







Analysis of 3D pathology samples using weakly-supervised Al

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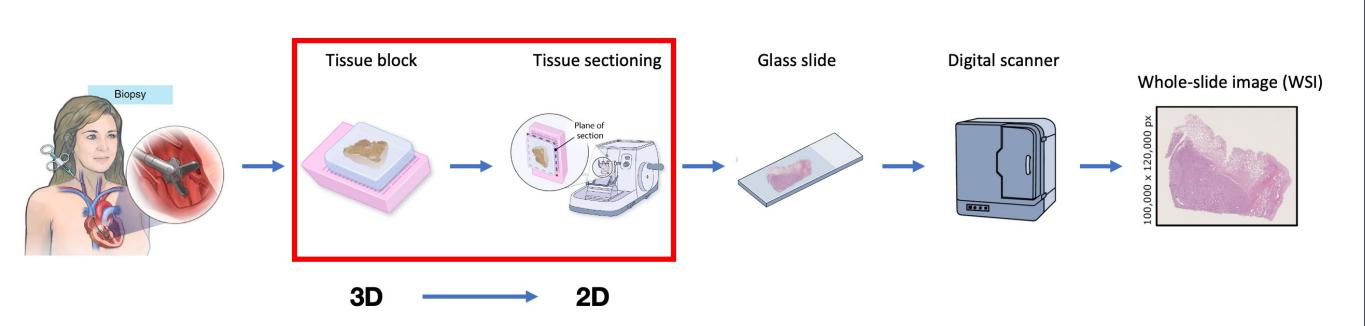
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Motivation

Human tissue is inherently 3D!

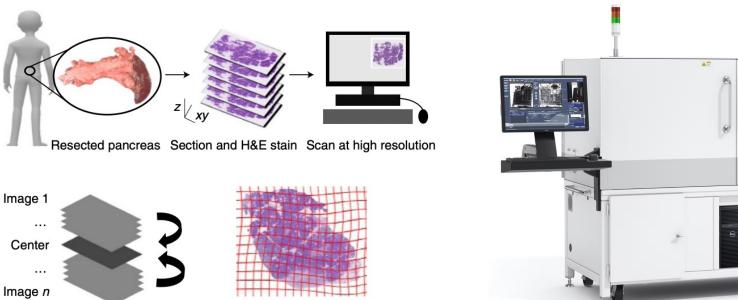


- However, the current clinical/computational practice entirely focuses on thinly-sliced 2D tissue sections [1,2]. This leads to
- ⇒ Sampling bias and misdiagnosis due to heterogeneous tissue volume
- ⇒ Mischaracterization of 3D morphological structure [3]

3D pathology paradigm can address these limitations!

The current state of 3D pathology

3D pathology requires innovations in hardware and software High-resolution 3D imaging modalities (hardware)



Serial sectioning & registration

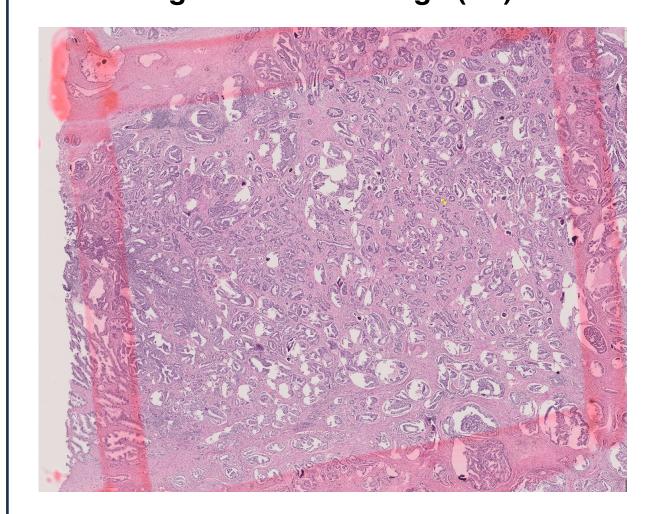
Xradia 620 Versa

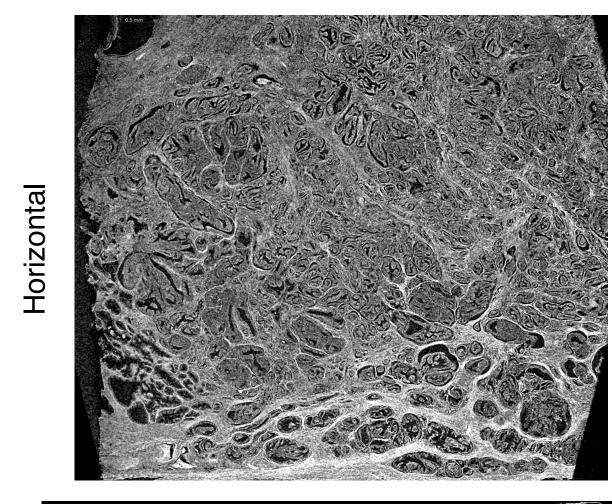
microCT

Open-top light-sheet microscopy [4]

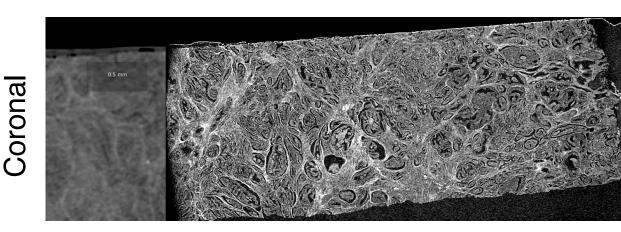
Examples of 3D pathology image (Prostate cancer)

Digitized tissue image (2D)

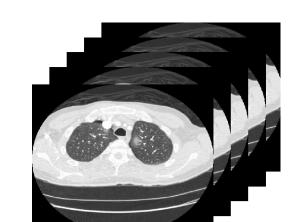




microCT image (3D)



No computational pipeline to process volumetric data exists!



Size: Mega-voxel
 Res.: 1~2 mm /voxel

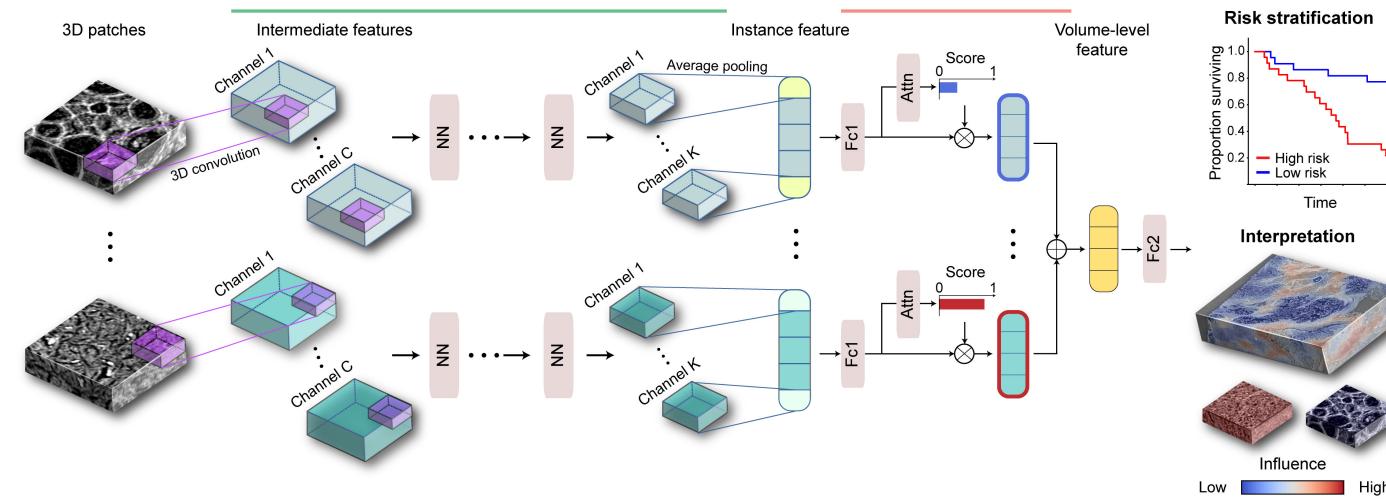
3D radiology



3D pathology

- Size: Giga ~ Tera-voxel
- **Res.:** 0.5~4 μm /voxel

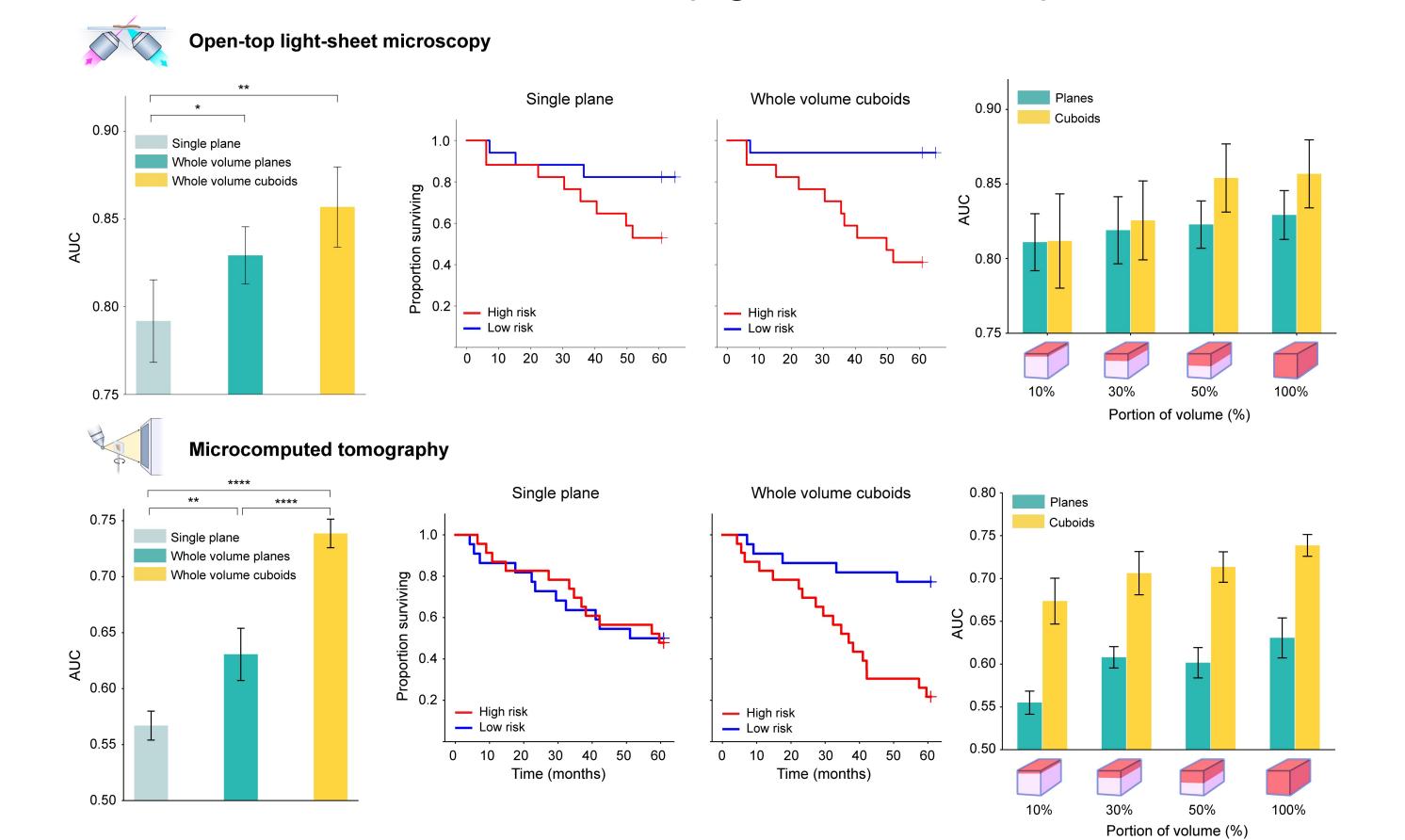
Proposed 3D workflow Blopsy block (3D) Block processing (3D to 3D) Resection Resection Raw volume Tissue segmented volume Stack of cuboids 3D patches Al-based computational processing Feature encoder Risk stratification Risk stratification



Multiple instance learning (weakly-supervised learning) in 3D

Risk stratification in Prostate cancer

Prostate cancer recurrence risk stratification (high-risk vs. low-risk)



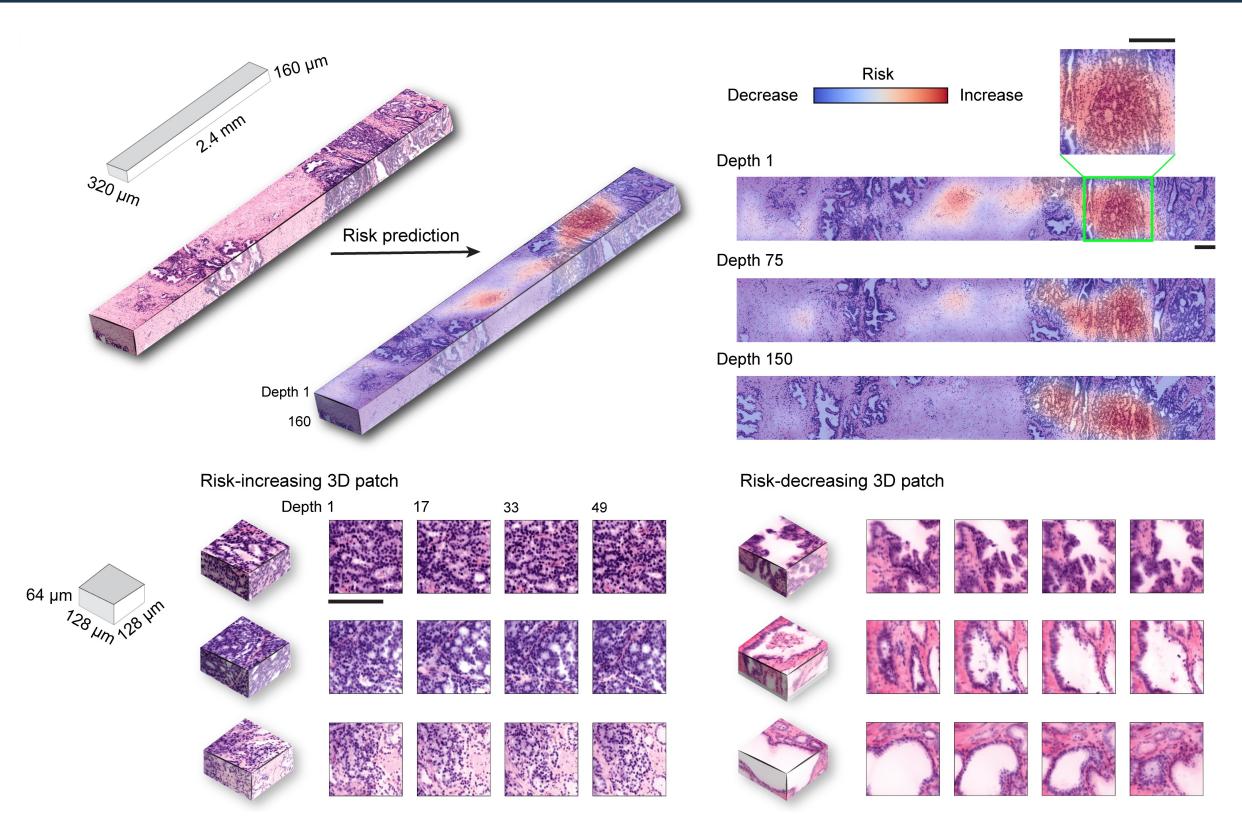
Panel of 6 pathologists Biopsy (OTLS 3 slices, 1µm/pixel) Pathologist grading (individual) Pathologist 1 (P1) 3+3 Redian Consensus 3+4 Pathologist 5 (P6) 3+4 TriPoth TriPoth O .90 O .75 O .70

Reader study with six board-certified pathologists

⇒ TriPath outperforms all clinical baselines, demonstrating clinical potential

Pathologists (consensus) TriPath (OTLS)

TriPath for Interpretability



Risk heatmap captures well-known prognostic biomarkers

- ⇒ Identifies risk-increasing/decreasing 3D patches automatically
- ⇒ With larger validation cohort, novel 3D biomarker discovery possible

References

- [1] Song AH et al., Analysis of 3D pathology samples using weaklysupervised AI. Cell, 2024
- [2] Song AH et al., Artificial intelligence for digital and computational pathology. Nature Reviews Bioengineering, 2023
- [3] Liu JTC et al., Harnessing non-destructive 3D pathology, Nature Biomedical Engineering, 2021
- [4] Bishop K et al., An end-to-end workflow for nondestructive 3D pathology, Nature Protocols, 2024



