# **MPR-Diff: a Self-Supervised Diffusion Model for Multi-Planar Reformation in Prostate Micro-Ultrasound Imaging**

Kaifeng Pang<sup>1,3</sup>, Qi Miao<sup>1</sup>, Alex Ling Yu Hung<sup>1,4</sup>, Kai Zhao<sup>1</sup>, Eunsun Oh<sup>1</sup>, Raymi Ramirez<sup>1</sup>, Wayne Brisbane<sup>2</sup>, Kyunghyun Sung<sup>1</sup> <sup>1</sup>Department of Radiological Sciences <sup>2</sup>Department of Urology <sup>3</sup>Department of Electrical and Computer Engineering <sup>4</sup>Department of Computer Science University of California, Los Angeles



# **Poster No. 1571074699**

#### Introduction

- > Micro-ultrasound is a novel 29-MHz ultrasound imaging with low cost and  $3-4 \times$  higher resolution than traditional ultrasound.
- > Acquisitions are fan-shaped with non-uniform angular intervals.
- > Multi-planar reformation (MPR) is crucial for joint evaluation with MRI, pathology, etc.
- > MPR is challenging due to low anterior resolution and discontinuity between adjacent slices.





#### Contributions

- > The first deep learning-based approach for micro-ultrasound MPR.
- A self-supervised learning strategy without the need for high-resolution ground truth.
- **Improved image quality**, validated by both quantitative metrics and expert reader studies.

#### Results

#### **Quantitative Results**

## Method

#### **Conditional DDPM**

> Polar coordinate system: represent each voxel in the MPR plane using polar coordinates  $(\theta, r)$  to reflect the fan-shaped acquisition. **Up-sampling strategy:** super-resolve low-resolution fan-shaped image  $y \in \mathbb{R}^{H \times N}$  to  $x \in \mathbb{R}^{H \times kN}$  with up-scaling factor k. **Conditional generation:** a DDPM is trained to generate the highquality image conditioned on the degraded image y, and the corresponding polar coordinates c.

#### > Best sharpness and perceptual quality!

Method	Sharpness [14] (×10 <sup>2</sup> ) $\uparrow$	<b>PIQUE</b> [15] ↓
Bilinear [3]	4.35	26.44
SwinIR [6]	4.36	26.24
SRConvNet [12]	4.39	25.18
MPR-Diff	4.44	19.47

#### **Qualitative Comparison**

Enhanced details and boundary of lesion!



#### **Expert Reader Study**



#### **Self-supervised Training**

- > No high-quality supervision: true high-resolution target-plane images are unavailable, making direct supervised training infeasible.
- > Patch-sampling training: simulate training pairs by randomly extracting LR and HR fan-shaped patches from the original plane.
- Simulated Acquisition Coordinates: use non-uniform angular coordinates  $\{\tilde{\theta}^{lr}\}$  during sampling to mimic the realistic MicroUS acquisition patterns.

#### > Superior reader indicates p < 0.05)



#### **Ablation Study**

- $\succ$  (a) remove the positional condition c.
- $\succ$  (b) replace random sampling of  $\{\tilde{\theta}^{lr}\}$  with uniform sampling.



LR (bilinear)

(a) w/o PC

**MPR-Diff** 





**David Geffen** Health UCLA **School of Medicine** 

Personal Lab website website

#### **Conclusion and Discussion**

- Improved delineation of prostate and lesions, enhancing the clinical value of MicroUS for prostate cancer diagnosis.
- > Limitation: DDPM suffers from long inference time, limiting practical use in volumetric reformation.
- > Future work: we are currently developing a single-step conditional consistency model-based approach – fast, stable and high-quality!

### Acknowledgment

This research was funded in part by NIH R01-CA248506, NIHR01-CA272702, and the Integrated Diagnostics Program of the Departments of Radiological Sciences and Pathology in the UCLA David Geffen School of Medicine.

#### References

[1] Pensa, Jake, et al. "Evaluation of prostate cancer detection using micro-ultrasound versus mri through coregistration to whole-mount pathology," Sci. Rep., 2024. [2] Liang, Jingyun, et al. "Swinir: Image restoration using swin transformer," ICCV, 2021.

[3] Li, Feng, et al. "Srconvnet: A transformer-style convnet for lightweight image super-resolution," IJCV, 2024.