

Soft Shadow Diffusion (SSD): Physics-inspired Learning for 3D Computational Periscopy

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Can we see (in 3D) by looking at soft shadows?



We demonstrate passive 3D non-line-of-sight (NLOS) imaging—or 3D Computational Periscopy—from photographs of soft shadows cast on a visible surface.

Prior approaches to 3D NLOS imaging:

- Active methods:** Rely on time-resolved measurements of the hidden scene in response to ultrashort laser pulses. They have long acquisition times and use expensive and complex optical hardware setups.
- Passive methods:** Rely on shadows of known simple occluders (e.g., a doorway).

Our contributions:

- Reformulation of NLOS imaging to enable 3D reconstructions for hidden occluders.
- Physics-based approach based on alternating minimization.
- Physics-inspired Soft Shadow Diffusion model which facilitates high-resolution 3D reconstructions from a soft shadow photograph.

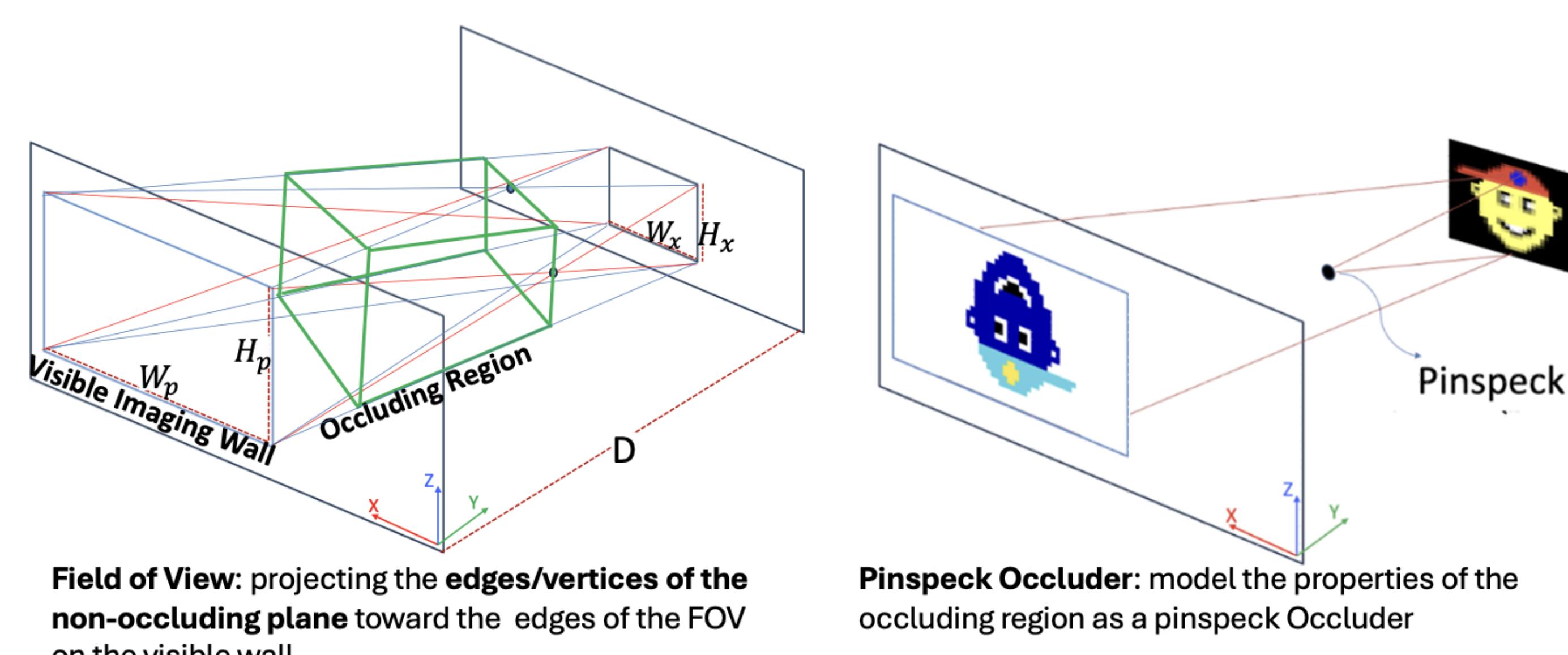
Separable Non-Linear Least Square Formulation

Generally, the brightness $i(\mathbf{p})$ of a patch \mathbf{p} on the visible wall surface is:

$$i(\mathbf{p}) = \int_{\mathbf{x} \in \mathcal{X}} \frac{g(\mathbf{p}, \mathbf{x})}{\|\mathbf{x} - \mathbf{p}\|_2^2} v(\mathbf{x}, \mathbf{p}; \theta_0) f(\mathbf{x}) d\mathbf{x} + b(\mathbf{p}),$$

where $f(\mathbf{x})$ is the brightness of some *non-occluding* hidden scene patch \mathbf{x} , $v(\cdot)$ models the visibility between \mathbf{x} and \mathbf{p} due to an unknown hidden 3D occluder represented by θ_0 , and $b(\cdot)$ ambient/visible side illumination.

Computational field of view: Hidden region that can be reliably reconstructed, for a measurement FOV



Hidden Region = Non-occluding sub-region (\mathbf{f}) & Occluding sub-region (θ_0)

Discretizing the integral yields a separable non-linear least square problem:

$$\mathbf{y} = \mathbf{A}(\theta_0)\mathbf{f} + \mathbf{b},$$

where \mathbf{y} is a soft shadow photograph, \mathbf{f} is a (discrete) representation of the non-occluding sub-region's brightness, θ_0 is representation of hidden 3D occluder, and \mathbf{b} is background.

Pure Physics with Voxelized 3D Occupancy Grid

By representing the occluding sub-region as being composed of discrete voxels (pinholes/antipinholes), and the non-occluding sub-region as discrete pixel patches, we can formulate a regularized LS problem for reconstructing (\mathbf{f}, θ_0) :

$$\arg \min_{(\mathbf{f}, \theta_0)} \|\mathbf{A}(\theta_0)\mathbf{f} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{f}\|_2^2.$$

An alternating minimization solution proceeds as:

$$\begin{aligned} \widehat{\mathbf{f}}^k &= \arg \min_{\mathbf{f}} \|\mathbf{A}(\widehat{\theta}_0^{k-1})\mathbf{f} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{f}\|_2^2 = (\mathbf{A}^\top(\widehat{\theta}_0^{k-1})\mathbf{A}(\widehat{\theta}_0^{k-1}) + \lambda \mathbf{I})^{-1} \mathbf{A}^\top(\widehat{\theta}_0^{k-1})\mathbf{y} \\ \widehat{\theta}_0^k &= \arg \min_{\theta_0} \|\mathbf{A}(\theta_0)\widehat{\mathbf{f}}^k - \mathbf{y}\|_2^2. \end{aligned}$$

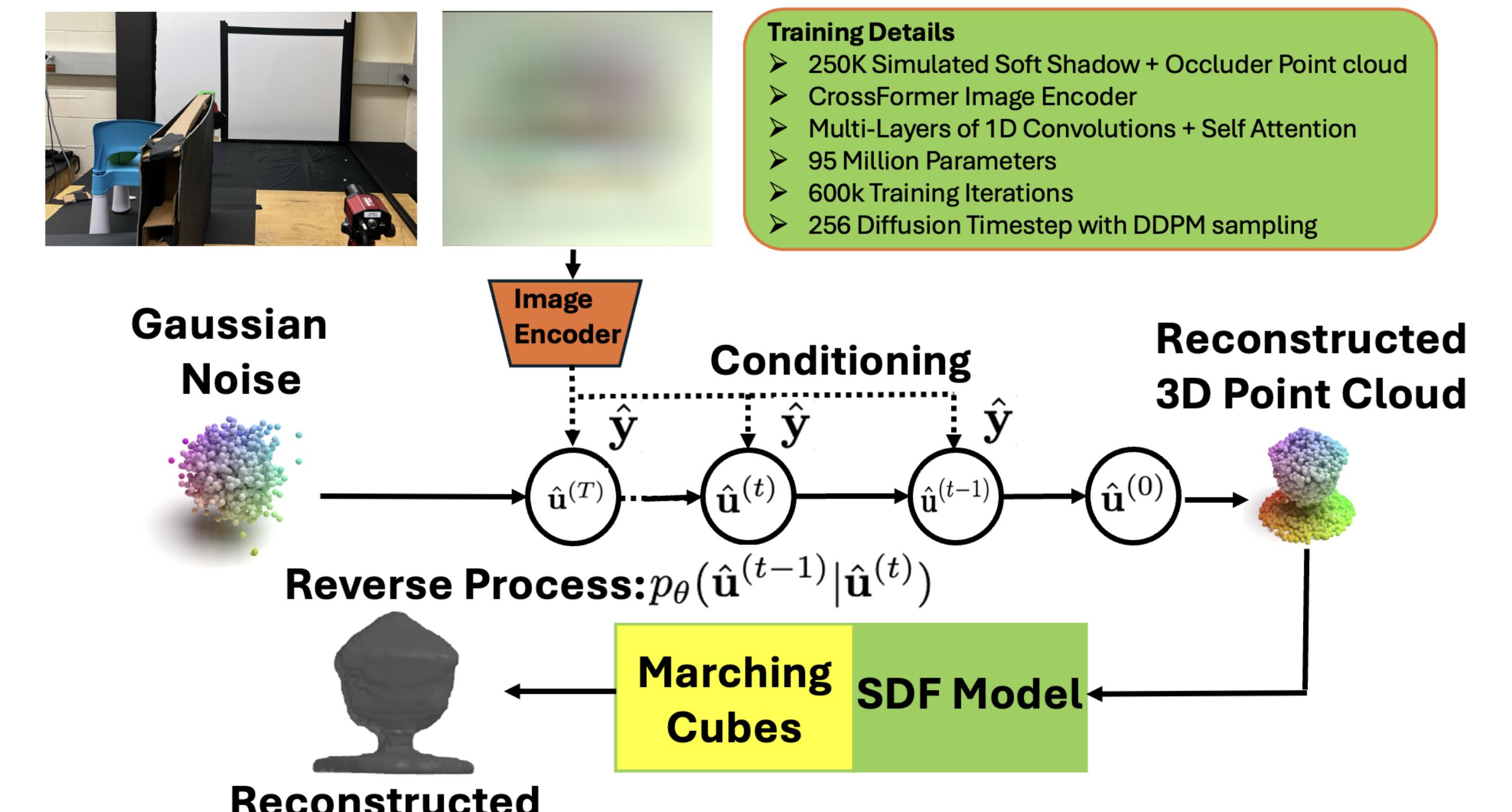
Drawback: 3D voxel-based representation is limited by voxel grid density (highly redundant representation), yet a dense grid increases computational complexity.

Soft Shadow Diffusion (SSD) Model

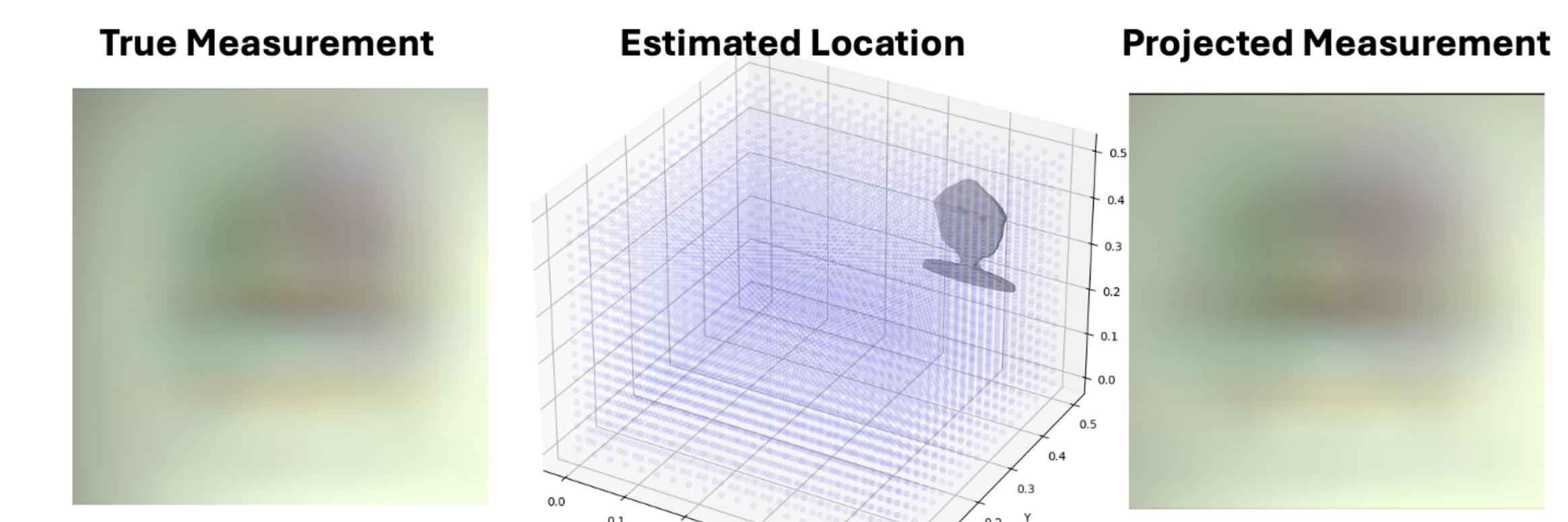
Promotes computational tractability as a learned lower dimensional representation for θ_0 and acts as an implicit prior for 3D structures. By conditioning on a soft shadow photograph, we train our novel soft shadow diffusion model to reconstruct a 3D point cloud of an occluding hidden scene object.

Intuition: From the LS-regularized problem above, we can reconstruct θ_0 without knowing \mathbf{f} , as $\widehat{\theta}_0 = \arg \min_{\theta_0} \|\mathbf{A}^\top(\theta_0)(\mathbf{A}^\top(\theta_0)\mathbf{A}(\theta_0) + \lambda \mathbf{I})^{-1} \mathbf{A}^\top(\theta_0) - \mathbf{I}\mathbf{y}\|_2^2$.

1. Reconstruct shape of θ from y via SSD:



2. Estimate location of the reconstructed occluder shape by grid search (over hidden region):

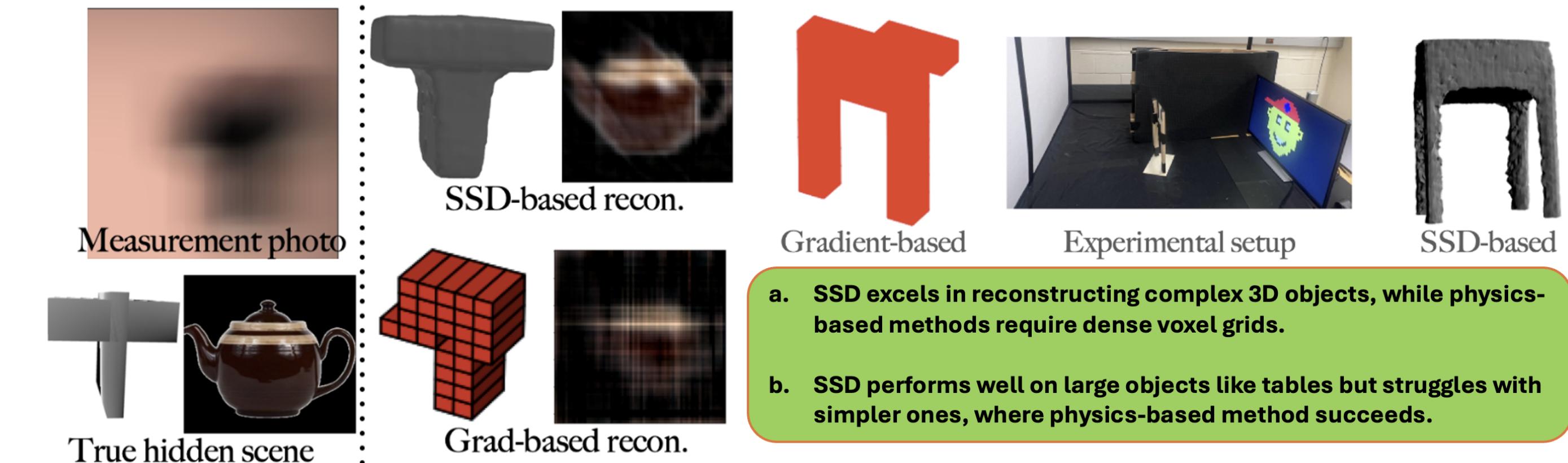


3. Reconstruct non-occluding hidden scene in 2D f by solving a TV-regularized inverse problem:

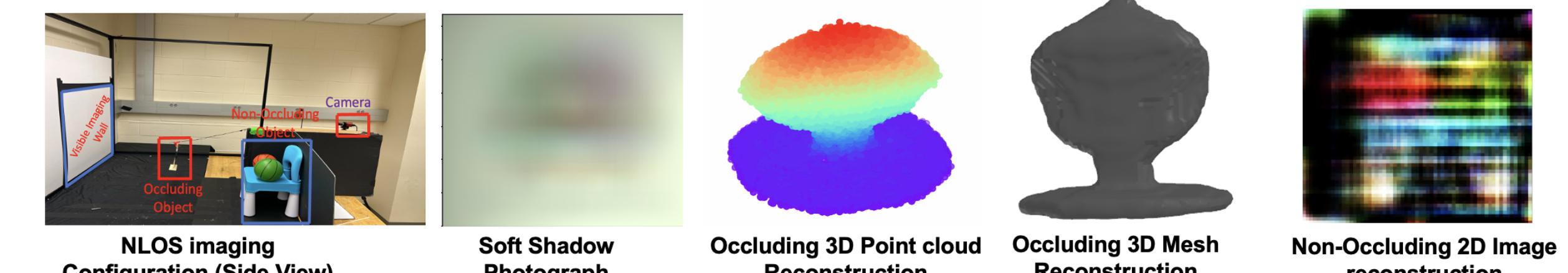
$$\widehat{\mathbf{f}} = \arg \min_{\mathbf{f}} \|\mathbf{y} - \mathbf{A}(\widehat{\theta}_0)\mathbf{f}\|_2^2 + \lambda \|\mathbf{f}\|_{TV}^2.$$

Results

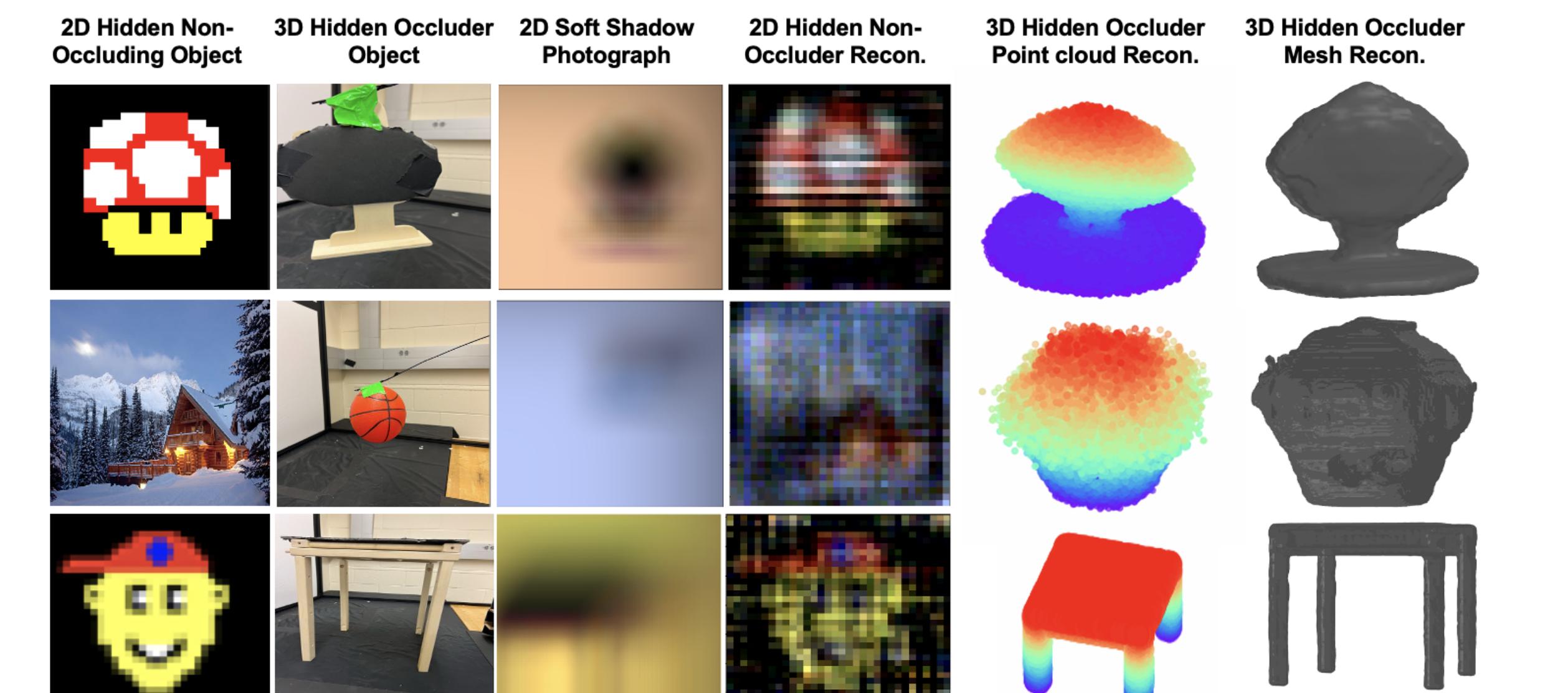
Soft Shadow Diffusion Vs Physics-Based Reconstruction



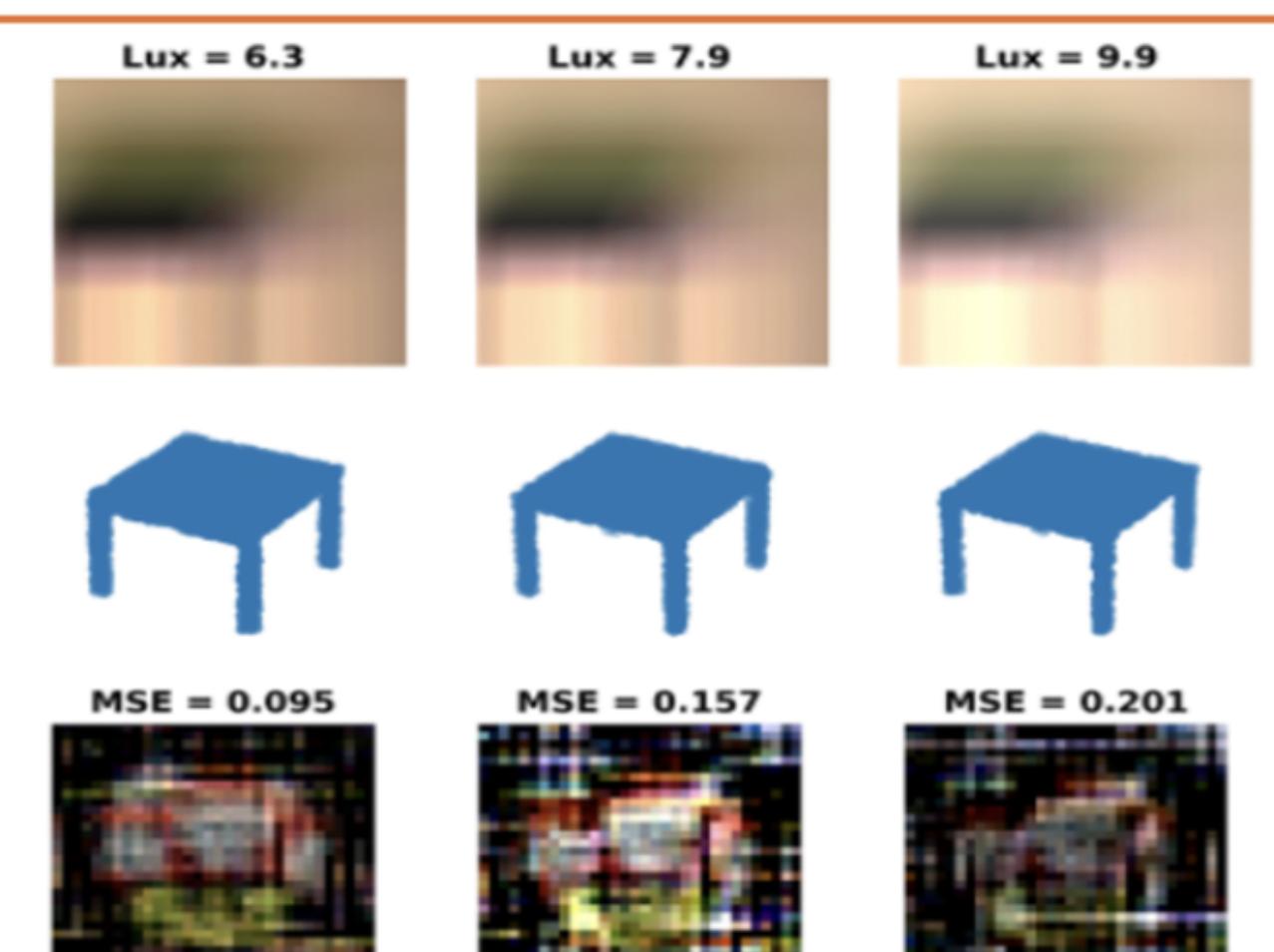
Soft Shadow Diffusion on real 3D light reflective object



SSD-based reconstruction



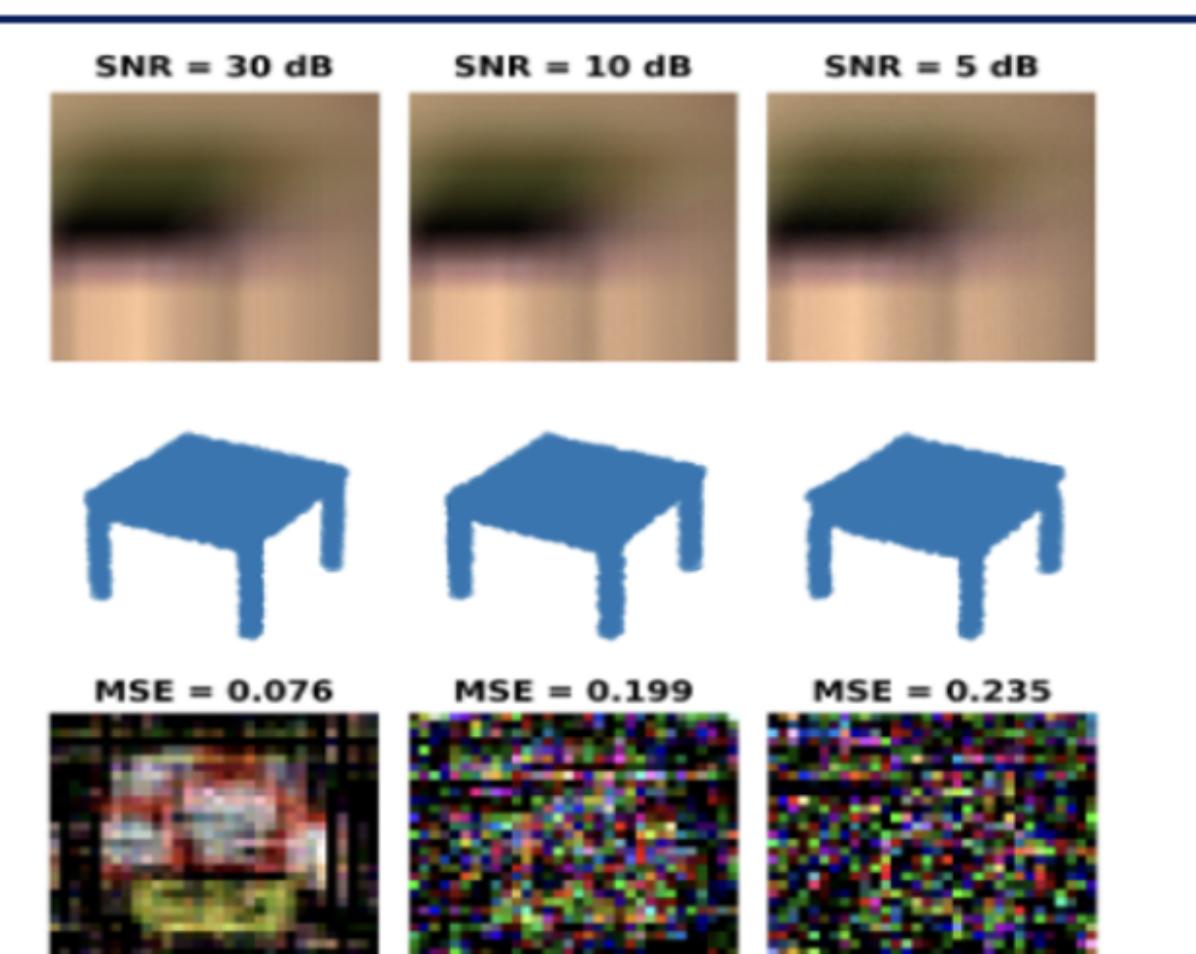
Reconstructions for increasing ambient light levels



True hidden scene setup



Reconstructions for increasing noise levels



- SSD is robust to background light, allowing reconstruction of non-occluded objects at different light levels.
- SSD shows robustness to noise, as it degrades gracefully with increasing noise levels.