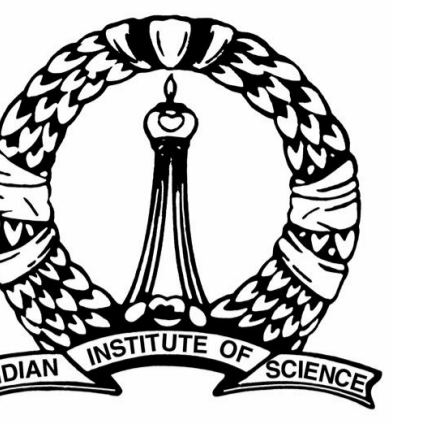




High Quality Photometric Reconstruction using a Depth Camera

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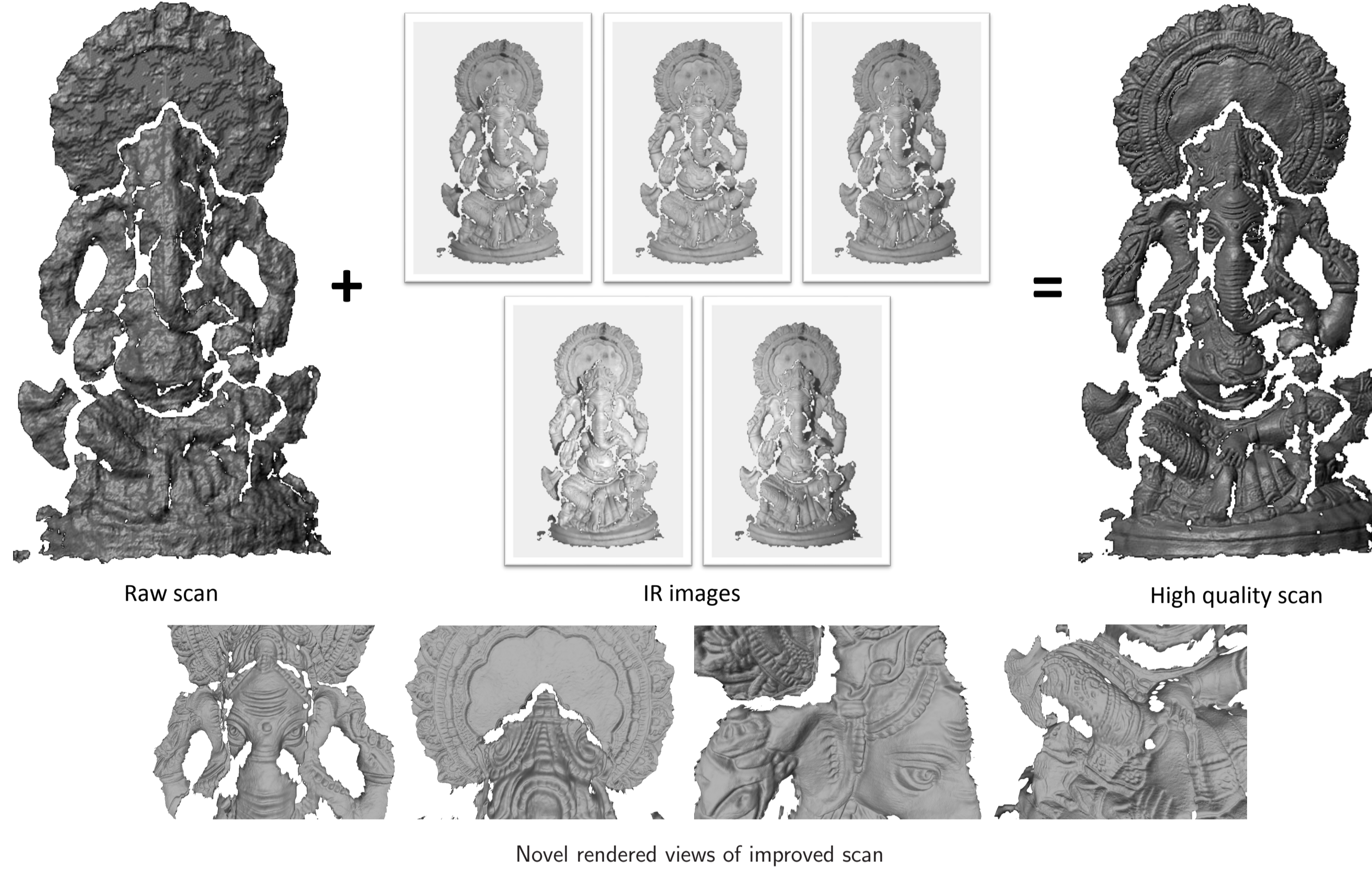
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Introduction

Consumer Depth Cameras (Kinect etc.):

- Advantages: Inexpensive, portable, easy to use, real-time depth acquisition
- Disadvantages: Highly noisy depth scans, fine-scale details are not captured
- Solution: Recover fine-scale details from photometry using same depth camera
- Novelty: Our is first method to use Kinect as depth and radiometric camera



Advantages of using Kinect's Infrared Camera

- High resolution of 1280×960 pixels
- High dynamic range (10 bit)
- Almost linear photometric response
- Depth and IR image in same reference frame: no uncommon occlusion
- No interference due to indoor lighting and a halogen lamp acts as a good source of IR

A Novel Approach to Depth and Intensity Imaging

- Through software Kinect's IR projector turned on (for depth imaging) and off (for intensity imaging)
- Exposure of IR camera lowered through software for very close range (about 40cm) depth scanning
- Light sources placed far enough to achieve parallel lighting

Depth-Normal Fusion

$Z(x, y)$: raw depth map; $N(x, y)$: photometric normal map; $\hat{Z}(x, y)$: estimated depth map

Relation between 3D points and depth map is given by

$$P(x, y) = \left[-\frac{x}{f}Z(x, y) \quad -\frac{y}{f}Z(x, y) \quad Z(x, y) \right]^T$$

with f being focal length of IR camera

Improved depth map is estimated as

$$\hat{Z} = \arg \min_{\hat{Z}} E(\hat{Z}) = \arg \min_{\hat{Z}} [E_d(\hat{Z}) + \lambda_n E_n(\hat{Z}) + \lambda_s E_s(\hat{Z})]$$

where

Depth penalty:

$$E_d(\hat{Z}) = \sum_p w_p \|\mu_p\|^2 (Z_p - \hat{Z}_p)^2$$

with $\mu_p = \left[-\frac{x}{f} \quad -\frac{y}{f} \quad 1 \right]^T$ and w_p is adaptive weighting, computed from the eigen values of local structure-tensor of photometric normal map

Normal penalty:

$$E_n(\hat{Z}) = \sum_p (N_p \cdot T_{x,p})^2 + (N_p \cdot T_{y,p})^2$$

where surface tangents T_x and T_y are given by

$$T_x = \frac{\partial P}{\partial x} = \left[-\frac{1}{f} \left(\hat{Z} + x \frac{\partial \hat{Z}}{\partial x} \right) \quad -\frac{1}{f} \frac{\partial \hat{Z}}{\partial x} \quad \frac{\partial \hat{Z}}{\partial x} \right]^T$$

$$T_y = \frac{\partial P}{\partial y} = \left[-\frac{1}{f} \frac{\partial \hat{Z}}{\partial y} \quad -\frac{1}{f} \left(\hat{Z} + y \frac{\partial \hat{Z}}{\partial y} \right) \quad \frac{\partial \hat{Z}}{\partial y} \right]^T$$

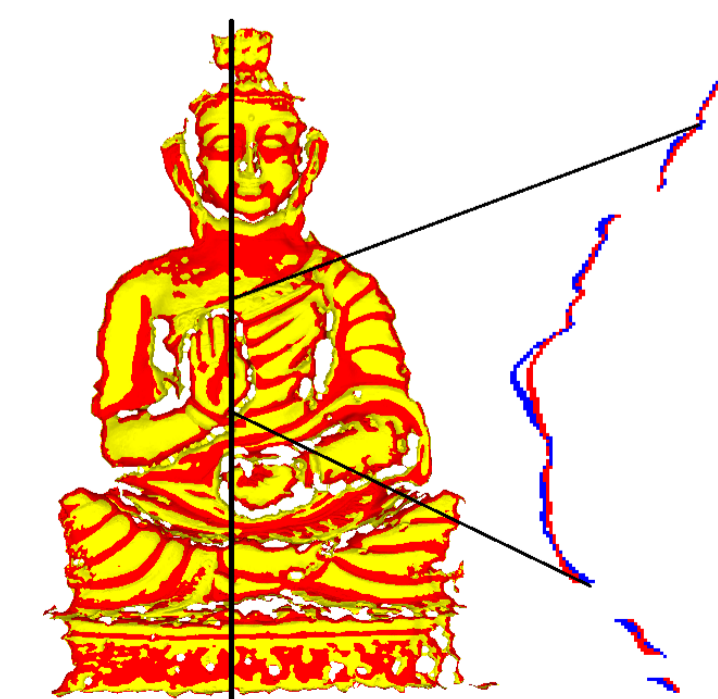
where depth derivatives are computed by adaptively weighting the forward and backward differences depending on the similarity of the estimated normals (instead of estimated depths) of the pixels. Similarity between p^{th} and q^{th} pixel is $w(p, q) = \exp^{-\frac{1}{2\sigma^2}(1-N_p^T N_q)}$

Smoothness penalty:

$$E_s(\hat{Z}) = \nabla^2(\hat{Z})$$

($\nabla^2(\hat{Z})$) is discrete version of the conventional Laplacian penalty

Multi-view Reconstruction



Non-rigid deformation from depth-normal fusion

- Depth-Normal fusion causes independent non-rigid deformations of scans
- ICP-based scan registration also has small alignment error
- Both cause blurring and loss of details after merging
- Solved by recovering the normals from individual scans by ray shooting
- Averaging normal estimates from different scans also results in blurring
- Solved by selecting only one normal using a priority ordering to preserve local coherence

Result: Ganesha Model

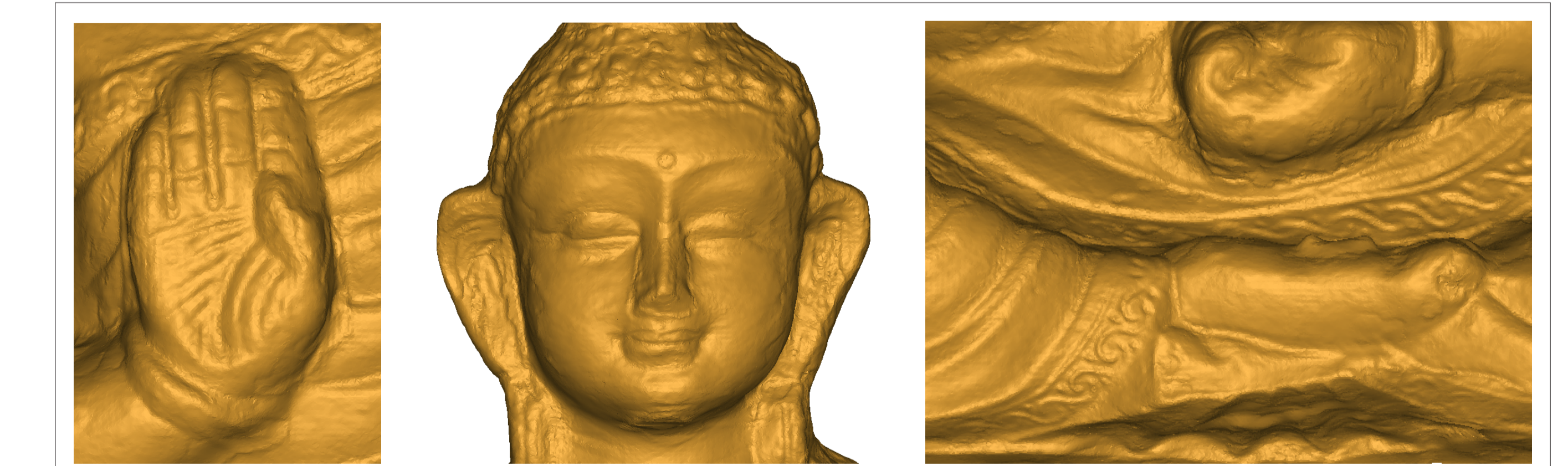


Result: Buddha Model



Reconstruction with raw kinect depth map

Our reconstruction

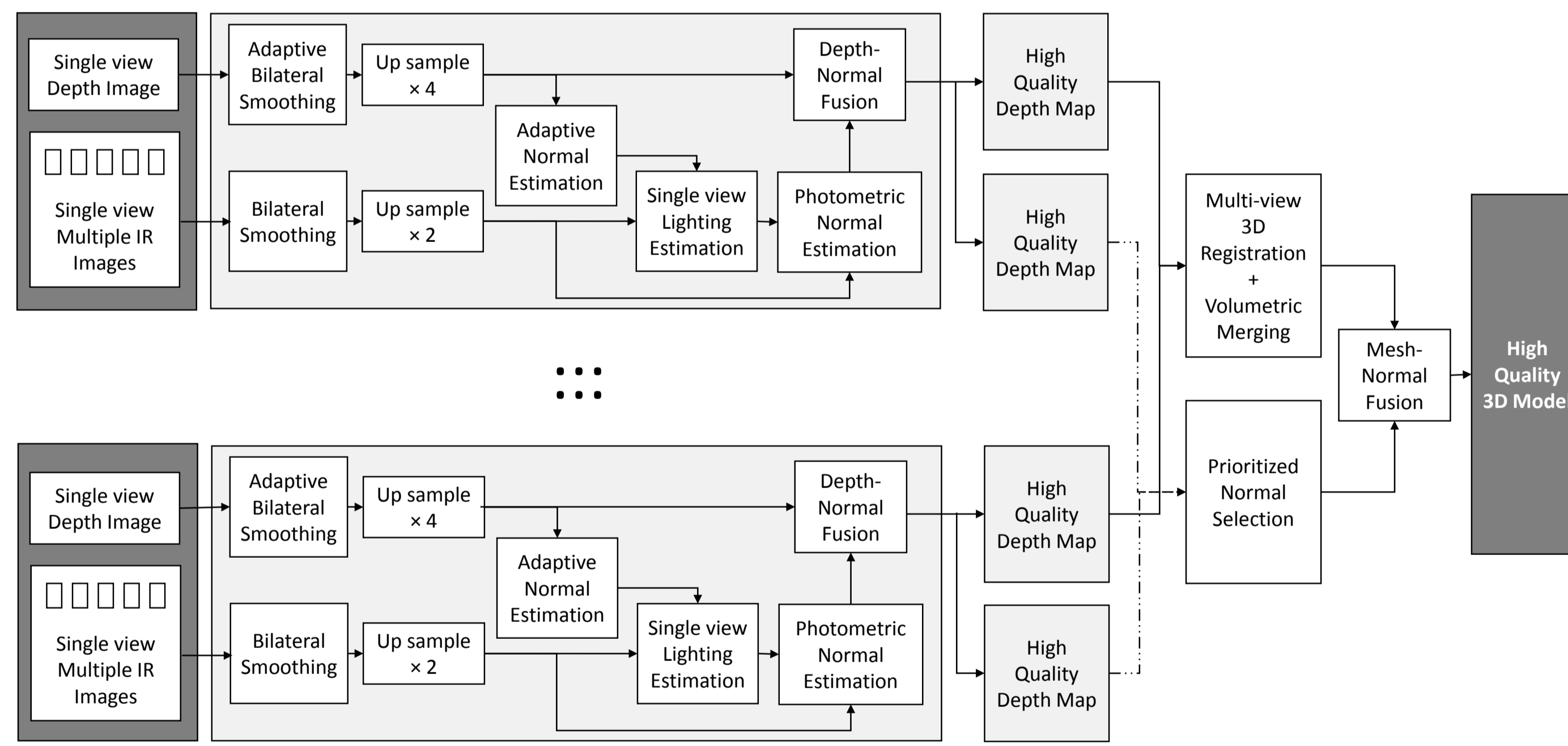


Novel rendered views of our reconstruction

Conclusion

- An approach to combine a raw depth map with photometric normal map is presented
- A novel method to use the same depth camera for both depth and radiometric imaging is developed
- Automatic calibration of lighting directions is addressed
- Novel adaptive weighting and differentiation schemes are described
- A multi-view reconstruction method with coherent normal selection is developed
- Preserves fine-scale details after registration and merging

Proposed Pipeline



Depth-map Guided Photometric-stereo

Linear lighting model:

I_p : intensity at pixel p ; A : illumination strength; α : albedo; N_p : surface normal; V : lighting direction. For Lambertian reflectance

$$I_p = A\alpha N_p \cdot V$$
$$\Rightarrow I_p^k = N_p \cdot S^k$$

with $S = A\alpha V$ (superscript k indexes different lighting conditions)

Linear lighting estimation:

Using estimate of surface normal N_p obtained from raw depth map

$$S^k = \arg \min_x \sum_p \rho(I_p^k - N_p \cdot x)$$

with $\rho(\cdot)$ being a robust loss function

Photometric normal estimation:

$$N_p = \arg \min_x \sum_k \rho(I_p^k - S^k \cdot x)$$