

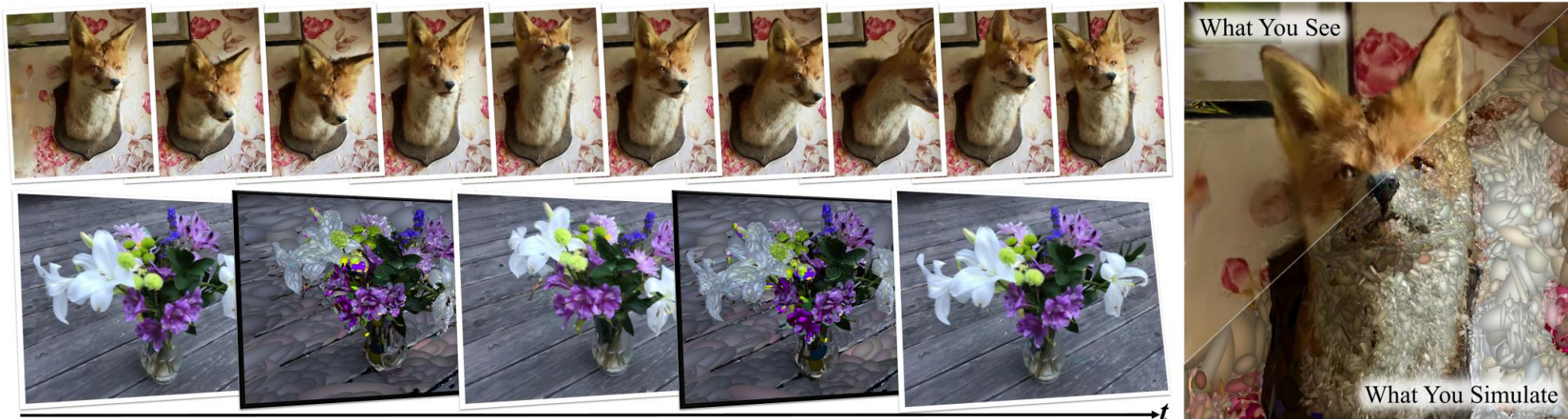
DecoupledGaussian: Object-Scene Decoupling for Physics-Based Interaction

Presenter: Miaowei Wang

Motivation: Simulation from Reality

PhysGaussian: start the trend of simulation captured from real scenes

- Introduce GS Kinematics: rotate covariance matrix, spherical harmonic coefficients
- Insert Gaussian Splatting (GS) into the Material Point Method(MPM) simulation engine.



[Image Source: PhysGaussian CVPR2024 Highlight]

Motivation: Simulation from Reality

VR-GS deforms semantically segmented objects by SAM through XPBD simulation engine

Object-level 3D Scene Reconstruction



Real Scene Capture

COLMAP
Calibration

Image
Segmentation

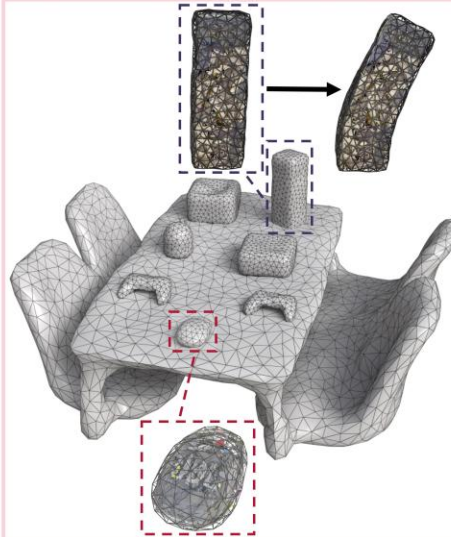


3D Gaussian Splatting

Segmentation

Inpainting

GS Embedded Geometry Reconstruction



Mesh Reconstruction

VDB Reconstruction

Tet Generation

Two-level Embedding

Local Embedding

Global Embedding

VR-GS Simulation and Rendering



Dynamics and Illuminations

Extended Position-based Dynamics

Collision Handling

Gaussian Rasterizer with Shadow Ray

[Image Source: VR-GS Siggraph 2024 Conference]

Limitation: Object Depart from Surface

Current methods fail to simulate object motion departing from the contact surface.



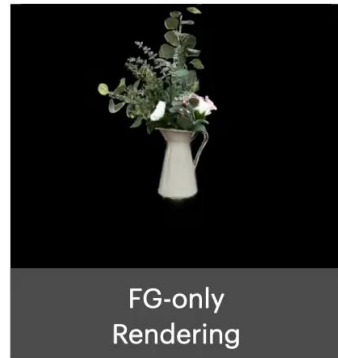
PhysGaussian



VR-GS



FeatureSplatting (ECCV2024, Language-driven)



FG-only
Rendering

The core issue is the failure to recover occluded geometry and texture.

Our task: Decoupling



Input Video

Decoupling

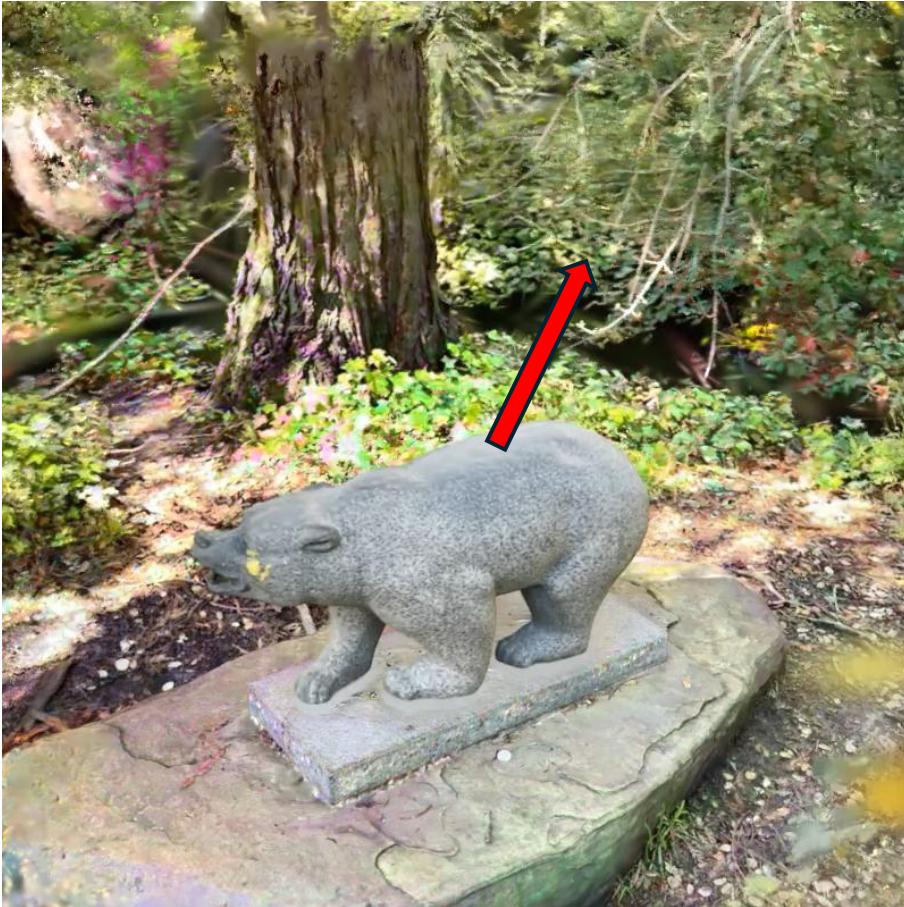


Scene



Object

Support Various Interactive Simulation



Scene Collisions



Object Melting

DecoupledGaussian: Meaning and Application



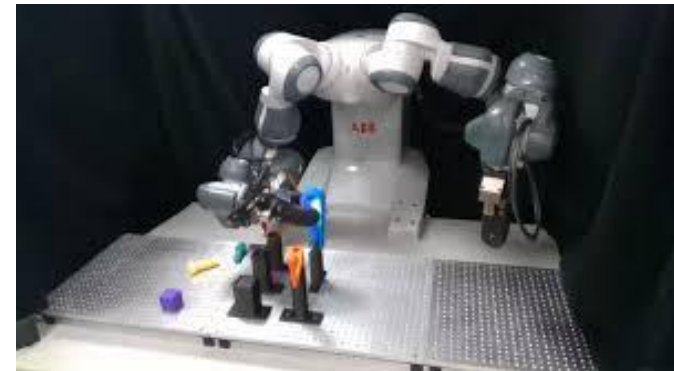
Autonomous driving Simulation



Entertainment industry



VR/AR Interaction



Robotic Learning Interaction

Our pipeline

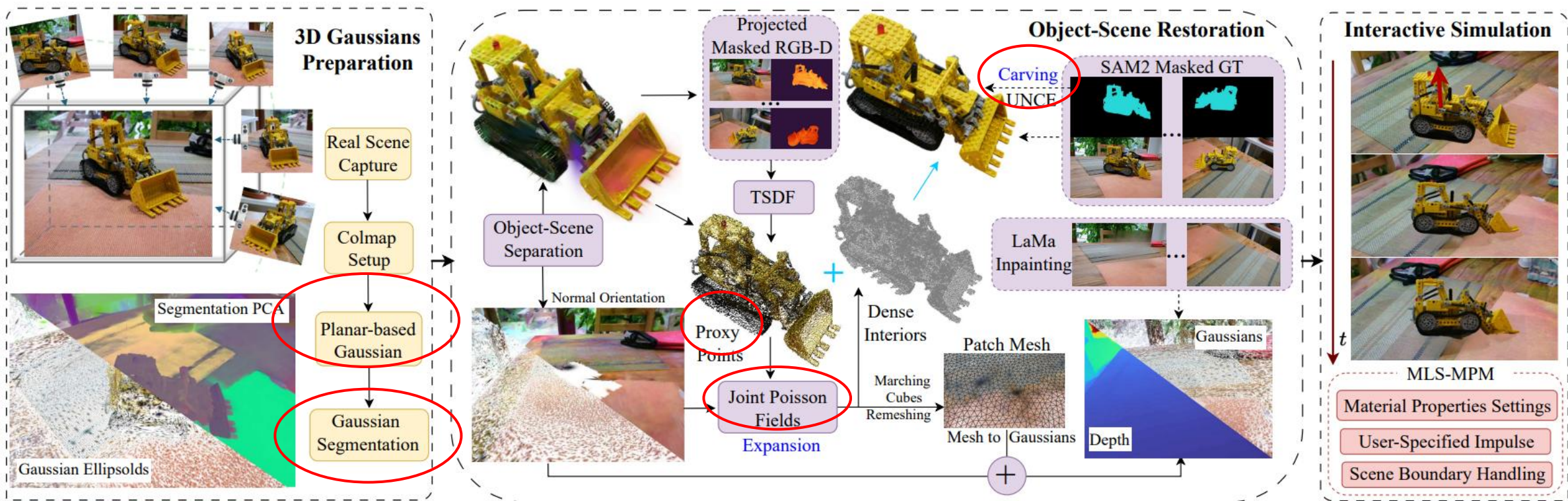


Figure 3. **System Overview.** DecoupledGaussian is an interactive simulation system that enables objects to detach from their initial contact surfaces after applying our proposed restoration pipeline, driven by user-specified impulses (red arrow on the right).

3D Gaussian Preparation: PGSR

The goal of using PGSR:

- 1) estimate correct geometry from GS
- 2) remove the large floaters in the scene

Planar-based Gaussian Splatting (PGSR):

- Flattening the ellipsoid
- Unbiased depth estimation
- Single/Multi-view geometry regularization
- Exposure compensation

w/o Planar-based GS



Ours

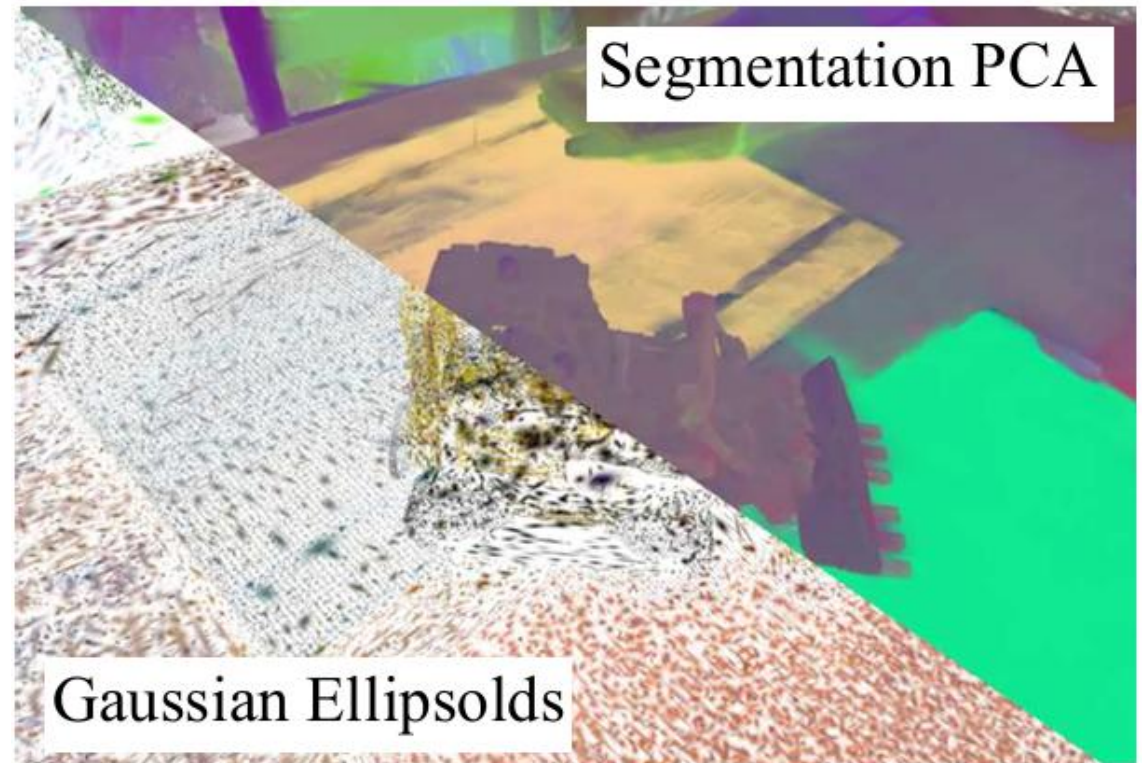


Opacity is set to 1, with $\times 0.4$ scaling for better Gaussian kernel visualization.

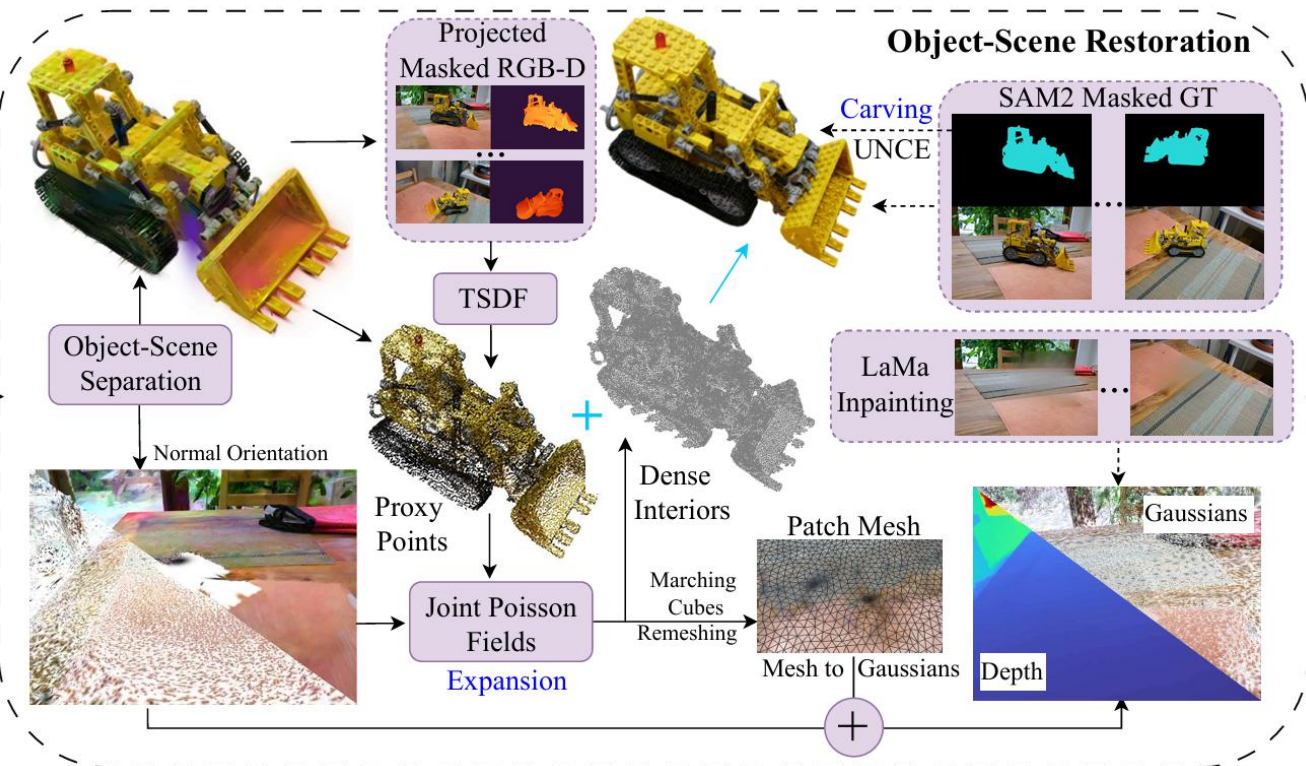
[PGSR: Planar-based Gaussian Splatting for Efficient and High-Fidelity Surface Reconstruction, TVCG2024]

3D Gaussian Preparation: Segmentation

1. Select 2D segmentation labels from SAM2
2. Manually selecting classes in the first frame
3. Each kernel is assigned semantic features
3. A gating network to predict segmentation
4. Local feature smoothing.



Object-Scene Restoration



1. User-specified click position to get object (affinity)
2. Contact-surface use KNN to remove nearby artifacts
3. Restore Object and Contact scene
 - Joint Poisson Fields
 - Proxy Points
 - Unilateral Negative Cross Entropy
 - Gaussian Restoration

2D Inpainting Tools



Masked Image

LaMa

Photoroom

Figure 2. Inpainting tools (LaMa [79]; PhotoRoom [1]) introduce artifacts and inconsistent textures across frames.

(1) inpainted regions often fail to blend seamlessly with surrounding geometry, creating artifacts

(2) texture inconsistencies across frames due to the lack of robust video inpainting tools.

Our approach: prioritizing geometry restoration, leveraging *intrinsic GS geometry priors* to ensure a coherent surface even when texture inpainting is imperfect

Joint Poisson Fields

Geometry Prior: assuming both Object and Scene are *smooth, closed* shapes

We resort to Poisson Surface Reconstruction, equivalent to the winding number field

Joint Poisson Fields STEPS

1. Solving Screen Poisson Indicator for object and scene. Points&Normals
2. Coordinates transformation
3. Solving conflicted regions, by curvature and prioritizing surface
4. Extract **dense interior points** for objects while marching cubes to get cropped **mesh patch** for surface

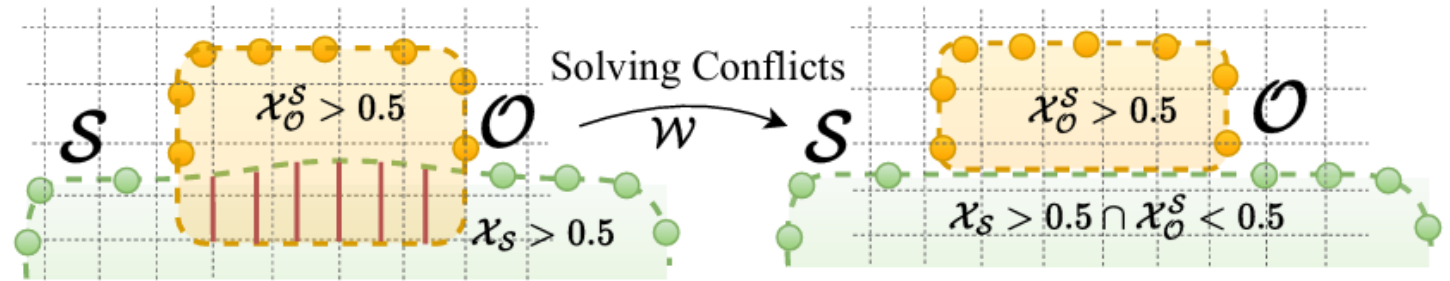


Figure 4. Joint Poisson Fields \mathcal{W} first reconstruct \mathcal{O} and \mathcal{S} independently, then resolve conflicts (red area) by defining a boundary that separates them into distinct, non-intersecting entities.

[Adaptive multigrid solvers]

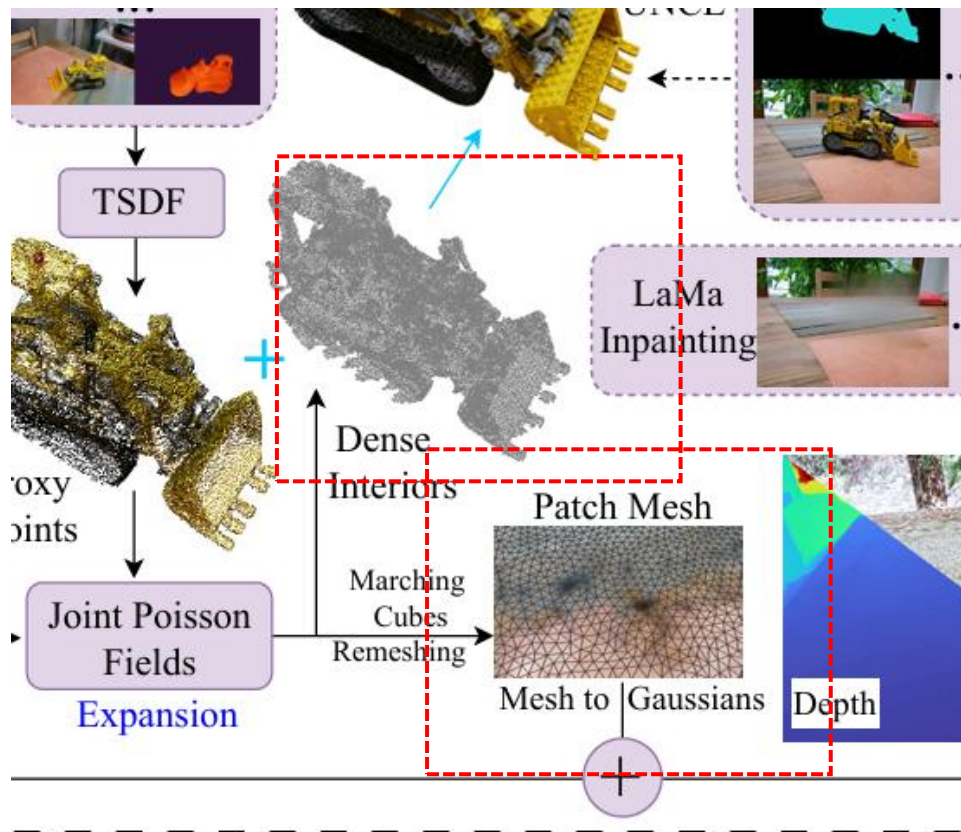
Joint Poisson Fields

Dense Points: helpful for MPM simulation

Patch Mesh: fix the broken geometry of the Surface

How about the input to Joint Poisson Fields:

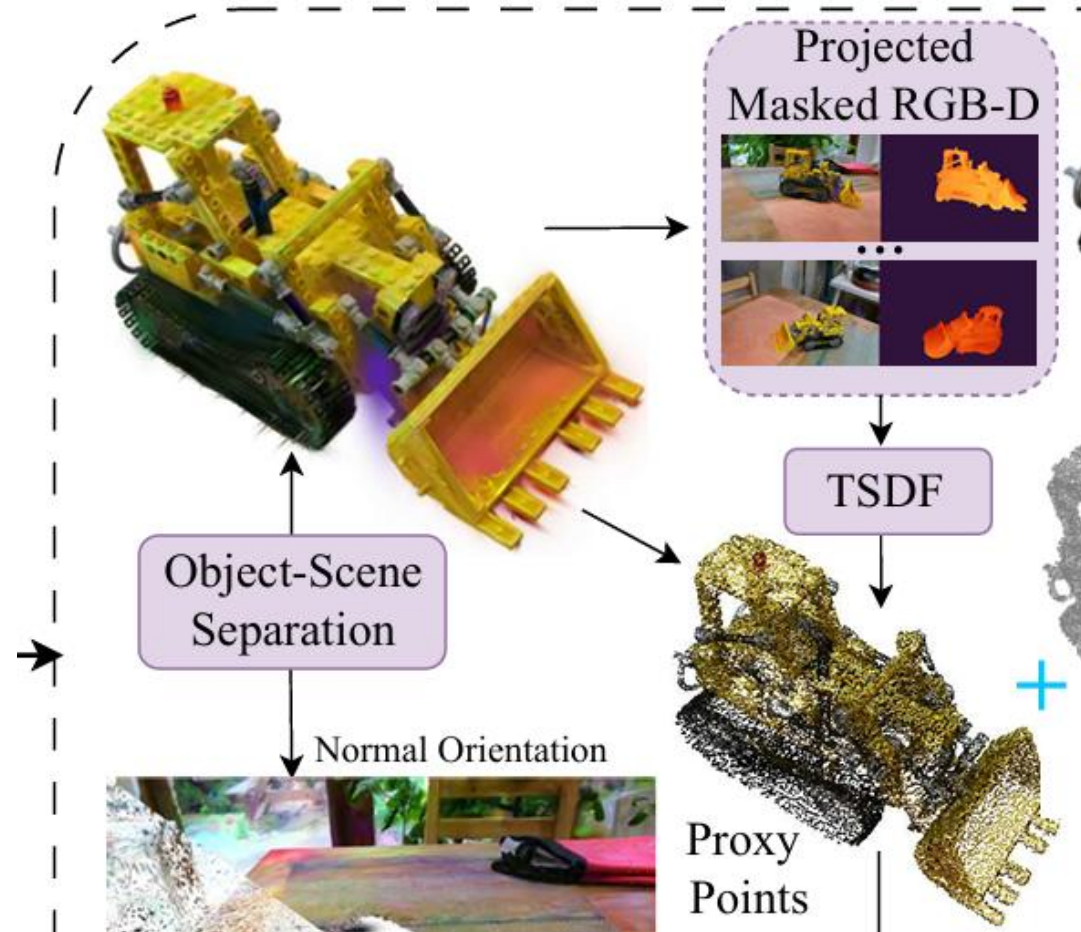
- Contact Surface
 - Gaussian Centers as points
 - Minimum scaling as normal (90-degree disambiguate)
- Target Object
 - Geometric complexity
 - Use Proxy Points instead of Gaussian Centers



Proxy Points: Enhance Geometry Estimations

To get the proxy points, we use TSDF but different from others:

1. Projected depth **mask** to boost fusion speed
Obtain the projected mask by setting zero/one opacity
2. Segment the final proxy points based on raw Gaussian kernel with nearest neighbor search



Proxy Points: Quick qualitative ablation

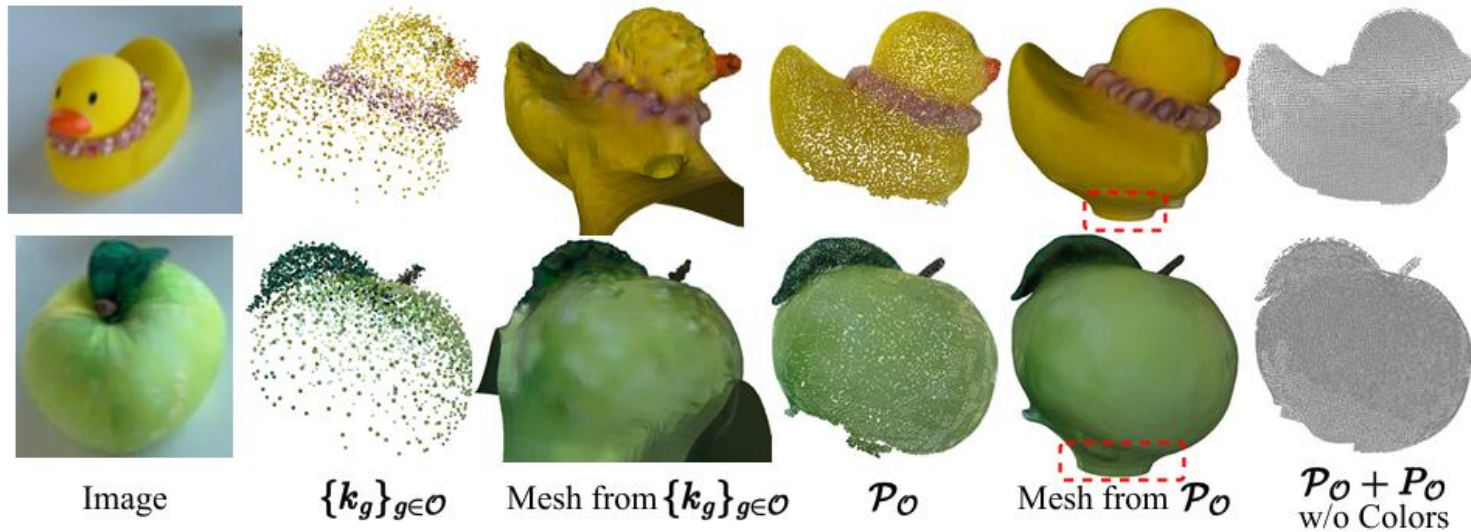


Figure 5. **Ablation for $\mathcal{P}_\mathcal{O}$.** Independent Poisson reconstruction of object \mathcal{O} using Gaussian centers $\{k_g\}_{g \in \mathcal{O}}$ yields poor mesh quality compared to using proxy points $\mathcal{P}_\mathcal{O}$. Our joint Poisson field \mathcal{W} , which integrates the scene surface \mathcal{S} , effectively removes the overextended regions (highlighted in red). The final dense points $P_\mathcal{O}$ are then combined with proxy points $\mathcal{P}_\mathcal{O}$ for Gaussian restoration and continuum simulation.

Unilateral Negative Cross Entropy

- Geometry Expansion from Poisson:
 - Over-smooth close surface
 - Influence by noises or outliers
 - Introducing particles beyond observable viewpoints
 - Points Converted to GS, we will get like:



Opacity is set to one for TRUCK to highlight artifacts

Unilateral Negative Cross Entropy

Multi-view carving:

- Silhouette consistency between rendered and GT
- Discrepancy between blended opacity and 2D GT object mask

$$\text{UNCE}(p) = -(1 - M_{\mathcal{O}}^{\text{GT}}(p)) \log(1 - \mathbb{1}_{\mathcal{O}}(p)).$$

2D GT object mask Blended opacity

- Every 100 iterations, we clean Gaussians with low opacity

w/o UNCE

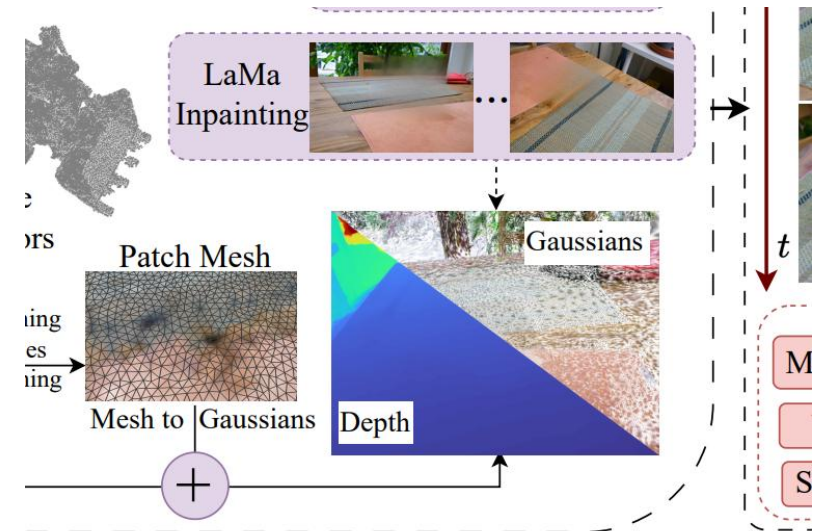


Ours



Gaussian Restoration

- Object, Points to Isometry Gaussian, during finetune:
 - UNCE multi-carve geometry artifacts
 - Fine-tune textures based on GT-masked train images.
- Contact Surface: fixed by patch mesh
 - From mesh to new 3D flattened Gaussian
 - 2D Tools to adjust the texture properties



Interactive Simulation

- User-specified force as an impulse
- Manually specified material properties
- Setting boundary conditions to bounce back
- Estimate Normals of the Ground plane by RANSAC to set up gravity.

.....

Evaluation

Object-Scene Interaction Simulation

Object Restoration

Scene Restoration

Qualitative Evaluation: Object Restoration



Segmented Gaussians



GIC [NIPS 2024]



PhysGaussian [CVPR 2024]



Ours

GIC and PhysGaussian don't account for incomplete object surfaces during simulation, unlike our approach.

Qualitative Evaluation: Scene Restoration



GScream [ECCV 2024]



VR-GS [SIGGRAPH 2024]



Ours

since we don't rely on 2D inpainting tools for geometry restoration, we avoid the catastrophic errors commonly seen in Gscream or VR-GS.

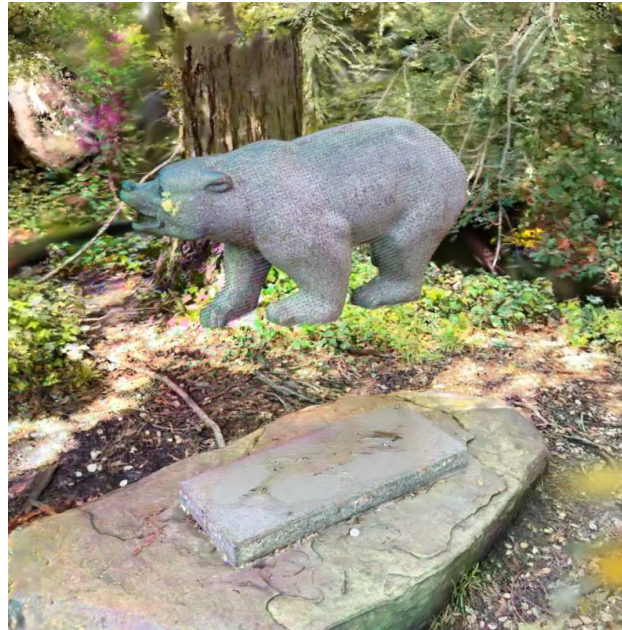
Object-Scene Interactive Simulation



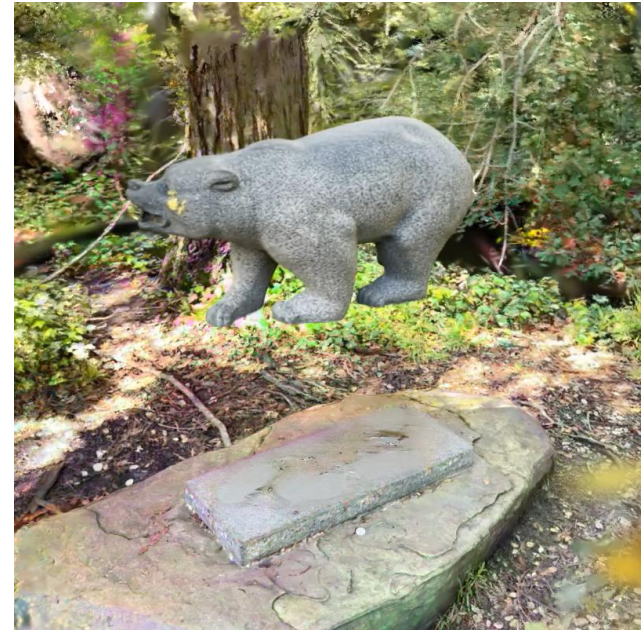
VR-GS (Scene) +
PhysGaussian (Object)



Ours (Scene) +
PhysGaussian (Object)



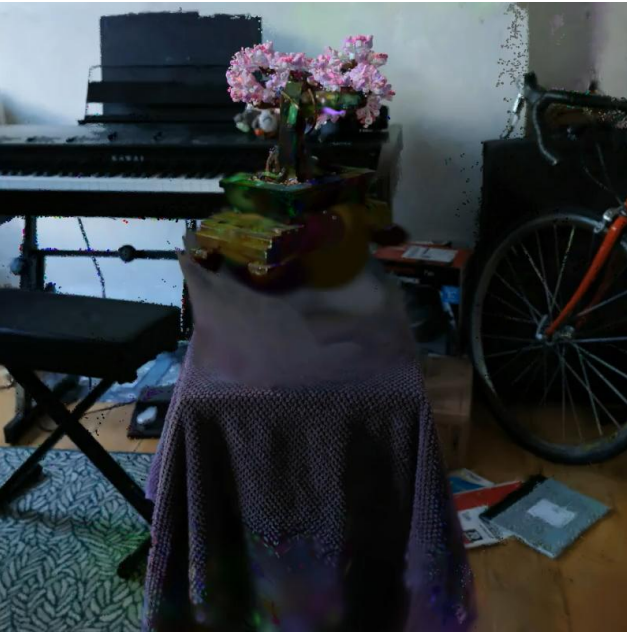
Ours (Scene) +
GIC (Object)



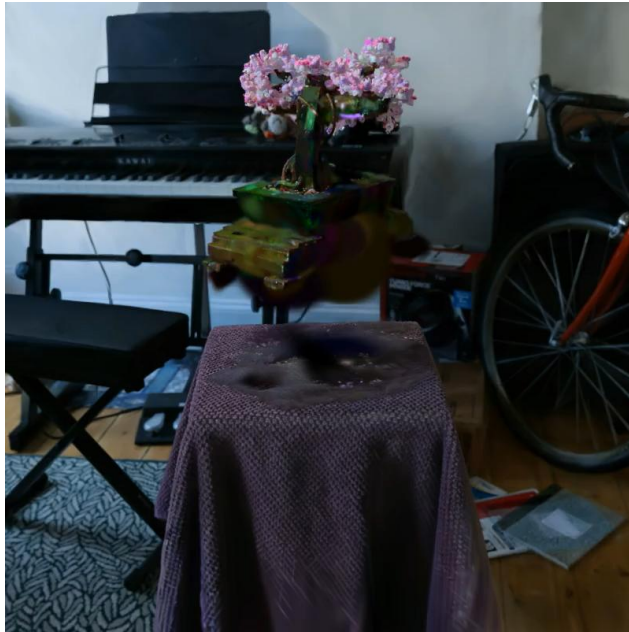
Ours (Scene) +
Ours (Object)

Correct geometry recovery of both the scene and object, as in our method, is crucial for the simulation.

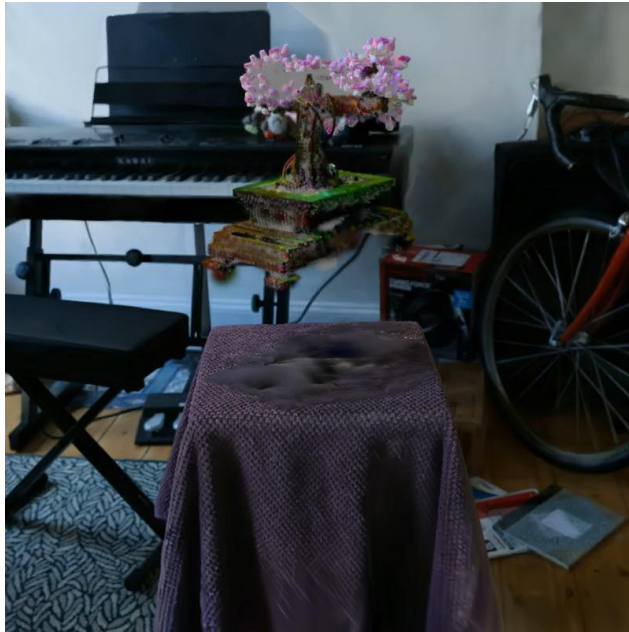
Object-Scene Interactive Simulation



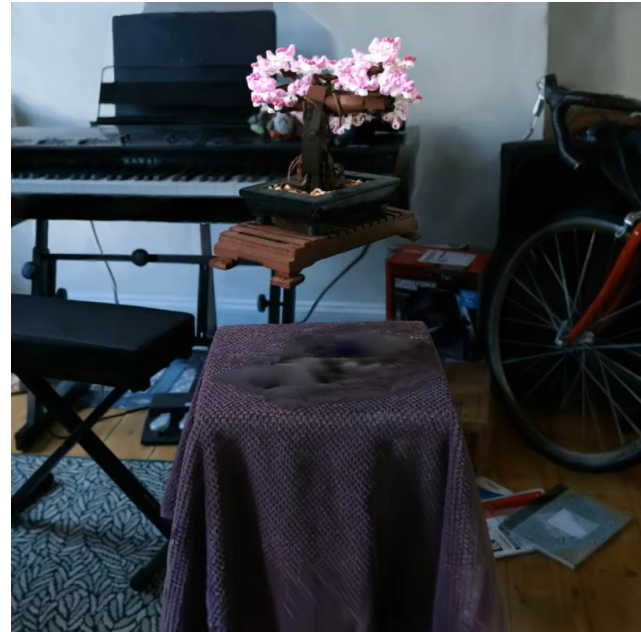
VR-GS (Scene) +
PhysGaussian (Object)



Ours (Scene) +
PhysGaussian (Object)



Ours (Scene) +
GIC (Object)



Ours (Scene) +
Ours (Object)

VR-GS fails to restore the surface of the contact scene, causing the bonsai to get stuck.

Object-Scene Interactive Simulation



VR-GS (Scene) +
PhysGaussian (Object)



Ours (Scene) +
PhysGaussian (Object)



Ours (Scene) +
GIC (Object)



Ours (Scene) +
Ours (Object)

Also, take note of the table in VR-GS. While it appears correct, it still traps the object.

Object-Scene Interactive Simulation

VR-GS (Scene) +
PhysGaussian (Object)

Ours (Scene) +
PhysGaussian (Object)



Ours (Scene) +
GIC (Object)



Ours (Scene) +
Ours (Object)

Of course, our method can also be applied to multiple objects.

Object Restoration



Segmented Gaussians

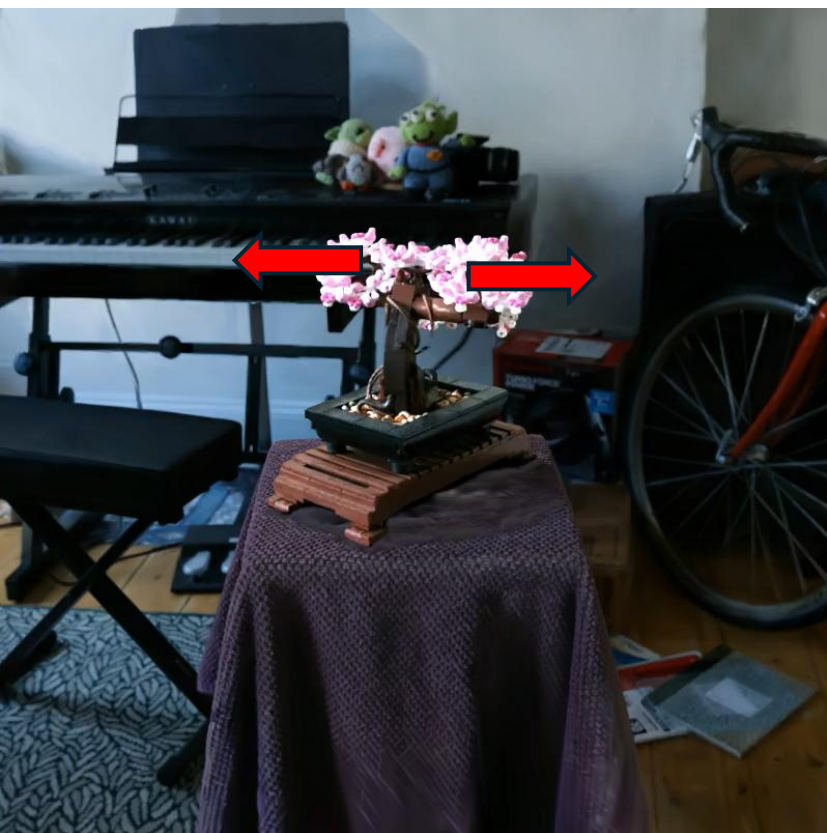
GIC [NIPS 2024]

PhysGaussian [CVPR 2024]

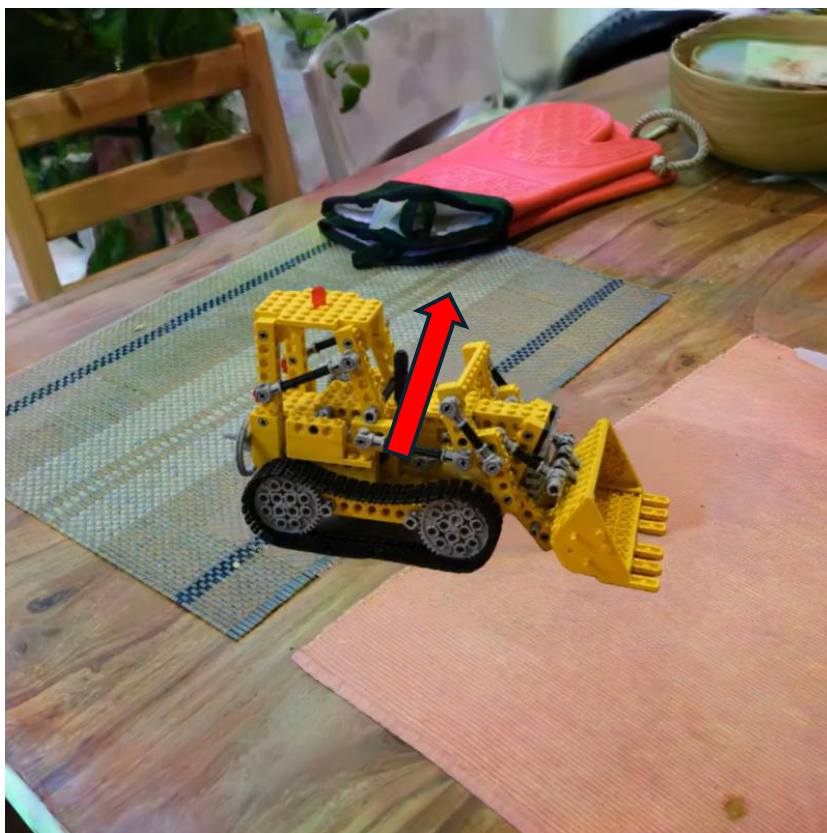
Ours

Our method recovers hidden object regions better than others, with slight issues in consistency and high-frequency textures.

More Cases (Ours)



Bonsai

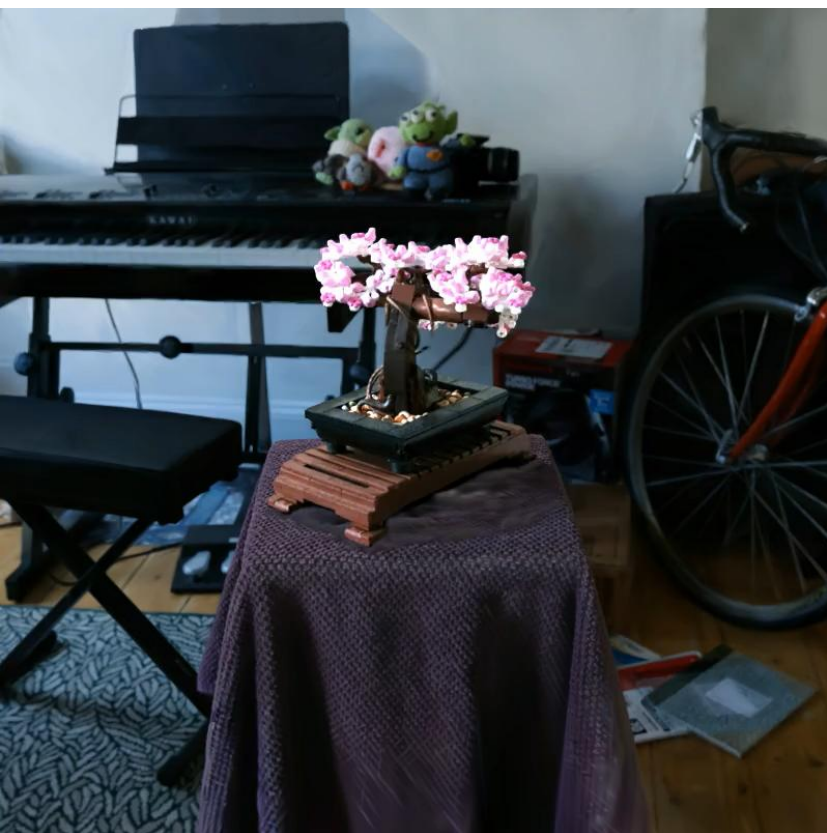


Kitchen



Room

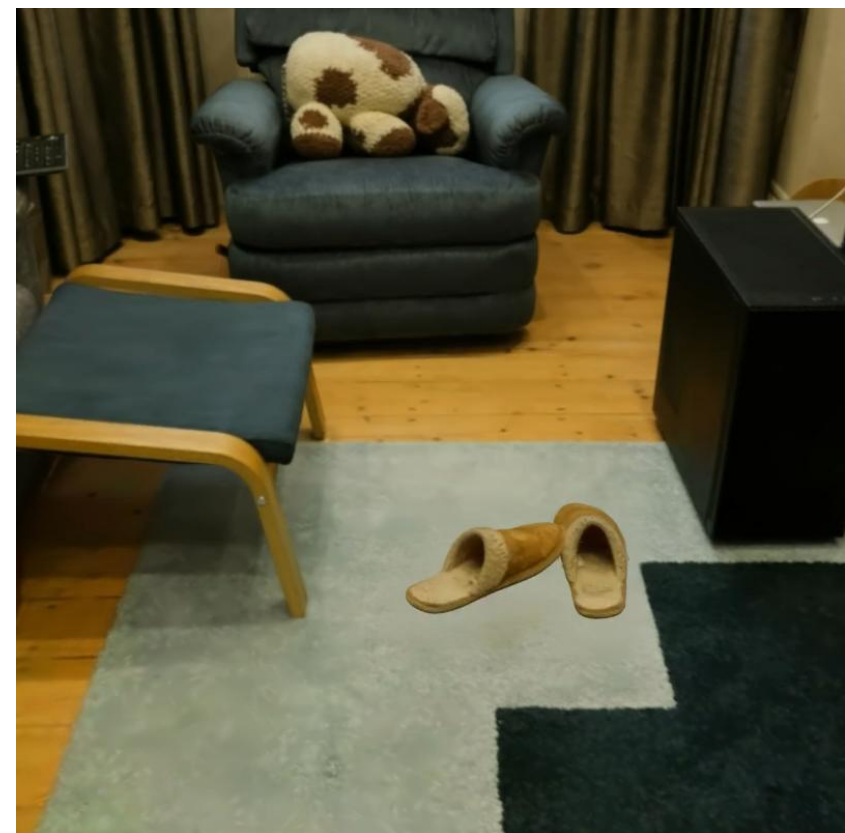
More Cases (Ours)



Bonsai



Kitchen



Room

Interactions Across Diverse Scenes



Input Video



Truck in Bicycle

Quantitative Evaluation

Table 1. **User Study.** Participants rated the fidelity of restoration and interactive simulation in a moving-camera video.

Scene Restoration			Object Restoration	
Methods	SRQ \uparrow	Time \downarrow	Methods	ORQ \uparrow
GScram [83]	1.94	$\sim 70\text{m}$	PhysGaussian [89]	1.40
VR-GS [38]	2.12	$\sim 7\text{m}$	GIC [6]	1.60
Ours	3.48	$\sim 1\text{m}$	Ours	4.03

Object-Scene Interactive Simulation	
Methods	ISF \uparrow
VR-GS(\mathcal{S}) + PhysGaussian(\mathcal{O})	1.50
Ours(\mathcal{S}) + PhysGaussian(\mathcal{O})	2.60
Ours(\mathcal{S}) + GIC(\mathcal{O})	2.73
Ours(\mathcal{S}) + Ours(\mathcal{O})	4.35

SQR: Scene Restoration Quality

OQR: Object Restoration Quality

ISF: Interactive Simulation Fidelity

Quantitative Evaluation

Composite complete
real scene and real
object

Placed by Rigid
simulation PyBullet

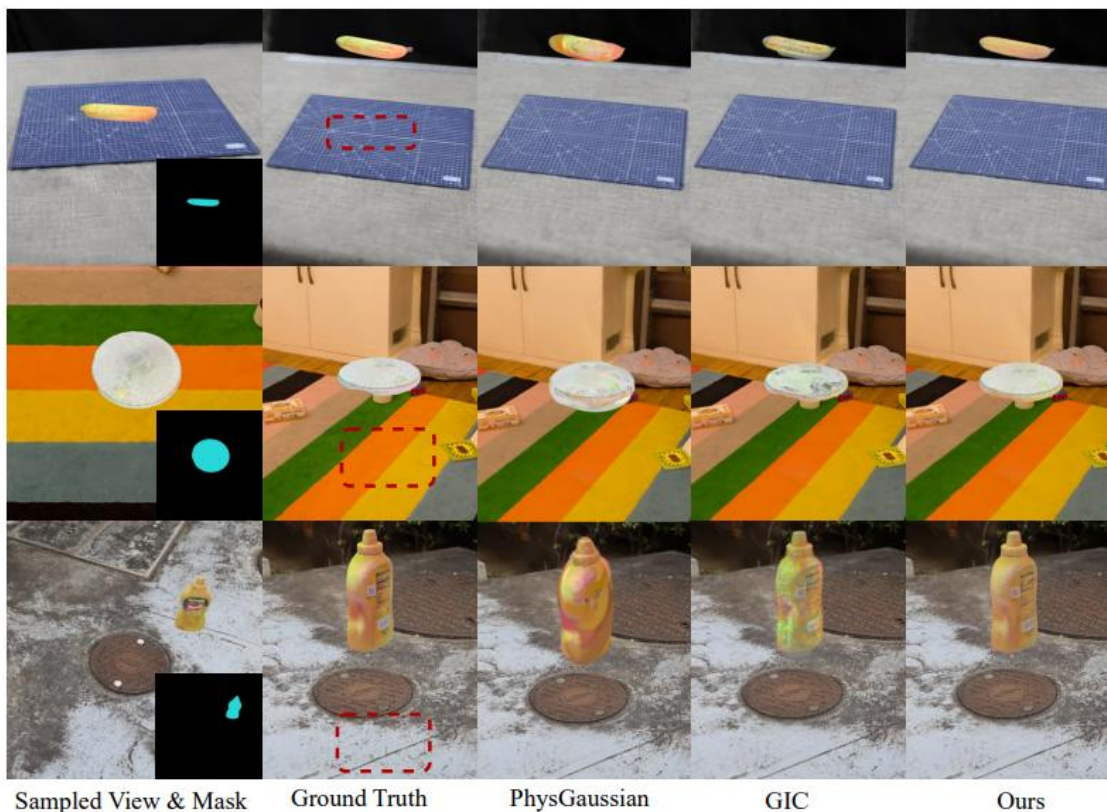


Figure 8. **Benchmark Comparisons.** A test viewpoint visualizes comparisons using the restored scene from our method, with inpainting regions marked by a red rectangle in the Ground Truth.

Table 2. **Quantitative Comparisons & Ablations.** We create a decoupling benchmark with comprehensive metrics comparing baselines and ablations to validate design choices.

Scene Restoration				
Methods	PSNR \uparrow	LPIPS \downarrow	FID \downarrow	CD ($\times 10^{-3}$) \downarrow
GScram [83]	17.82	0.56	42.28	44.00
VR-GS [38]	25.13	0.32	58.50	6.41
Ours	27.32	0.30	32.07	4.40
Object Restoration				
Methods	PSNR \uparrow	LPIPS \downarrow	FID \downarrow	CD ($\times 10^{-3}$) \downarrow
PhysGaussian [89]	24.46	0.07	227.60	0.53
GIC [6]	26.62	0.06	201.91	0.73
Ours	30.32	0.04	138.75	0.17
Object-Scene Interaction Simulation				
Methods	PSNR \uparrow	LPIPS \downarrow	FID \downarrow	Motion-FID \downarrow
PhysGaussian [89]	19.48	0.37	112.55	54.79
GIC [6]	20.90	0.31	134.56	47.47
w/o dense P_O	21.19	0.29	98.19	48.39
w/o Proxy P_O	21.08	0.30	90.26	36.01
w/o \mathcal{W}	20.97	0.30	96.16	42.27
Ours	21.33	0.29	86.98	31.69

Geometry accuracy using CD

Motion accuracy use Motion-FID

Ablations

Dense Points prevent collapse under gravity, while Joint Point Fields eliminate intersection regions.



Ours (Scene) +
Ground Truth (Object)



Ours (Scene) +
Ours (Object)



Ours (Scene) +
w/o Dense Points



Ours (Scene) +
w/o Joint Poisson Fields

Ablations

Note: Opacity is set to one for TRUCK to highlight artifacts



Ours (Scene) +
Ours (Object)



Ours (Scene) +
w/o UNCE (Object)

Take away

- Restore the object, and contact scene to simulate.
- Not rely on 2D inpainting models to repair broken geometry.



Thank you!

Scan to our project website