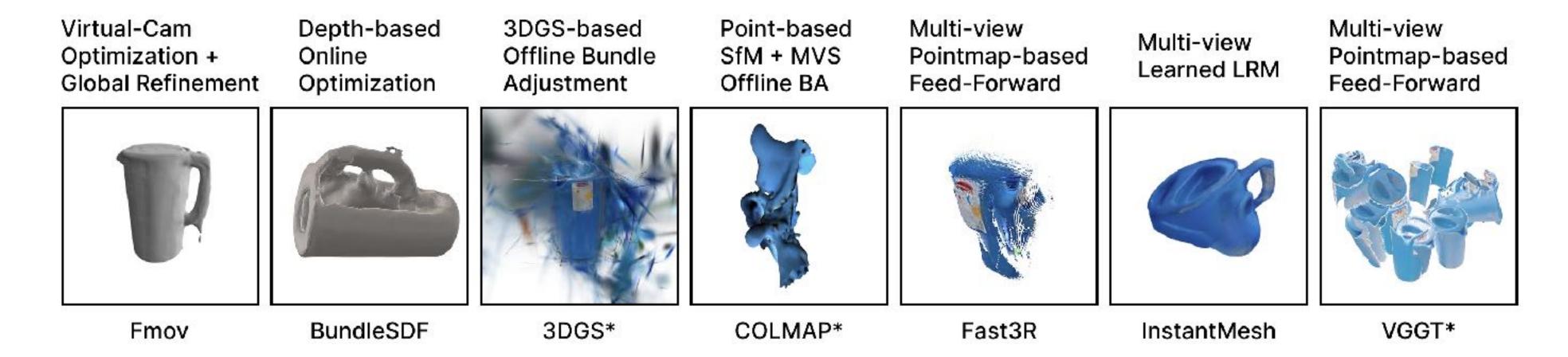
Dual-Key Spatial Object Memory Insertion and Readout





It is a Pretty Difficult Task

Many classical and latest methods fails



Free-Moving Object Reconstruction and Pose Estimation with Virtual Camera (AAAI 2025)

BundleSDF: Neural 6-DoF Tracking and 3D Reconstruction of Unknown Objects (CVPR 2023)

3D Gaussian Splatting for Real-Time Radiance Field Rendering (SIGGRAPH 2023)

Structure-from-Motion Revisited (CVPR 2016)

Fast3R: Towards 3D Reconstruction of 1000+ Images in One Forward Pass (CVPR 2025)

InstantMesh: Efficient 3D Mesh Generation from a Single Image with Sparse-view Large Reconstruction Models (2024)

VGGT: Visual Geometry Grounded Transformer (CVPR 2025)

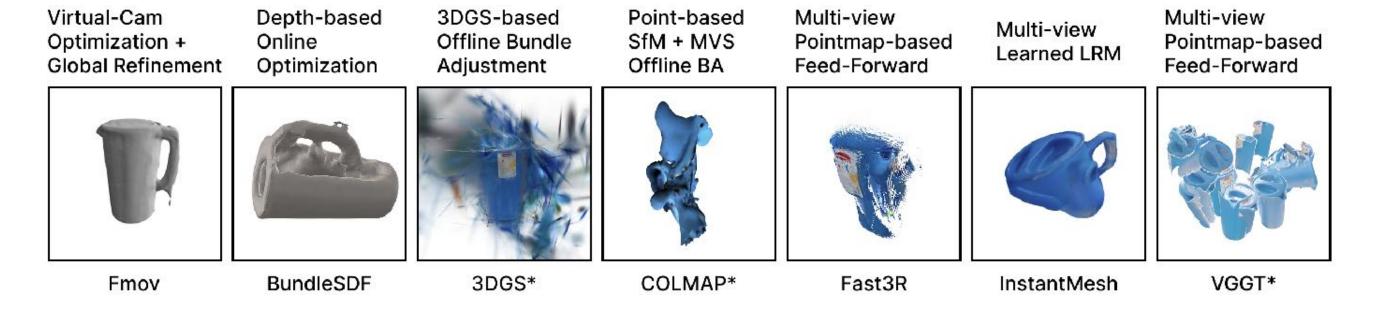


Quantitative and Qualitative Results

GSO										
Method	Early-Stage			Mid-Stage			Late-Stage			
	PSNR↑	SSIM↑	LPIPS↓	PSNR↑	SSIM↑	LPIPS↓	PSNR↑	SSIM↑	LPIPS↓	
FSO _{rand4}	21.358	0.861	0.177	21.919	0.877	0.181	21.737	0.855	0.181	
FSO _{dist4}	22.365	0.874	0.119	<u>23.757</u>	0.862	0.117	23.751	0.873	0.120	
NPS _{dist2}	22.986	0.859	0.155	23.050	0.863	0.162	22.949	0.878	0.156	
NPS_{dist3}	<u>23.331</u>	0.862	0.149	23.206	0.861	0.138	<u>24.141</u>	0.863	0.125	
Ours	26.329	0.921	0.084	27.553	0.933	0.066	31.737	0.969	0.075	

HO3D											
Method	Early-Stage			Mid-Stage			Late-Stage				
	PSNR↑	SSIM↑	LPIPS↓	PSNR↑	SSIM↑	LPIPS↓	PSNR↑	SSIM↑	LPIPS↓		
FSO _{rand4}	18.488	0.820	0.187	18.552	0.817	0.191	17.683	0.810	0.199		
FSO_{dist4}	18.594	0.837	0.177	19.215	0.848	0.184	19.619	0.843	0.183		
NPS_{dist2}	21.063	0.855	0.160	22.803	0.841	0.158	22.134	0.846	0.164		
NPS _{dist3}	<u>21.134</u>	0.853	0.162	22.967	0.869	0.165	22.947	<u>0.860</u>	<u>0.163</u>		
Ours	23.627	0.910	0.152	25.803	0.912	0.122	27.928	0.952	0.099		

Table 1: Comparison of different baselines on two datasets. Results are shown for early-stage mid-stage, and late-stage settings. Best results are **bolded** and second best results are <u>underlined</u>.



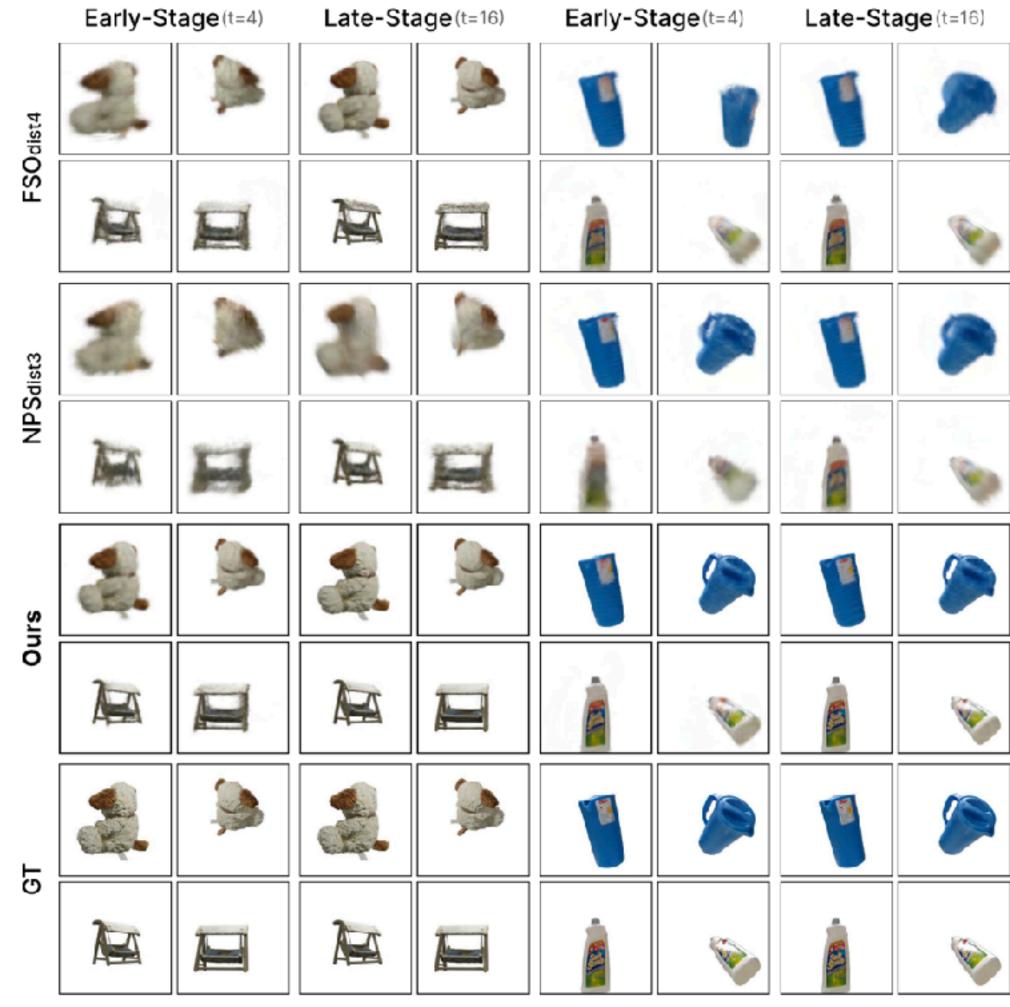


Figure 3: Qualitative results of different baselines and our method on the GSO (left) and HO3D (right) datasets. We visualize the results at inference timestep t=4 and t=16, which corresponds to the early-stage and late-stage settings, respectively. Our reconstructed outputs shows significant better visual quality and geometric accuracy as more observations become available.



Future Works

- ·Feedforward Model + Real-time Lightweight Optimization
- Pose-free Training
- Non-rigid Object On-the-fly Reconstruction
- Hybrid Object Representation



Concurrent Works Exploring Stateful Continuous Reconstruction

- •CUT3R: Continuous 3D Perception Model with Persistent State
- •TTT3R: 3D Reconstruction as Test-Time Training



markhh.com/OnlineSplatter

