

*Decoupled Gaussian: Object-
Scene Decoupling for Physics-
Based Interaction*

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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

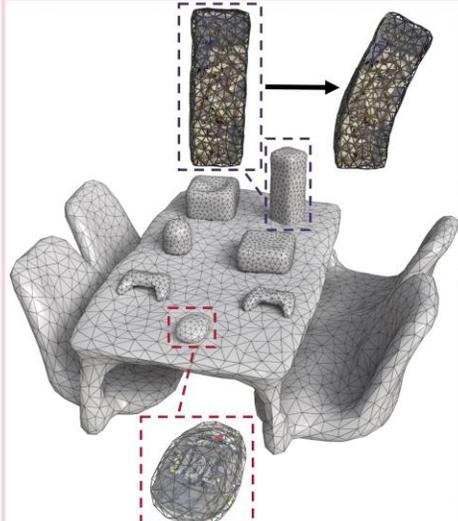
Object-level 3D Scene Reconstruction



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

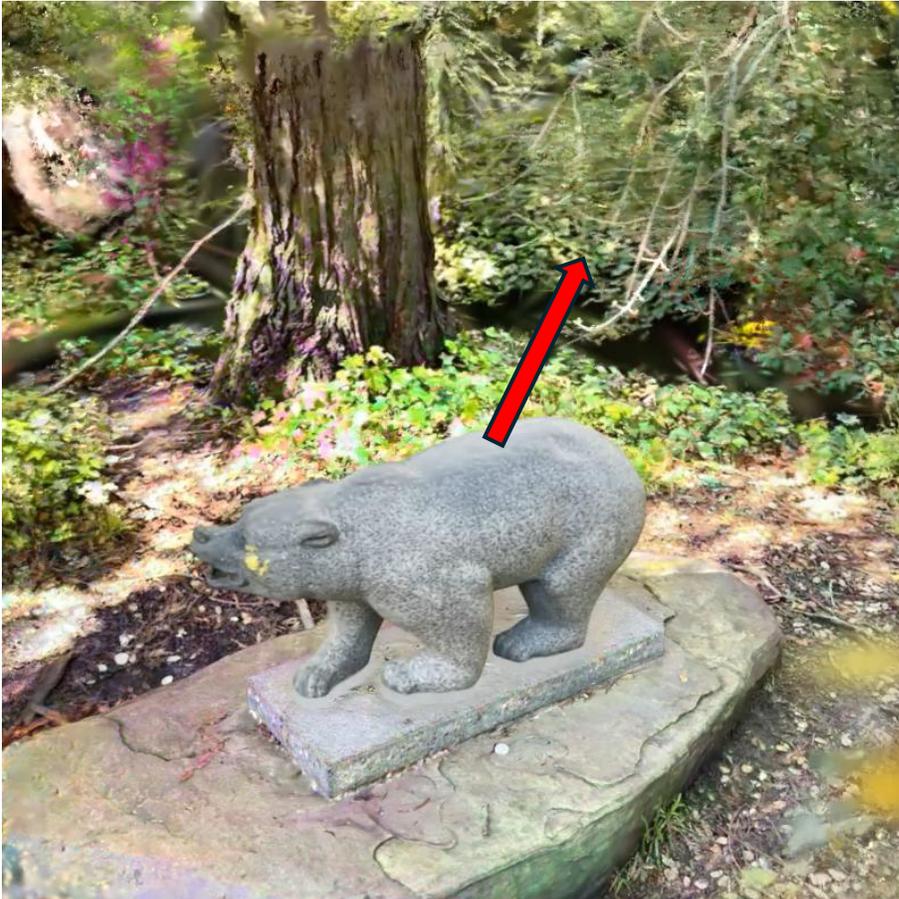


Scene



Object

Support Various Interactive Simulation



Scene Collisions



Object Melting

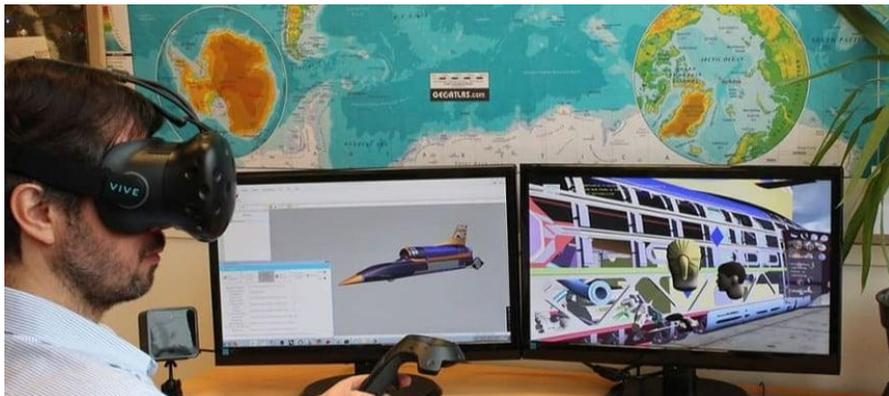
DecoupledGaussian: Meaning and Application



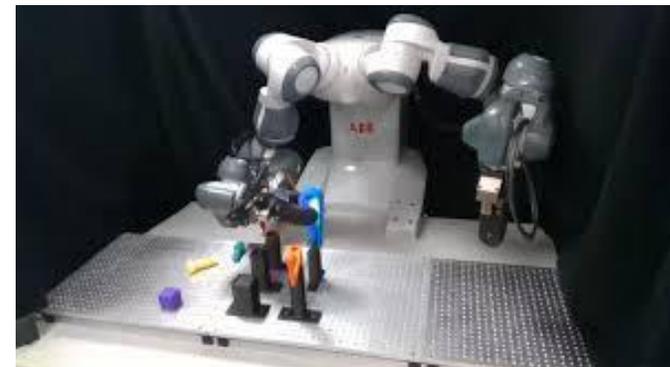
Autonomous driving Simulation



Entertainment industry



VR/AR Interaction



Robotic Learning Interaction

Our pipeline

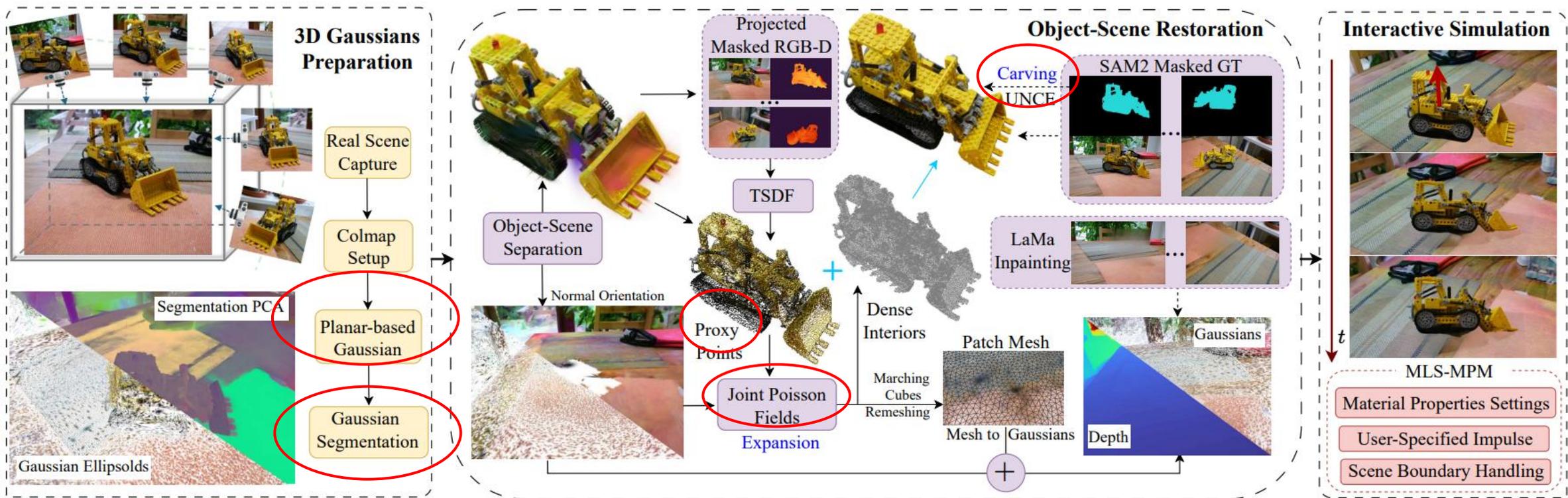


Figure 3. **System Overview.** DecoupledGaussian is an interactive simulation system that enables objects to detach from their initial contact surfaces after applying our proposed restoration pipeline, driven by user-specified impulses (red arrow on the right).

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

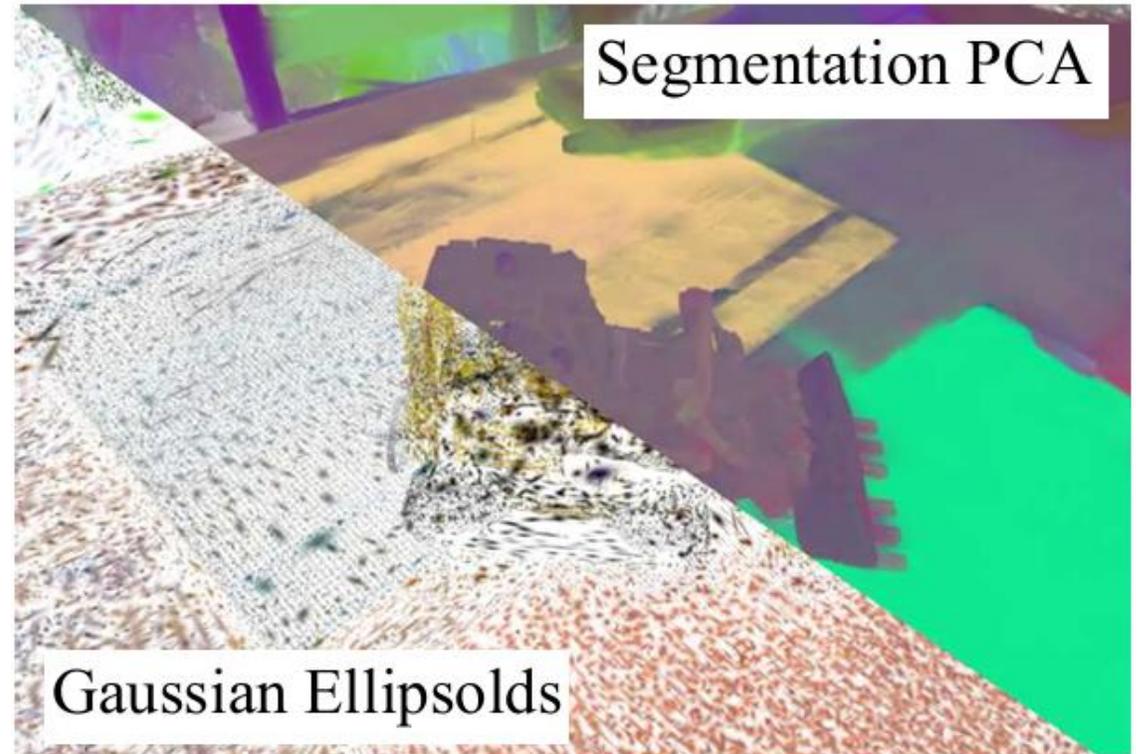


Opacity is set to 1, with $\times 0.4$ scaling for better Gaussian kernel visualization.

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

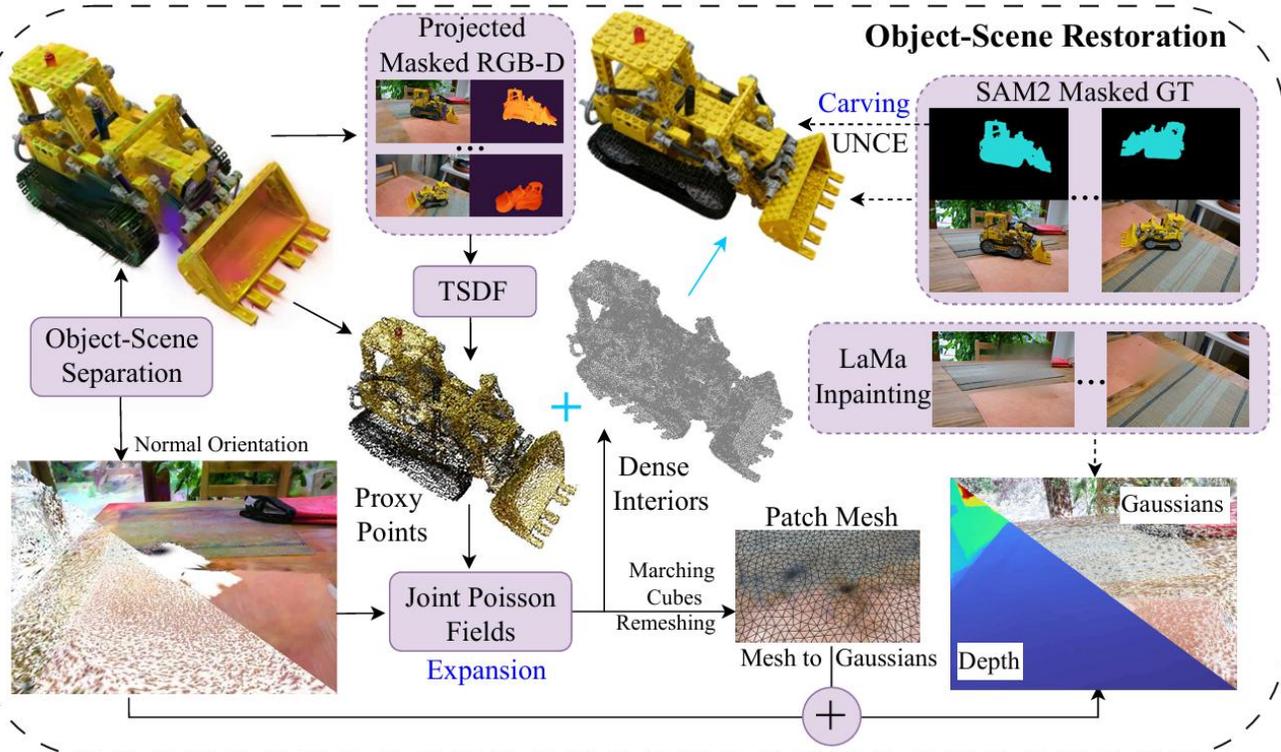
3D Gaussian Preparation: Segmentation

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

- (1) inpainted regions often fail to blend seam lessly 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

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

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

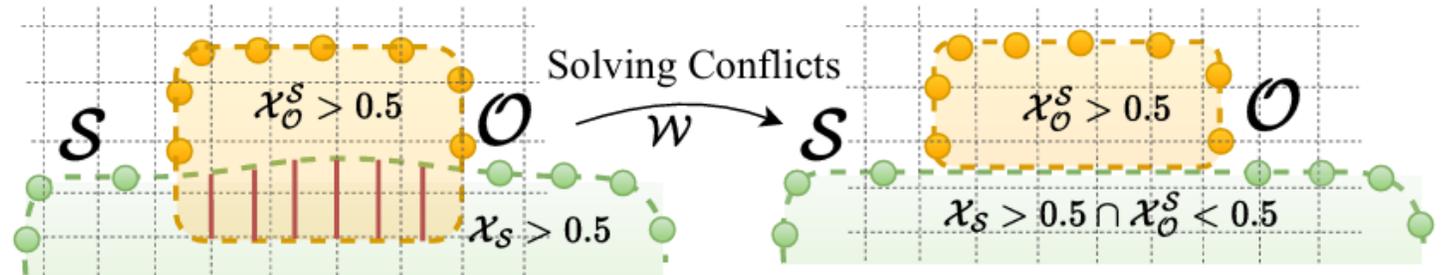


Figure 4. Joint Poisson Fields \mathcal{W} first reconstruct \mathcal{O} and \mathcal{S} independently, then resolve conflicts (red area) by defining a boundary that separates them into distinct, non-intersecting entities.

[Adaptive multigrid solvers]

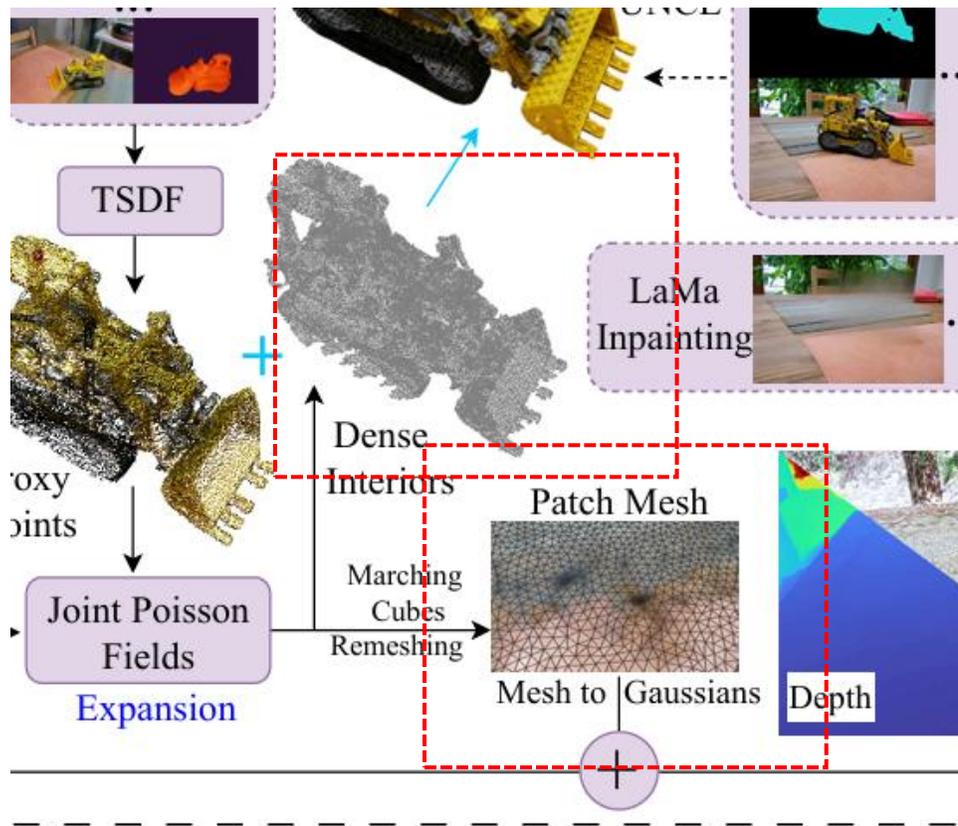
Joint Poisson Fields

Dense Points: helpful for MPM simulation

Patch Mesh: fix the broken geometry of the Surface

How about the input to Joint Poisson Fields:

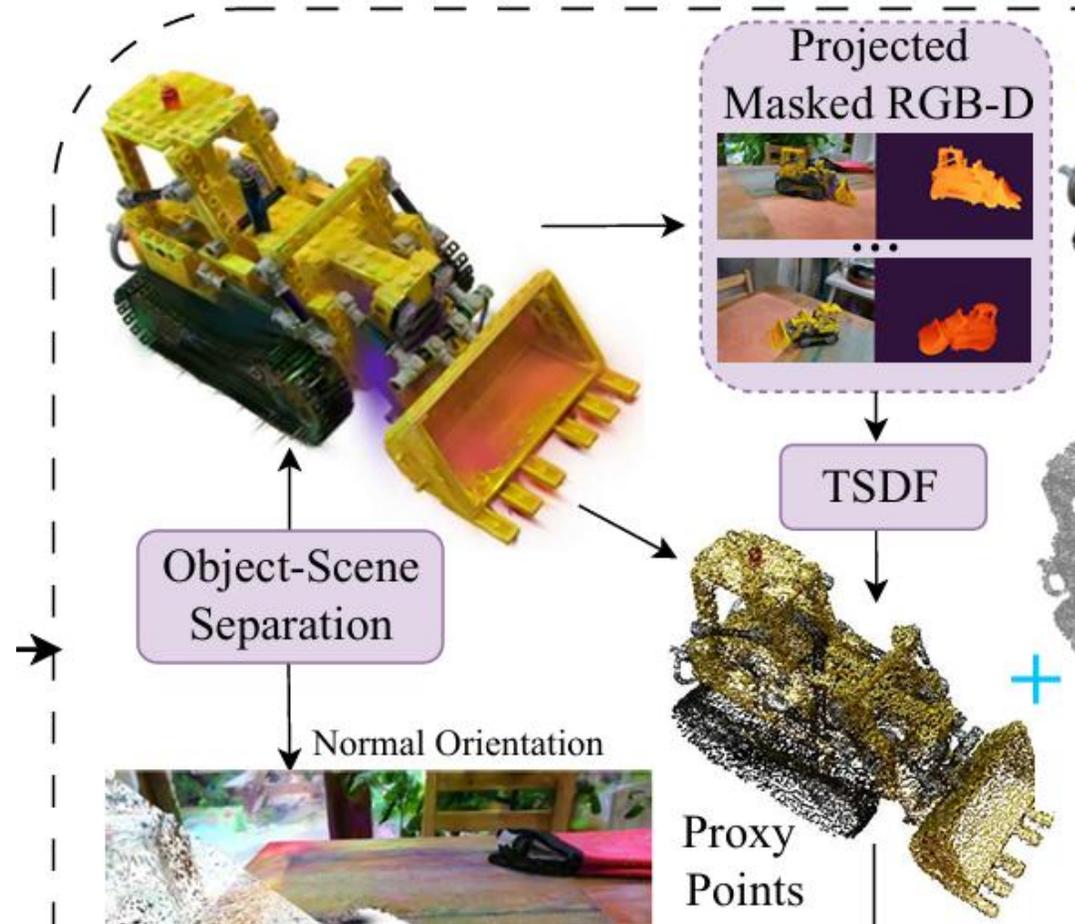
- Contact Surface
 - Gaussian Centers as points
 - Minimum scaling as normal (90-degree disambiguate)
- Target Object
 - Geometric complexity
 - Use Proxy Points instead of Gaussian Centers



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

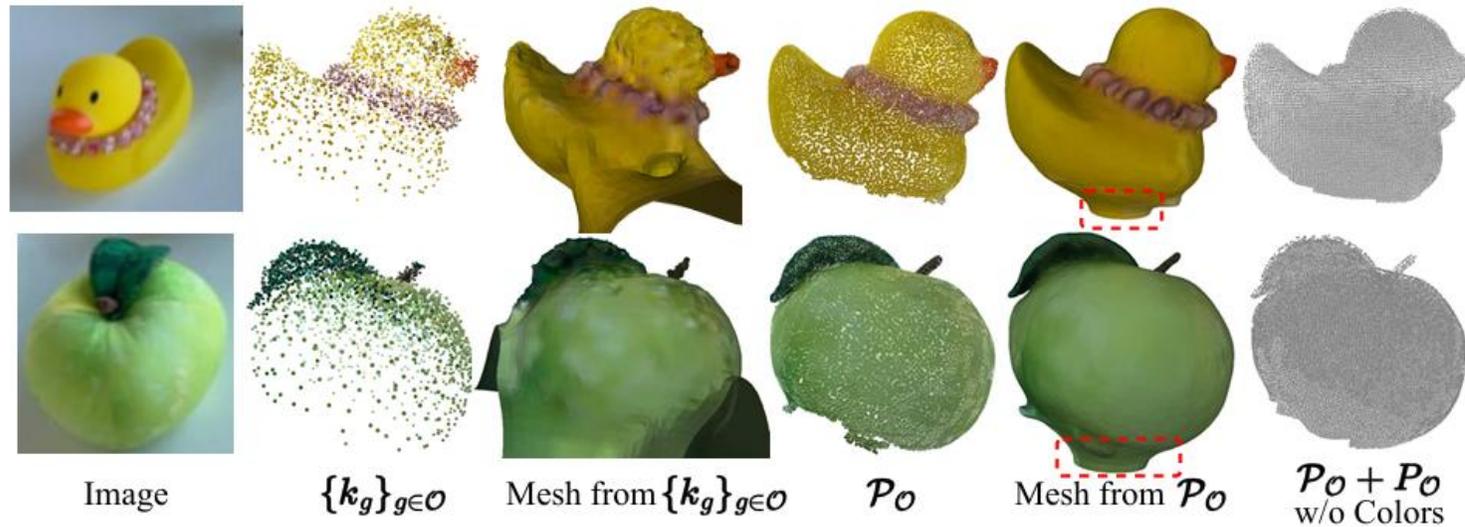


Figure 5. **Ablation for $\mathcal{P}_\mathcal{O}$.** Independent Poisson reconstruction of object \mathcal{O} using Gaussian centers $\{k_g\}_{g \in \mathcal{O}}$ yields poor mesh quality compared to using proxy points $\mathcal{P}_\mathcal{O}$. Our joint Poisson field \mathcal{W} , which integrates 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



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]



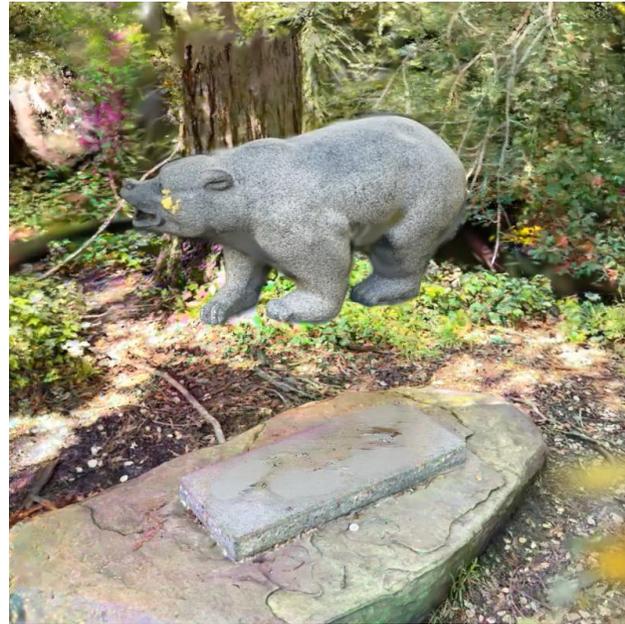
Ours

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

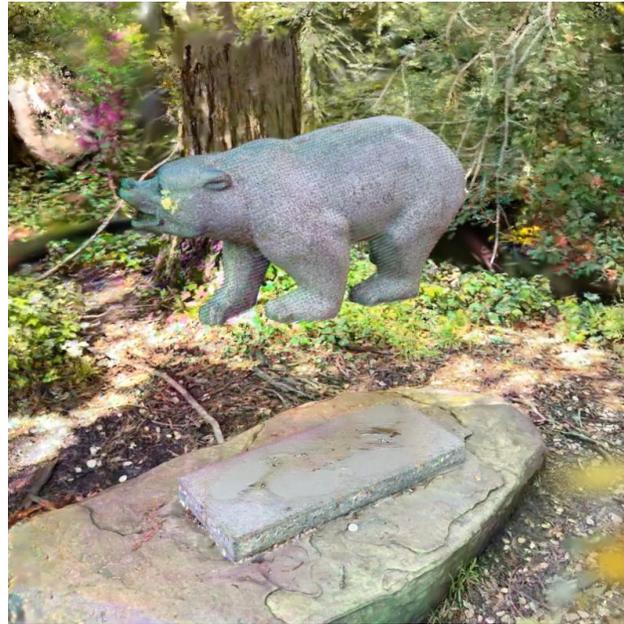
Object-Scene Interactive Simulation



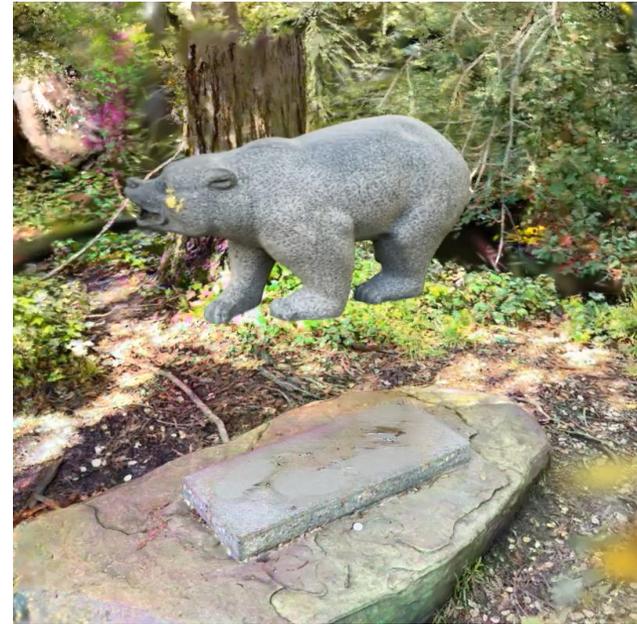
VR-GS (Scene) +
PhysGaussian (Object)



Ours (Scene) +
PhysGaussian (Object)



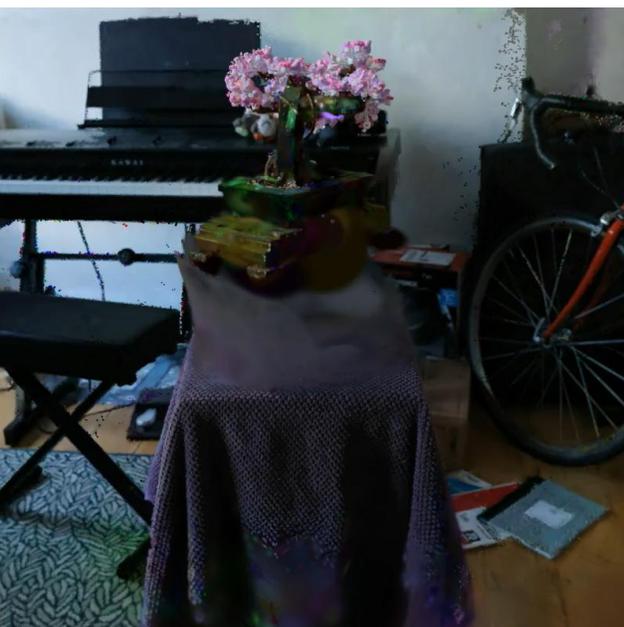
Ours (Scene) +
GIC (Object)



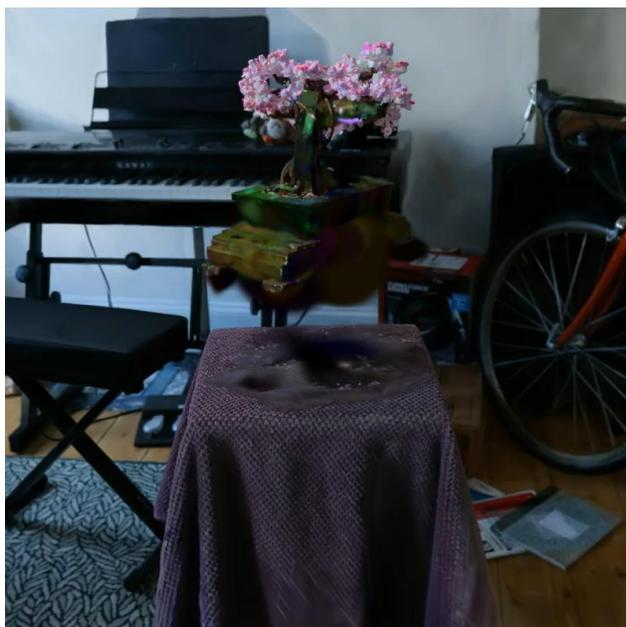
Ours (Scene) +
Ours (Object)

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

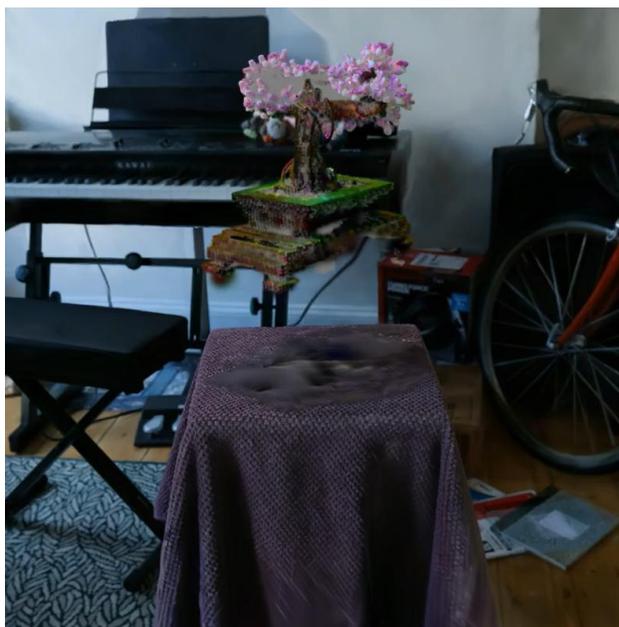
Object-Scene Interactive Simulation



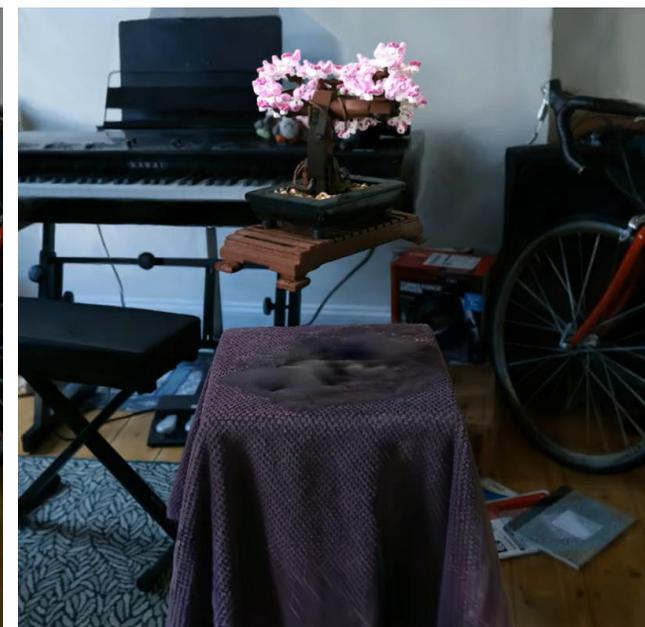
VR-GS (Scene) +
PhysGaussian (Object)



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.

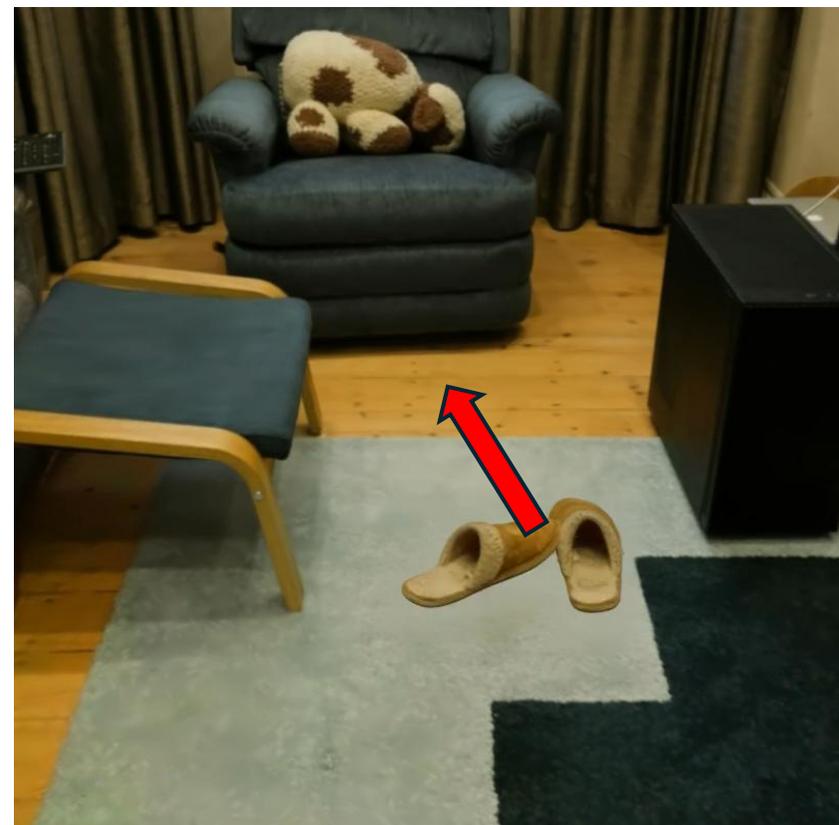
More Cases (Ours)



Bonsai

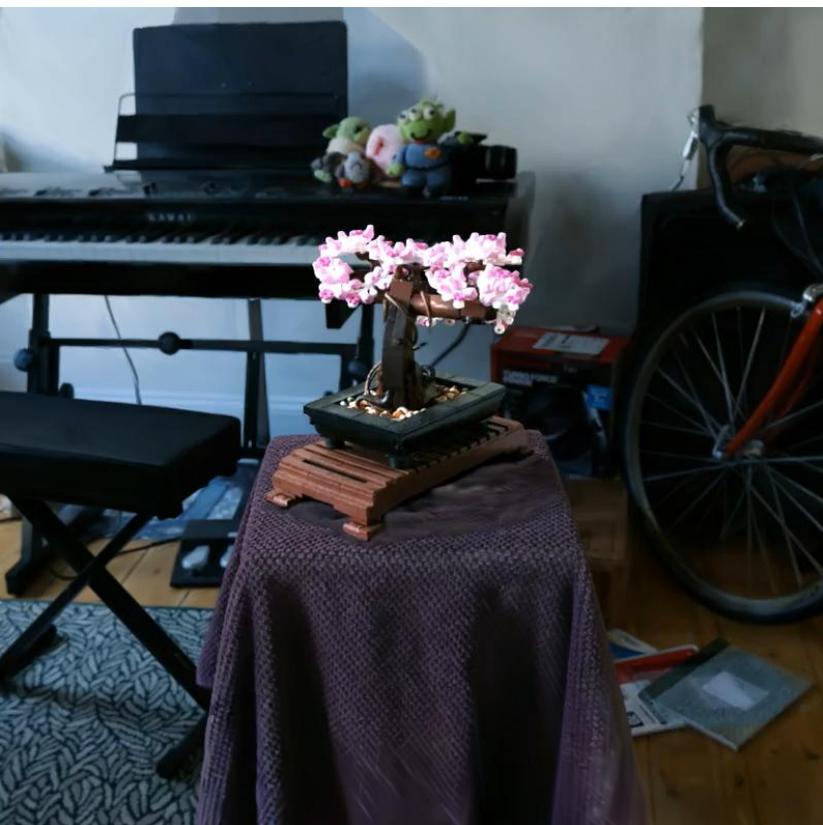


Kitchen



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
GScreen [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

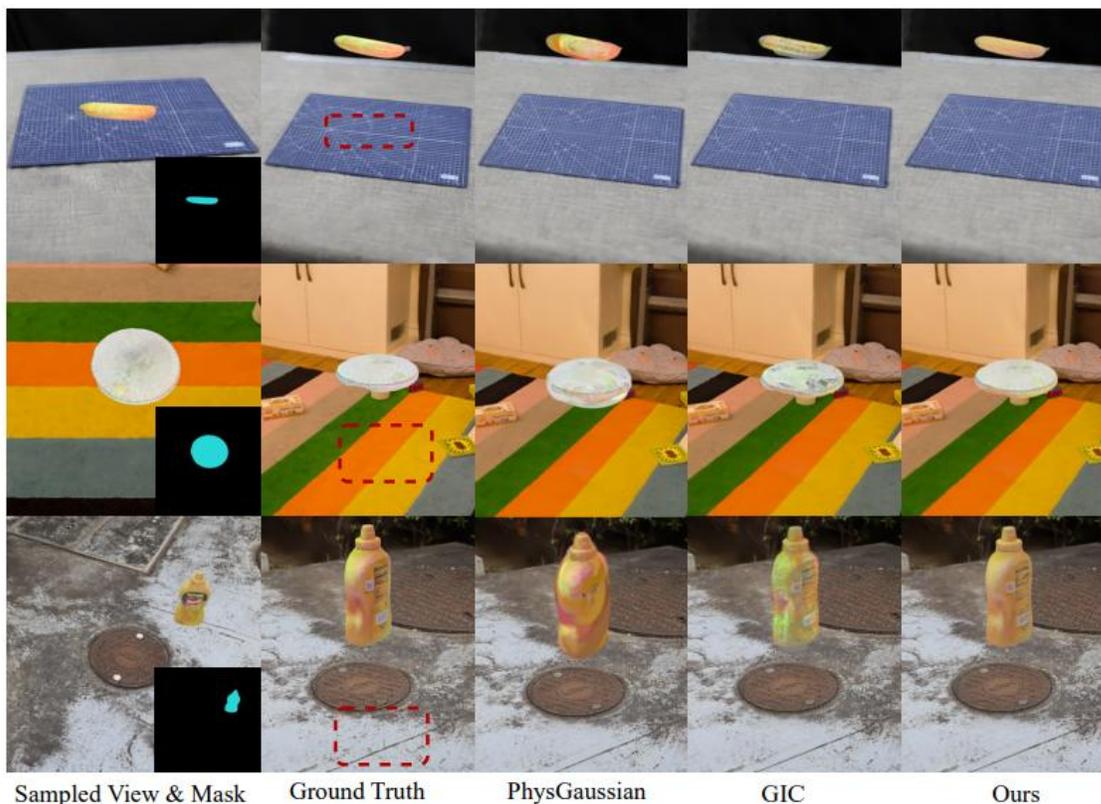


Figure 8. **Benchmark Comparisons.** A test viewpoint visualizes comparisons using the restored scene from our method, with inpainting regions marked by a red rectangle in the Ground Truth.

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
GScreen [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 $P_{\mathcal{O}}$	21.19	0.29	98.19	48.39
w/o Proxy $P_{\mathcal{O}}$	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

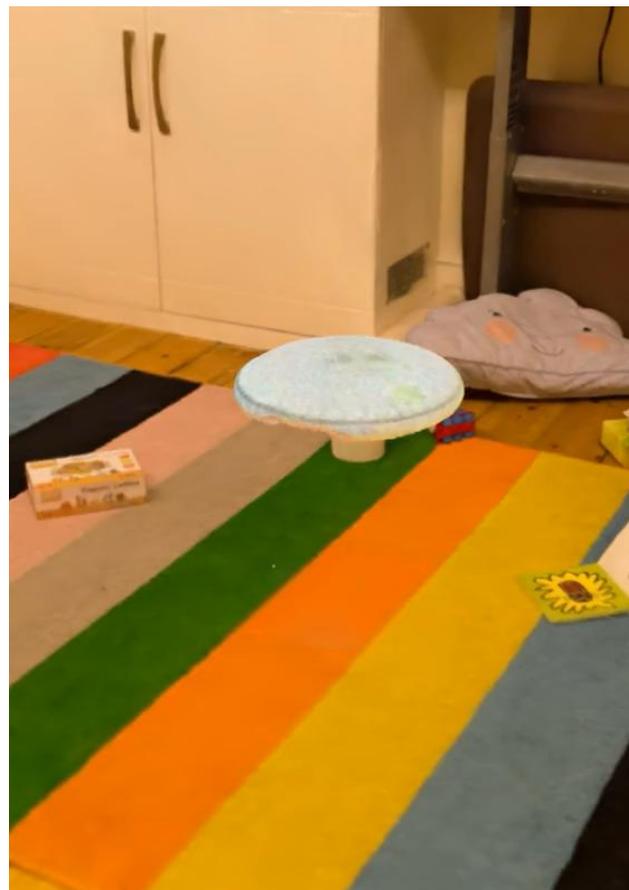
Dense Points prevent collapse under gravity, while Joint Point Fields eliminate intersection regions.



Ours (Scene) +
Ground Truth (Object)



Ours (Scene) +
Ours (Object)



Ours (Scene) +
w/o Dense Points



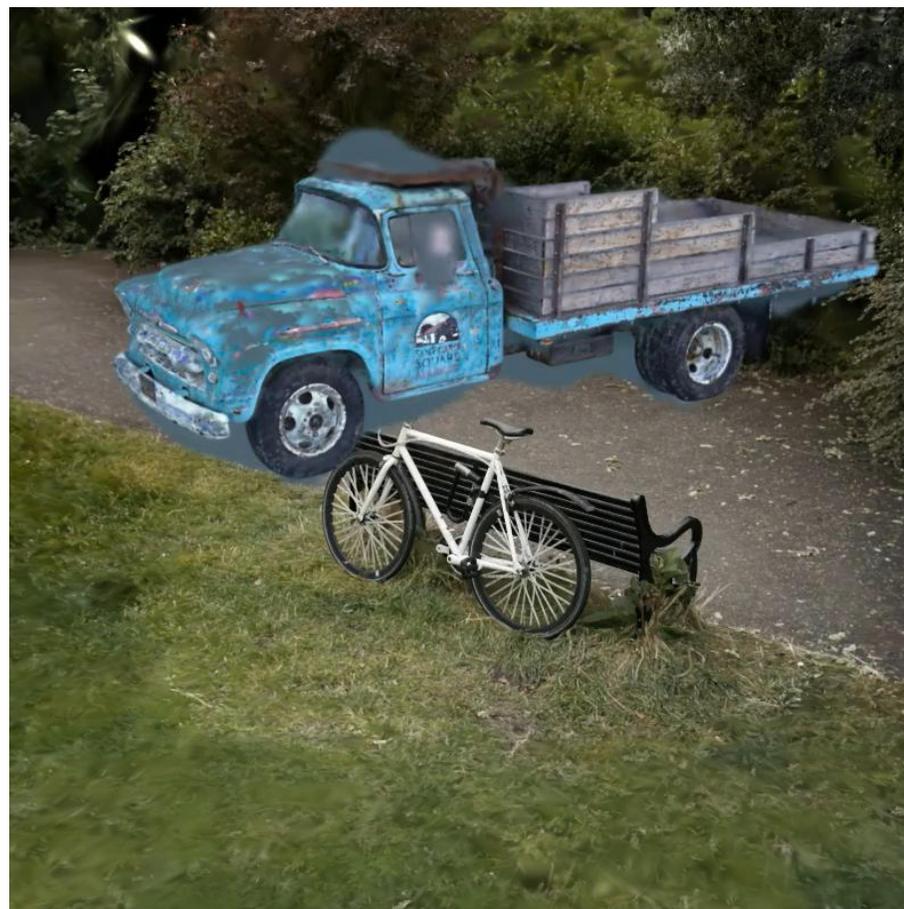
Ours (Scene) +
w/o Joint Poisson Fields

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