

Seeing Through Clutter: Structured 3D Scene Reconstruction via Iterative Object Removal

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Current Directions in Scene Generation

LLM Scene Generation

Victorian-style living room



- Examples often setting restricted (e.g. indoor scenes)
- Rely on LLM domain knowledge

HOLODECK (2023)

Diffusion/NeRF-Based Methods

(a) Generated scene



"a chicken hunting for easter eggs"

(b) Disentangled objects



- Great results, but panorama based
- Unstructured representation

Disentangled 3D Scene Generation
with Layout Learning (2023)

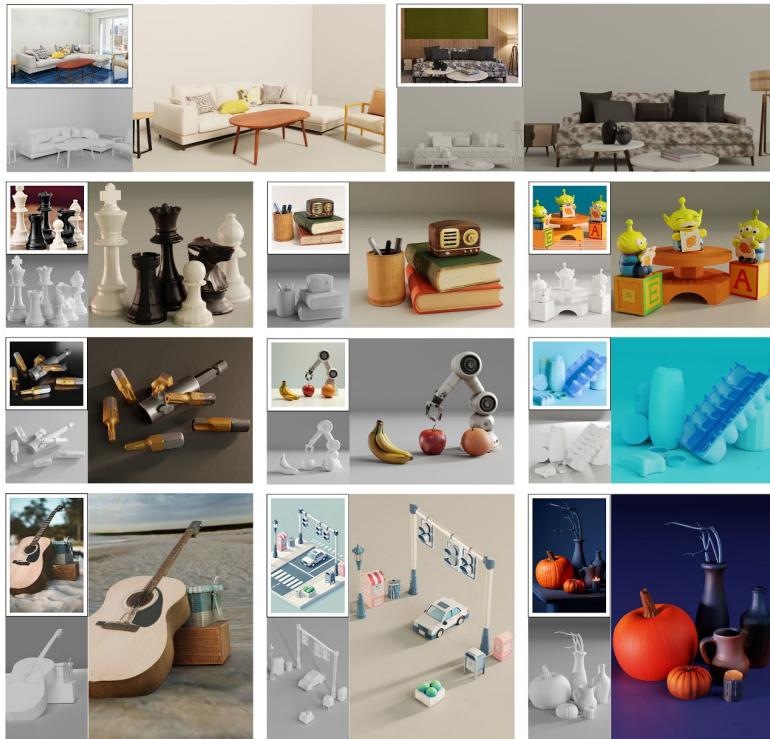
Single-View Scene Synthesis



- Often limited, examples either:
 - Require additional training
 - Rely on limited 3D object databases for retrievals
 - Aren't open-vocabulary

Diorama (2024), CAST (2025), MIDI (2025)

Why is single-view scene synthesis limited?



Smaller scale scenes



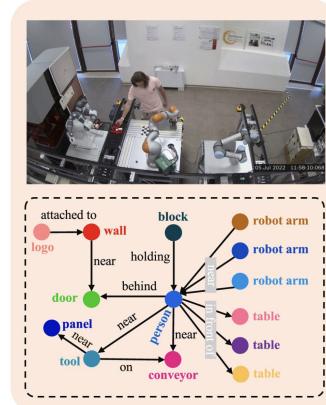
Poor Fitting

Why is single-view scene synthesis limited?



Image Segmentation

Depth Estimation



VLM Annotations

Why is single-view scene synthesis limited?



Why is single-view scene synthesis limited?

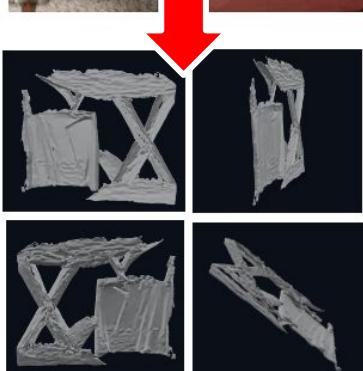


Why is single-view scene synthesis limited?



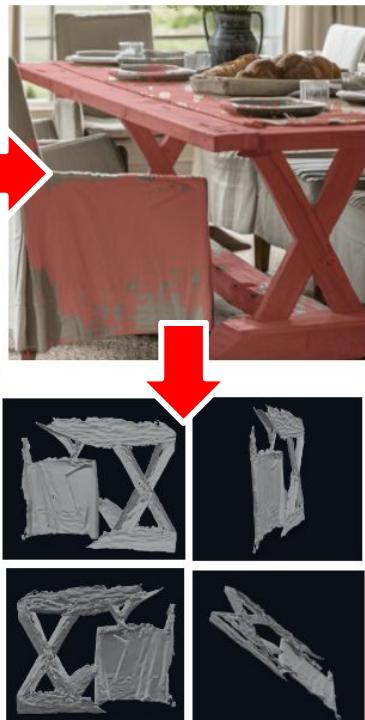
Why is single-view scene synthesis limited?

A

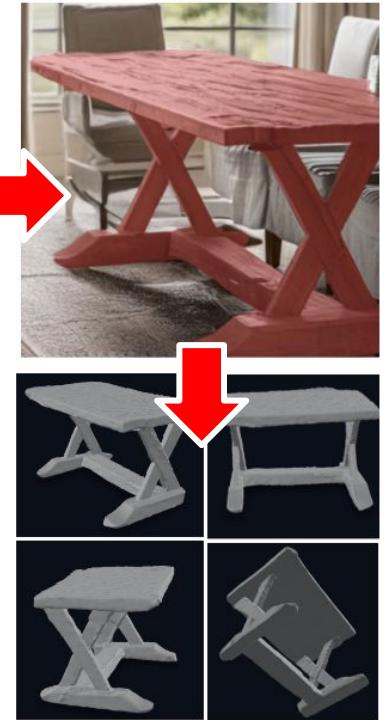


Why is single-view scene synthesis limited?

A



B



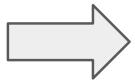
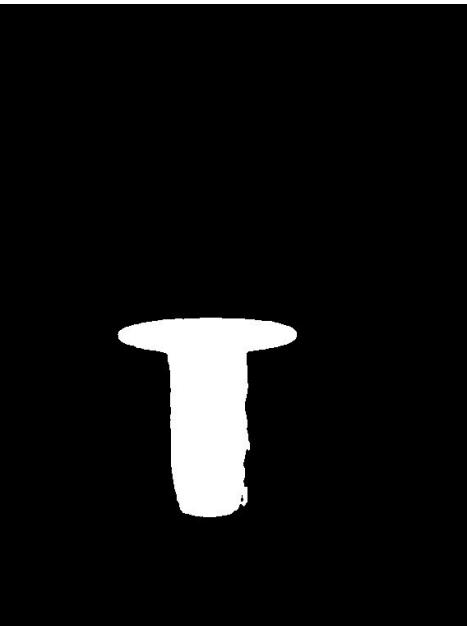
Automating Object Removal?



Promising Results in Object Removal

SmartEraser (2025)

Caveats in Object Removal



Automating Object Removal?



Tables are symmetrically placed in the room; each table should have two chairs on opposite sides of the table facing each other, ready for dining ...

All the tables aligned to form a line, dividing the room up into two halves, place all the chairs on one side of the line and the buffets on the other side



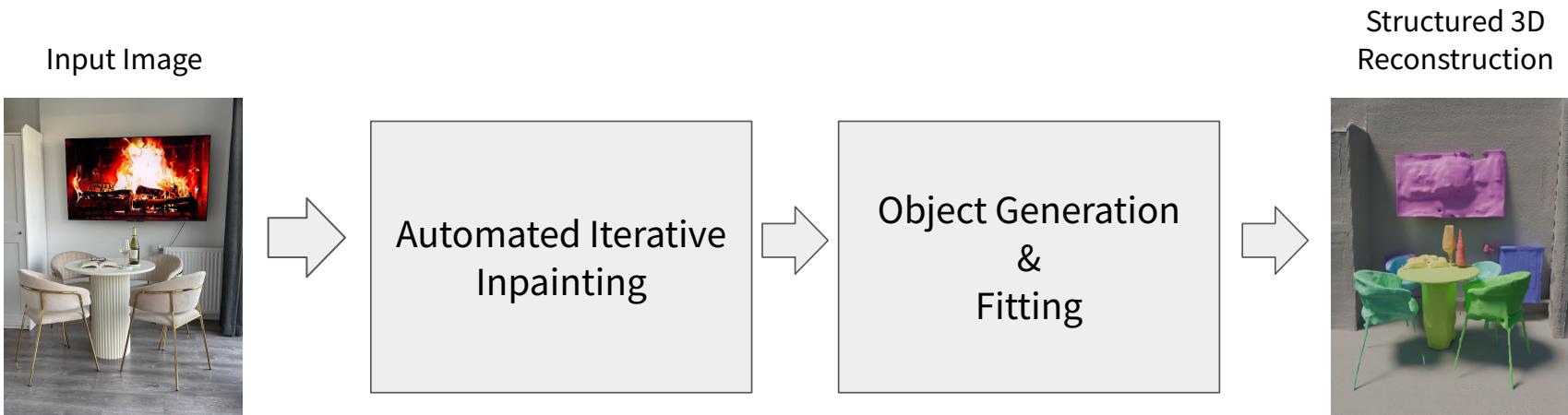
Promising Results in Object Removal

SmartEraser (2025)

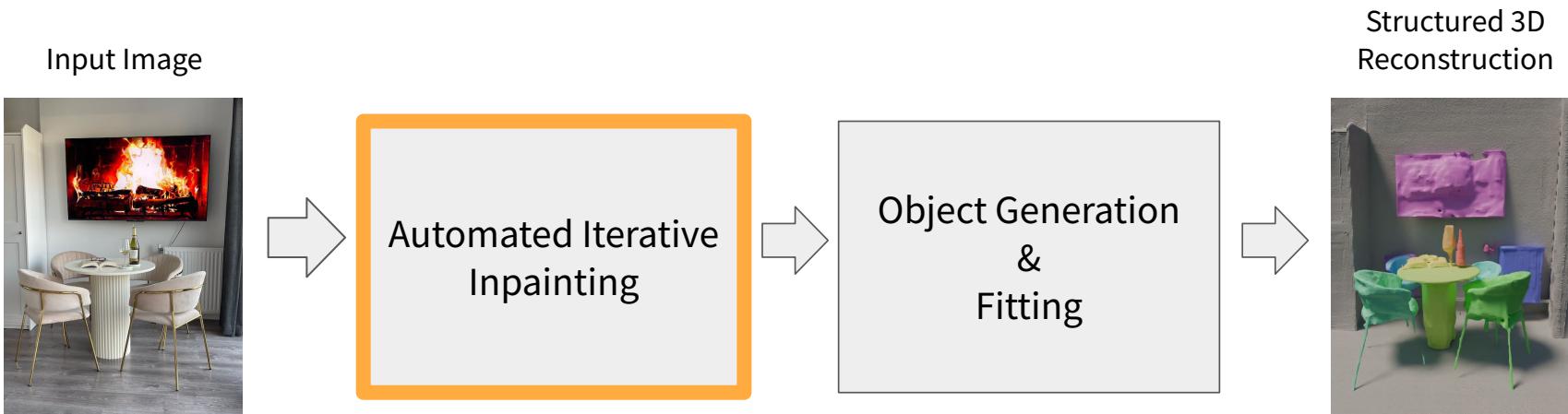
Spatial Reasoning in VLMs?

LayoutVLM (2025)

Method Overview



Method Overview



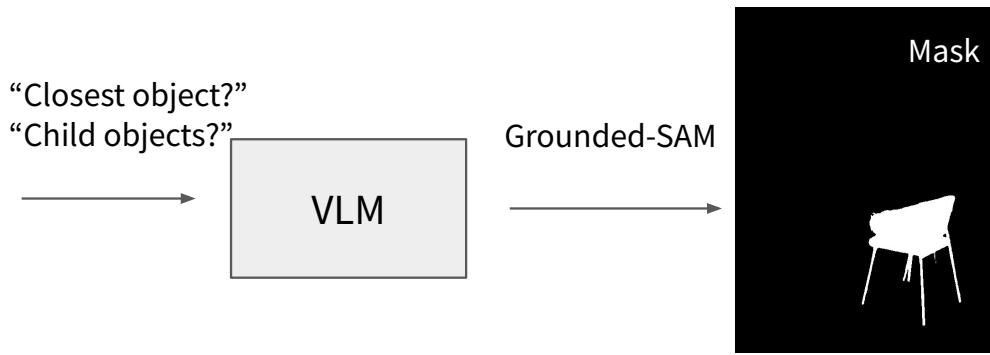
Method: Automated Inpainting Pipeline

Input Image



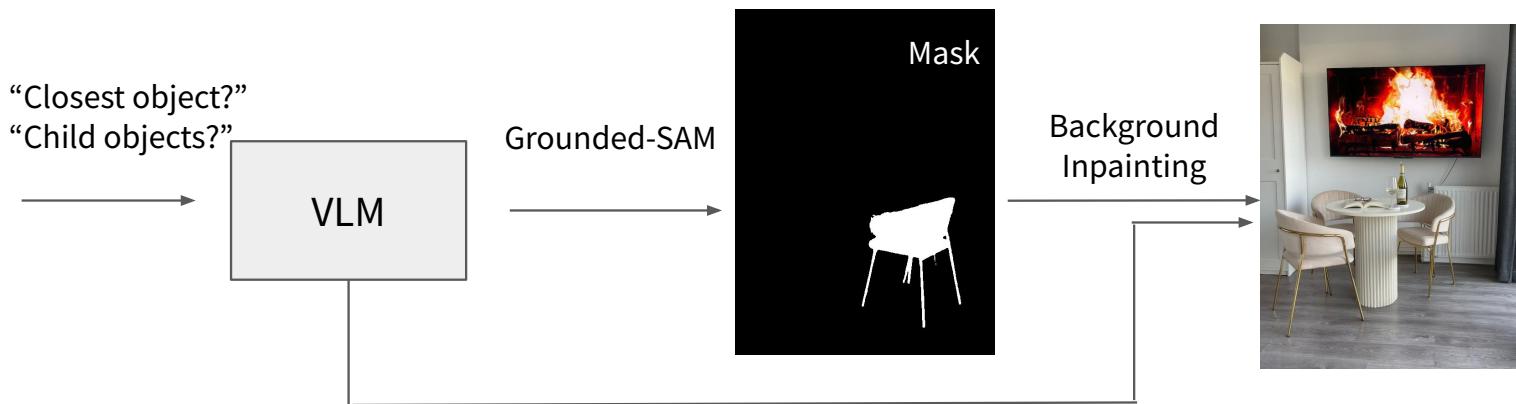
Method: Automated Inpainting Pipeline

Input Image



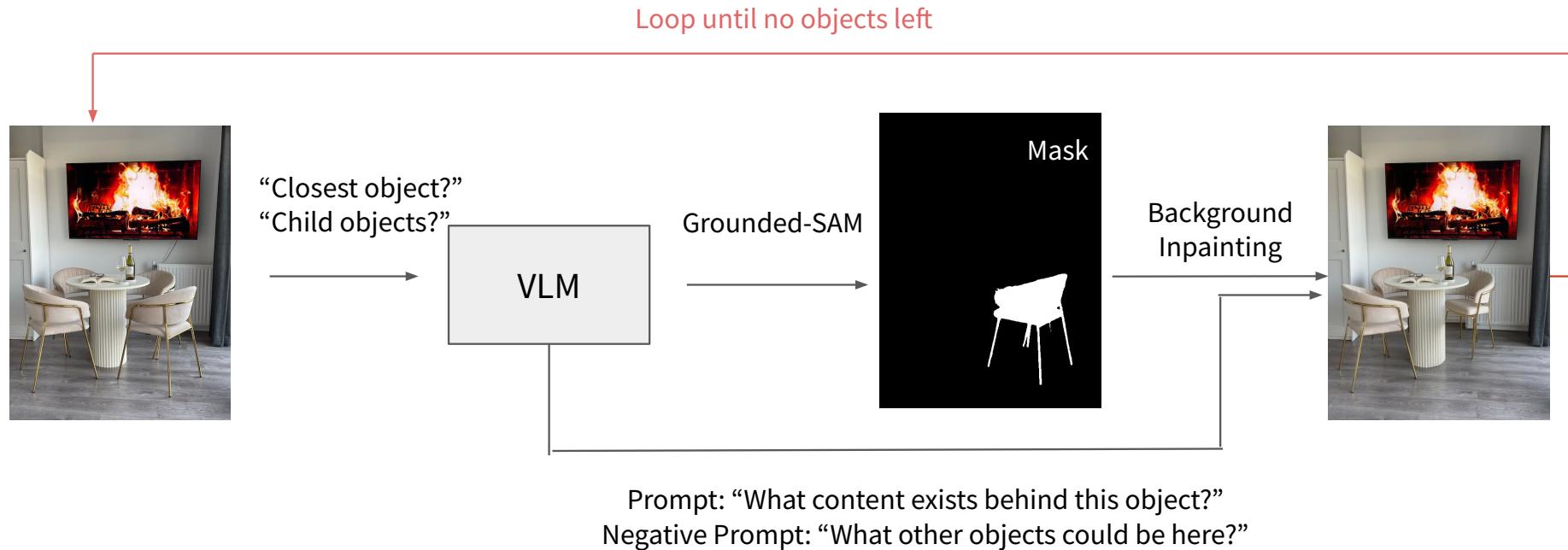
Method: Automated Inpainting Pipeline

Input Image

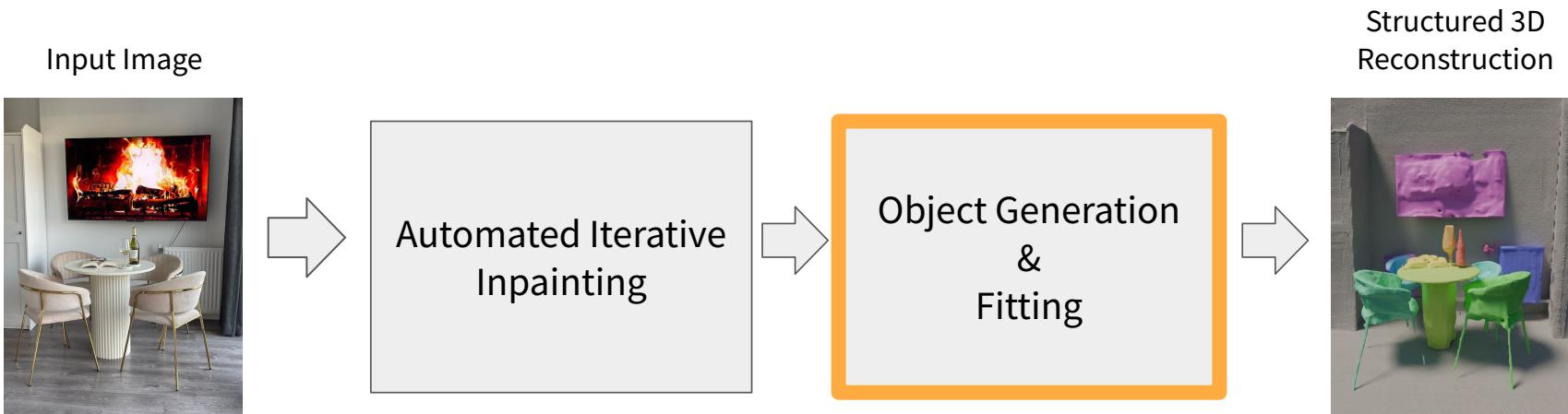


Prompt: "What content exists behind this object?"
Negative Prompt: "What other objects could be here?"

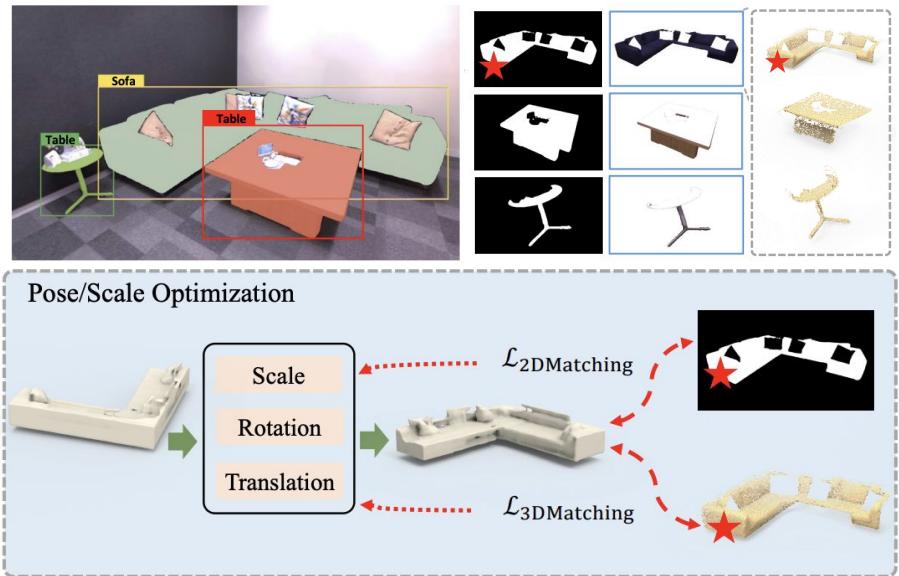
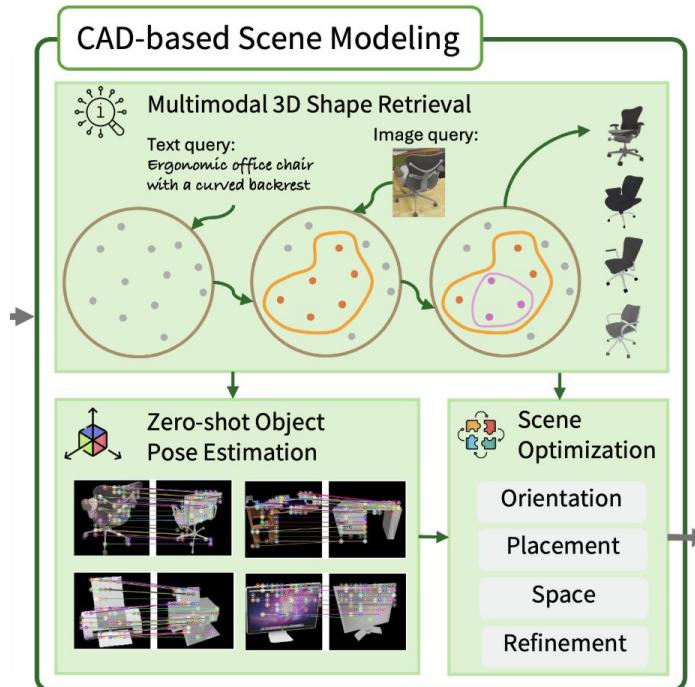
Method: Automated Inpainting Pipeline



Method Overview



Prior work: Object Generation & Fitting



Diorama
(2024)

DeepPriorAssembly
(2024)

Method: Object Generation & Fitting

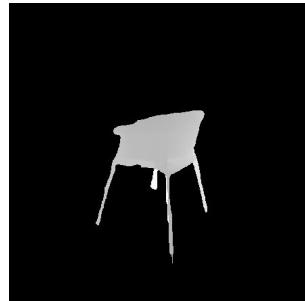


Amodal Depth
Segmentations

+



Camera-view centric 3D
reconstruction



Depth renders

Method: Object Generation & Fitting

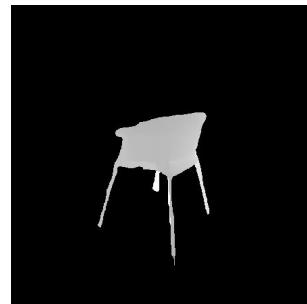


Amodal Depth
Segmentations

+



Camera-view centric 3D
reconstruction



Depth renders

Fitting



Final Reconstruction

Object fitting becomes a simplified 4 DoF (uniform scale & translation) RANSAC problem!

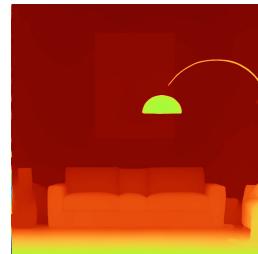
Method: Depth Alignment

$i = 1$



,

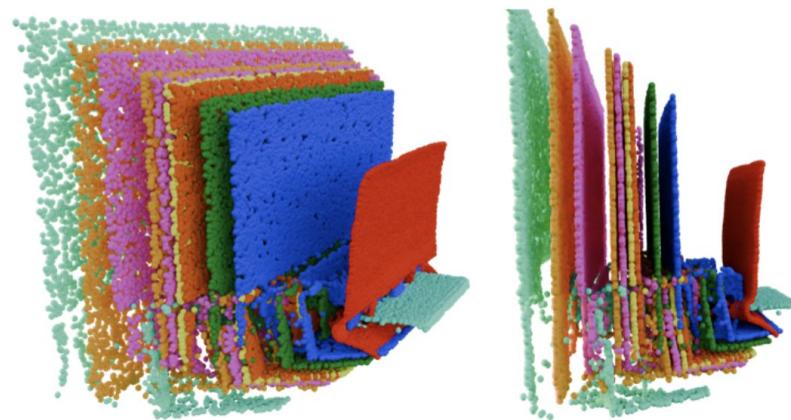
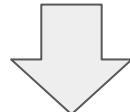
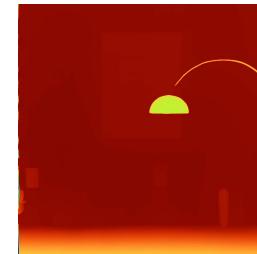
$i = 2$



,

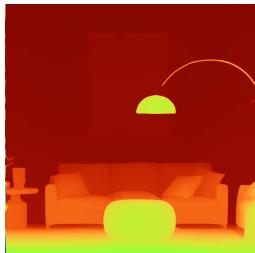
\dots

$i = n$



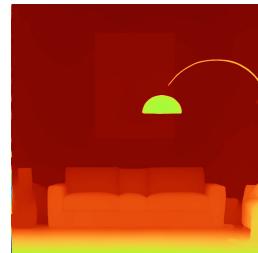
Method: Depth Alignment

$i = 1$



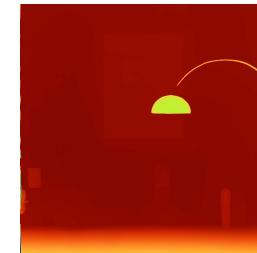
Fixed
Reference

$i = 2$



\dots

$i = n$

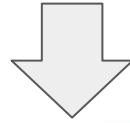


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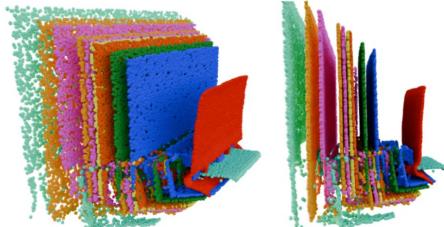
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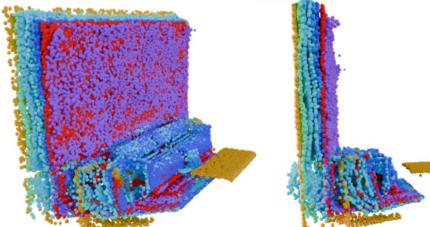
$$\mathcal{L}(\theta) = \sum_{n=1}^{N-1} \sum_{x,y} (1 - M_n(x,y)) |\mathcal{D}'_n(x,y) - \mathcal{D}'_{n+1}(x,y)|.$$



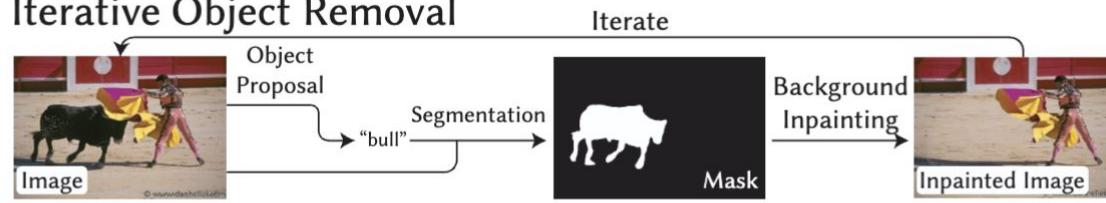
Before



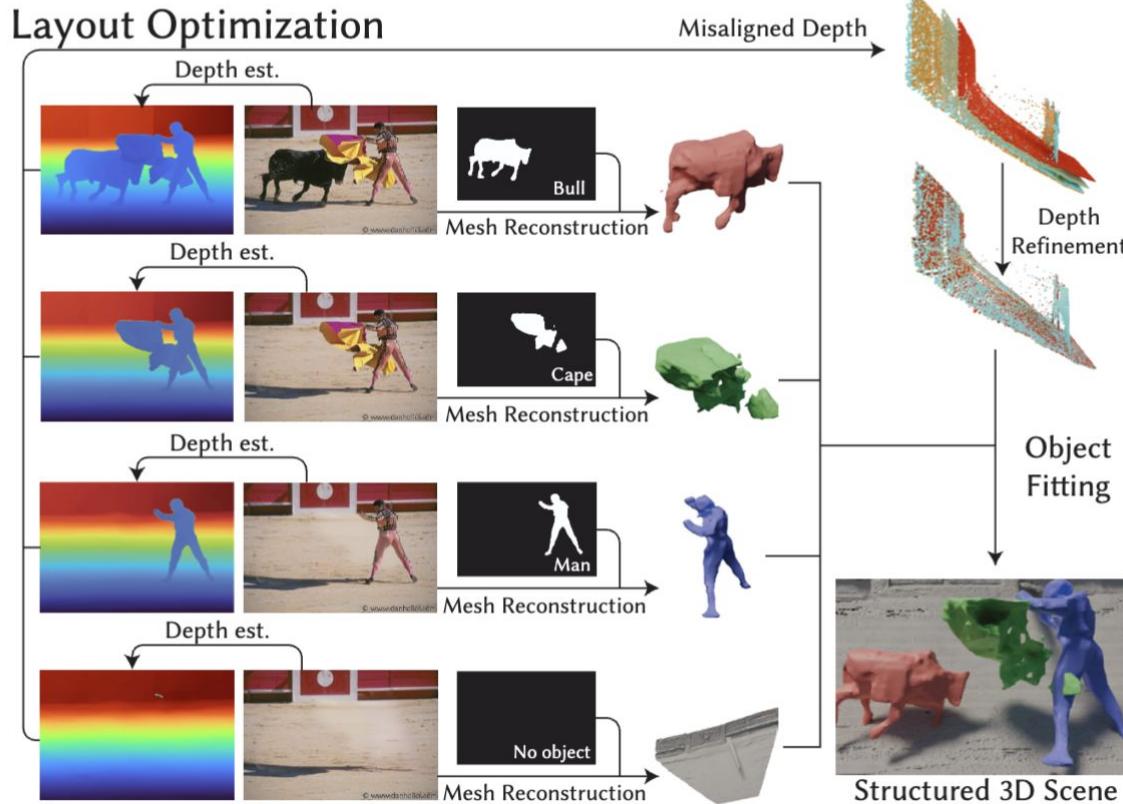
After



Iterative Object Removal

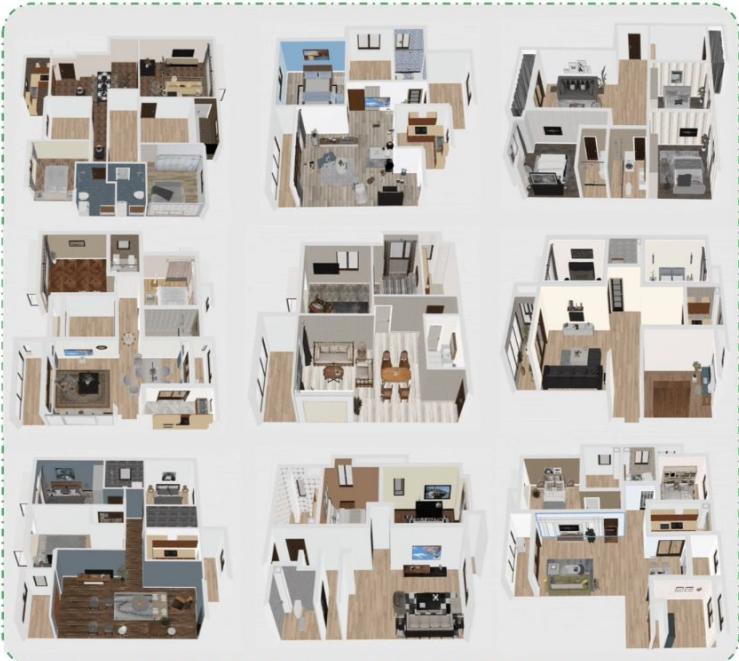


Layout Optimization



Results: Existing Benchmarks

3D FRONT



MIT Scene Parsing Benchmark



- Synthetic Indoor Scenes w/ 3D GT
- Well-used benchmark for single-view reconstruction

- Scene segmentation benchmark w/ semantic annotation (e.g. things vs. stuff)
- No 3D GT, strictly segmentation

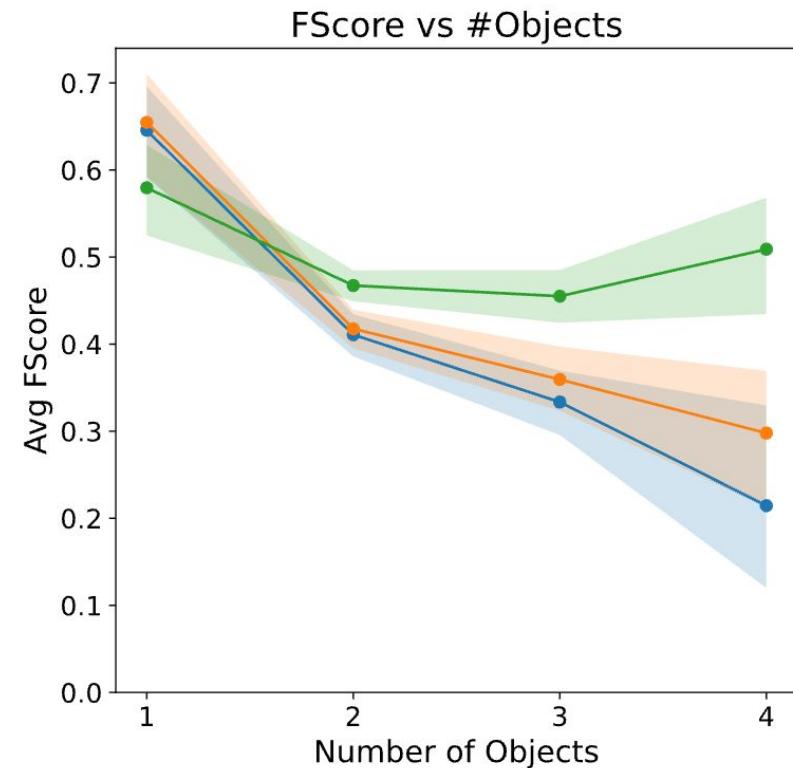
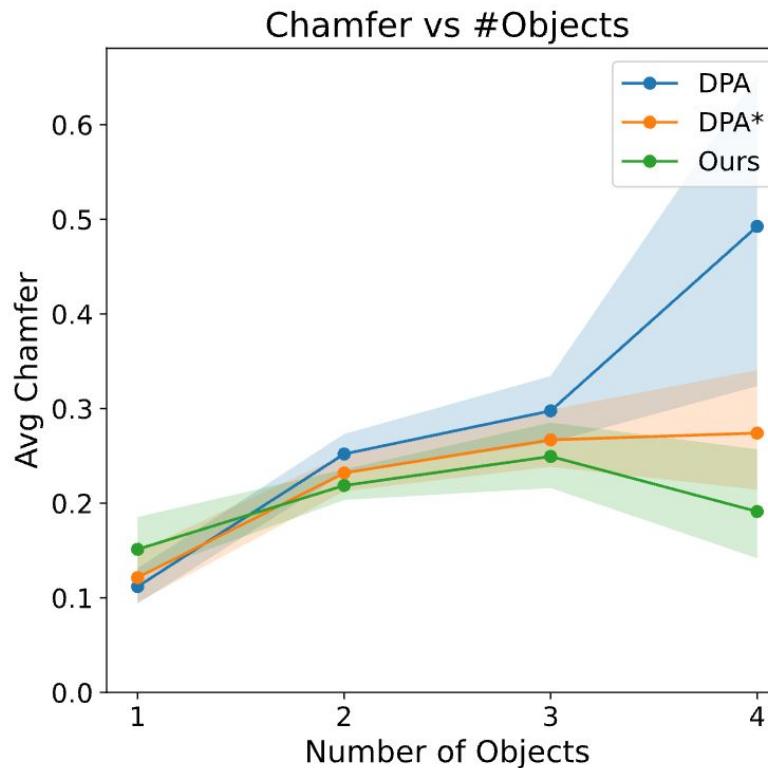
Results: Indoor Scenes (3D Front)

Baselines:

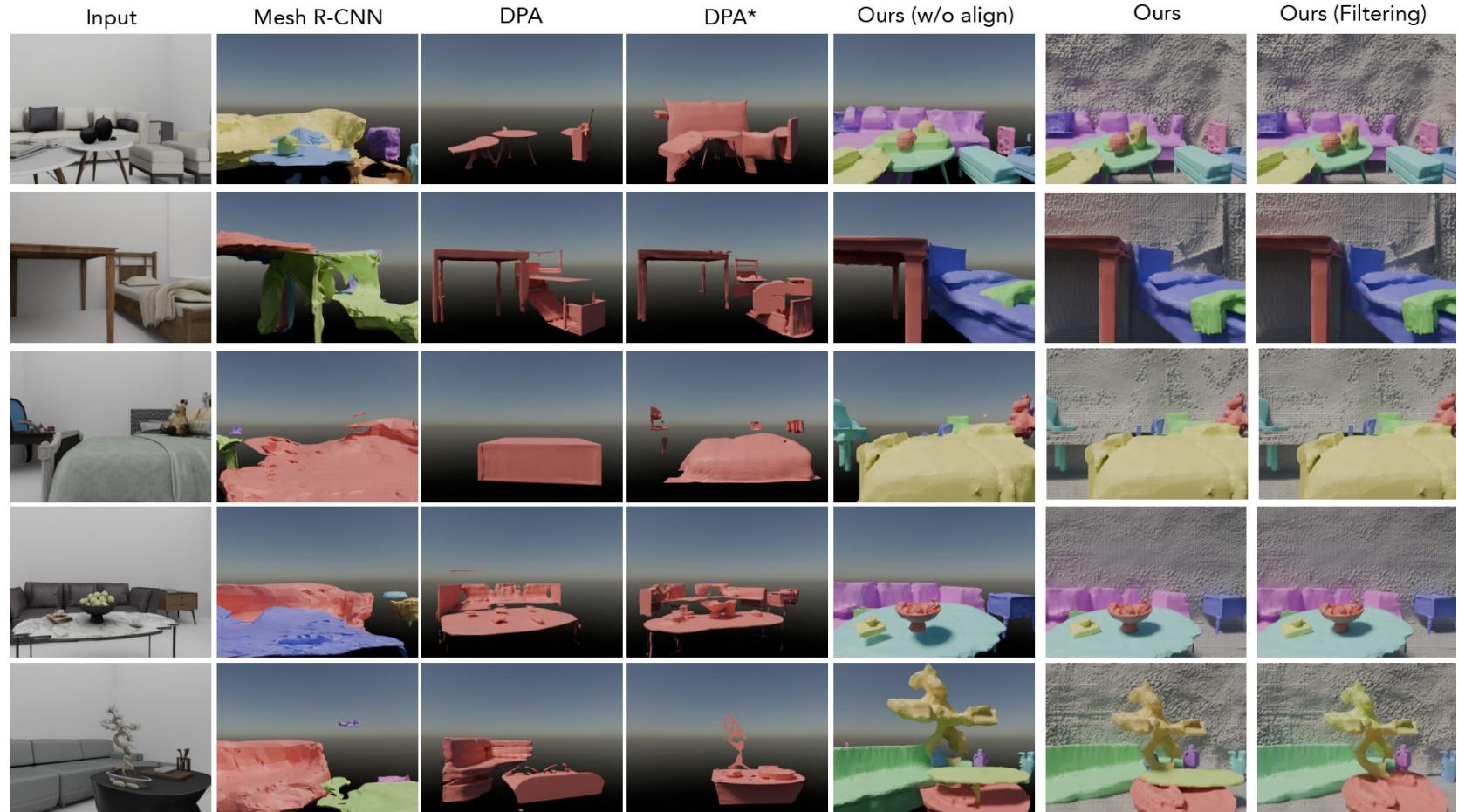
- DeepPriorAssembly (or DPA; closed vocabulary)
- DPA* (augmented w/ VLM labels)
- Ours (without depth alignment)
- Ours (w/ object artifact filtering)

Model	CD ↓	FS ↑	Obj-FS ↑	Depth ↓	IoU ↑	M-IoU ↑
DPA	24.90	42.49	9.357	0.287	0.726	0.213
DPA*	23.04	42.73	9.037	0.286	0.751	0.205
ours (w/o depth ref.)	23.38	45.36	11.21	0.095	0.787	0.529
ours (w/ object filt.)	21.80	47.67	12.70	0.082	0.817	0.534
ours	21.66	48.07	12.53	0.085	0.817	0.539

Results: Indoor Scenes (3D Front)



Results: Indoor Scenes (3D Front)

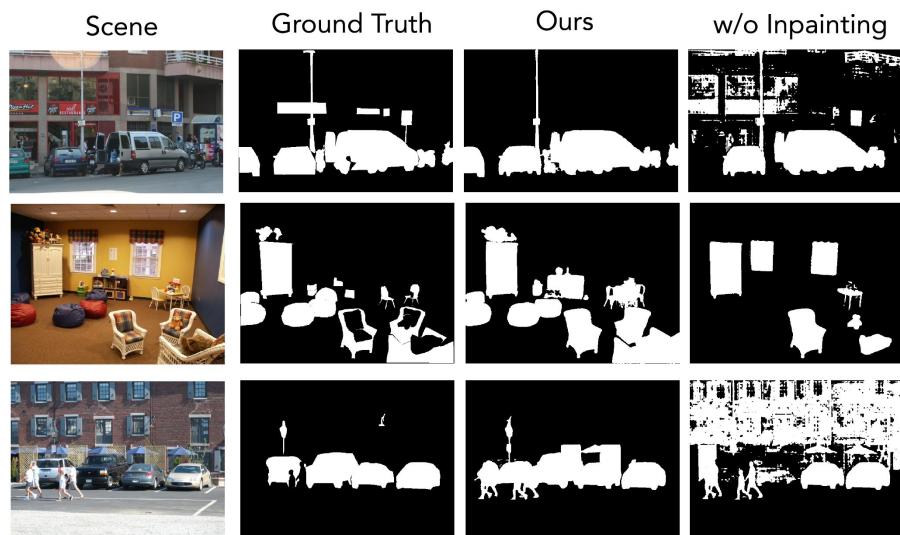


Results: Outdoor Scenes (ADE20K)

Baselines:

- Ours (w/o iterative object removal)

Method	Average IoU
Ours (things)	0.33
w/o obj. removal (things)	0.29
Ours (things + stuff*)	0.28
w/o obj. removal (things + stuff*)	0.24

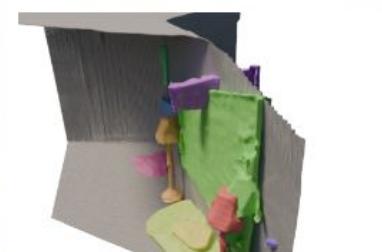
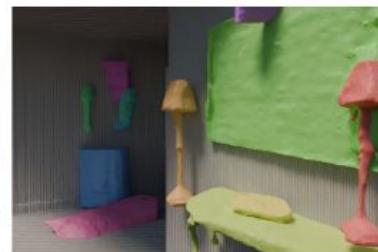


Qualitative Results: Indoor Scenes

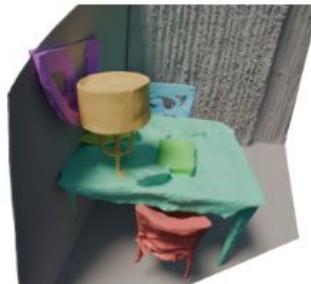
Input



Structured 3D Reconstruction



Qualitative Results: Indoor Scenes

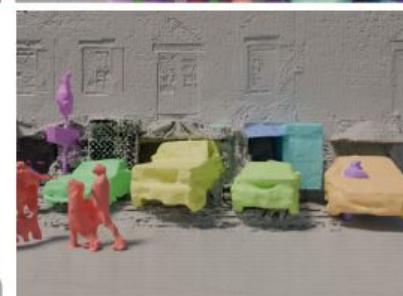


Qualitative Results: Outdoor Scenes

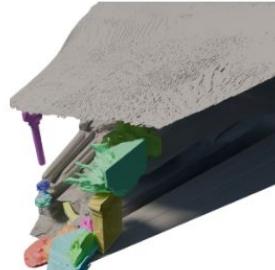
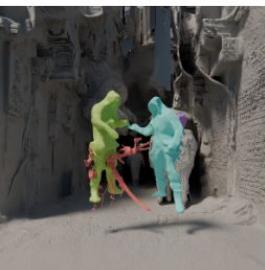
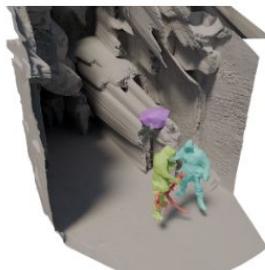
Input



Structured 3D Reconstruction



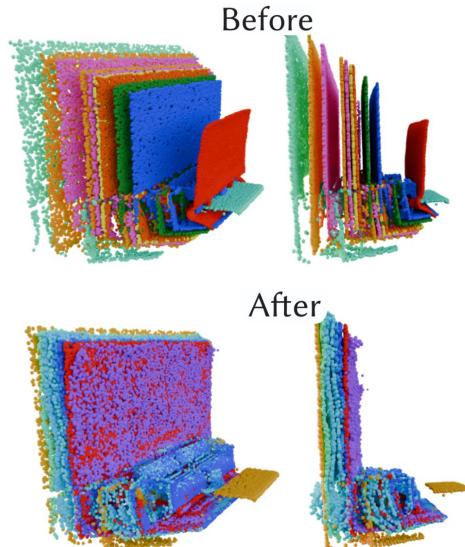
Qualitative Results: Text2Img

Prompt	Input	Structured 3D Reconstruction
<p><i>Moody fantasy illustration of a wizard and his ancient library.</i></p>		 
<p><i>Underwater ruins draped in coral and kelp, a carved sea-god relief partly obscured, shafts of turquoise light</i></p>		  
<p><i>Cyberpunk neon alley at midnight, two hooded operatives exchanging data behind steaming vents and tangled chrome cables</i></p>		  

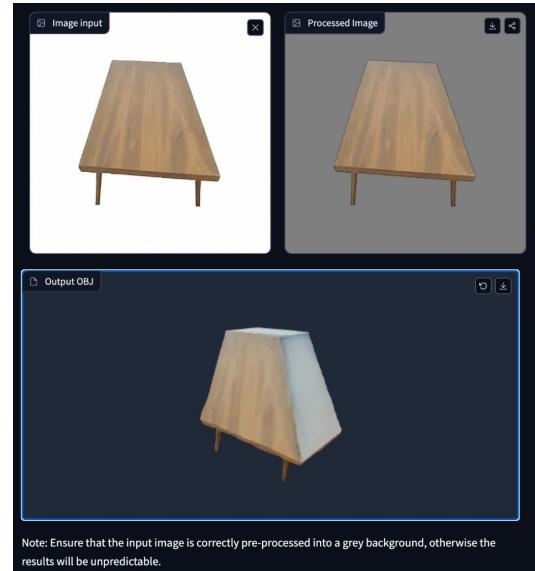
Limitations & Future Work



Artifacts Produced by
Inpainting



Depth alignment
==
Heuristic solution



Camera-view
centric object
generation

A quick thank you to my collaborators & mentors!

Matheus, Thibault, Kevin, Vova and other mentors at Adobe I met along the way... :)

