# Neural Rendering for Stereo 3D Reconstruction of **Deformable Tissues in Robotic Surgery**



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## Introduction

#### **Background:**

Surgical reconstruction plays an important role in Minimally Invasive Surgery (MIS), since it illuminate downcan many stream tasks:

#### Stereo Videos



### **Experimental Results**

**Experiment Setup:** Evaluate on robotic surgery stereo videos from 6 cases of our in-house DaVinci robotic prostatectomy data. We choose **E-DSSR** (MICCAI'21) as a strong comparison.

**Quantitative Results (Photometric errors):** 

- □ Intra-operative navigation.
- **Given Surgical planning**.
- **Context-awareness in MIS.** □ Surgery education. AR guidance in MIS.
- □ Surgical visualization.

### 3D Shapes

#### **Challenges:**

- Surgical scenes will undergo large deformations.
- Endoscopic videos show sparse viewpoints due to constrained movement in confined spaces.
- The surgical tools occlude part of the soft tissues. 3.

## Method The proposed method:



Metrics	Ours	E-DSSR
PSNR	29.831 ± 2.208	13.398 ± 1.387
SSIM	0.925 ± 0.020	$0.630 \pm 0.057$

#### **Qualitative Results (Visit our project page for more):**





Adopt dynamic **neural radiance fields (NeRF)** to represent deformable surgical scenes.

Surgical Scene



- Design a new tool-mask guided ray casting for handling tool occlusion. • Reject training rays that shoot towards tool pixels.
- Incorporate **depth-cueing ray marching** and dense **depth-supervised optimization** to impose explicit geometric clues.

#### Close-ups results of "cutting tissues twice" with topology changes.



## Conclusion

**Contribution:** We have introduced our novel neural rendering-based framework for dynamic surgical scene reconstruction from singleviewpoint binocular captures, addressing complex tissue deformations and tool occlusion.



• Concentrate points around tissue surface indicated by stereo depth. Add loss between rendered optical depth and estimated stereo depth. • Statistically refine estimated depth to patch corrupt depth.

**Applications: 1)** AR/VR immersive surgery education. 2) Robotic surgery simulation.



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