

Learning to predict cutting angles from histological human brain sections

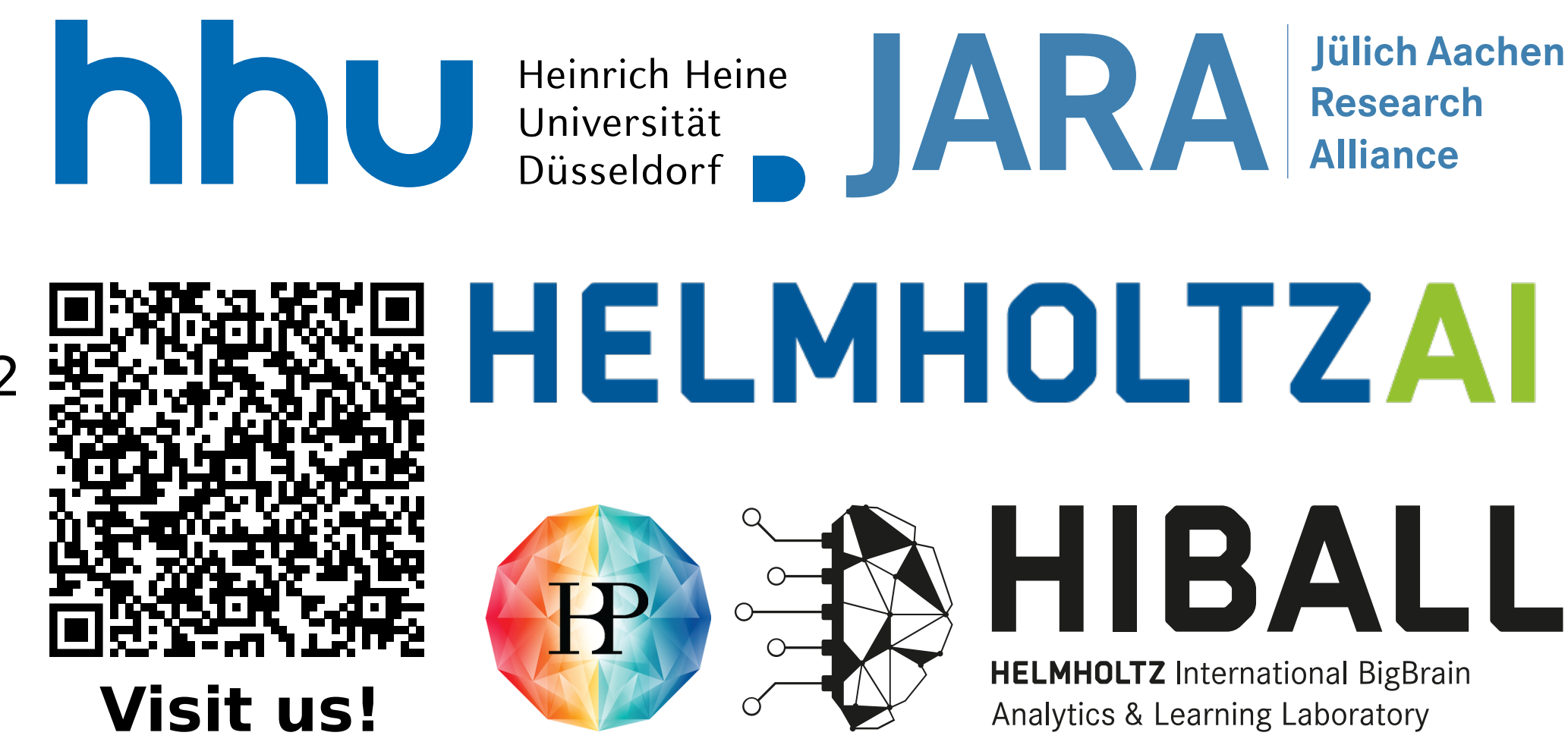
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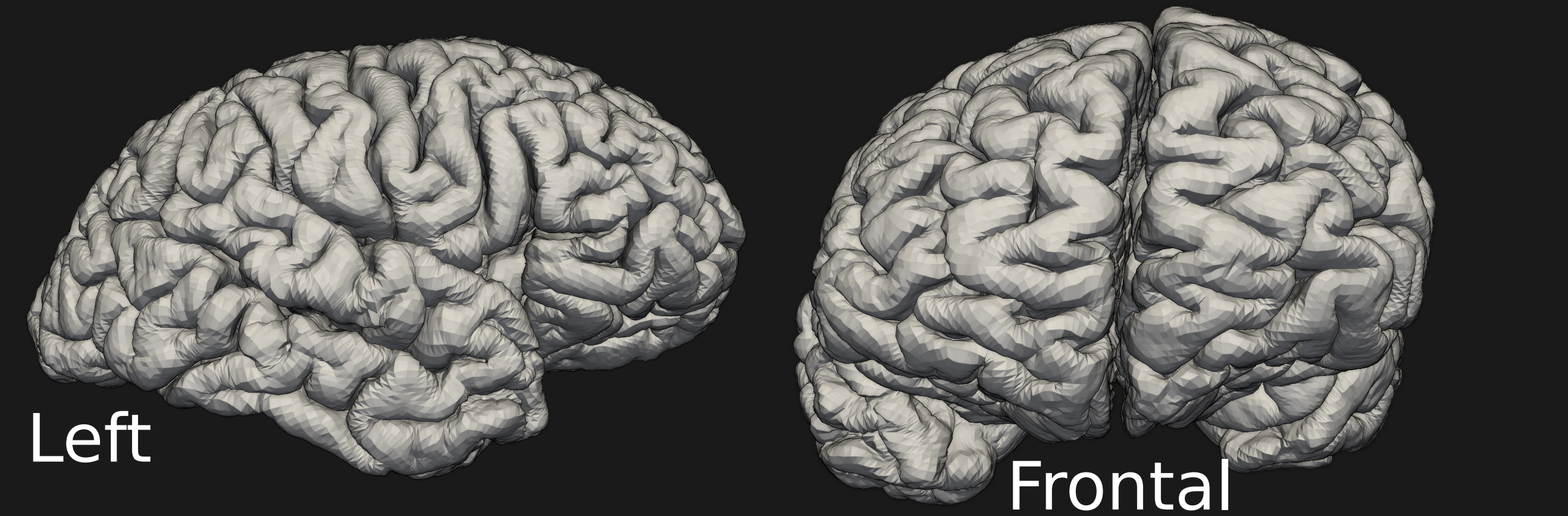
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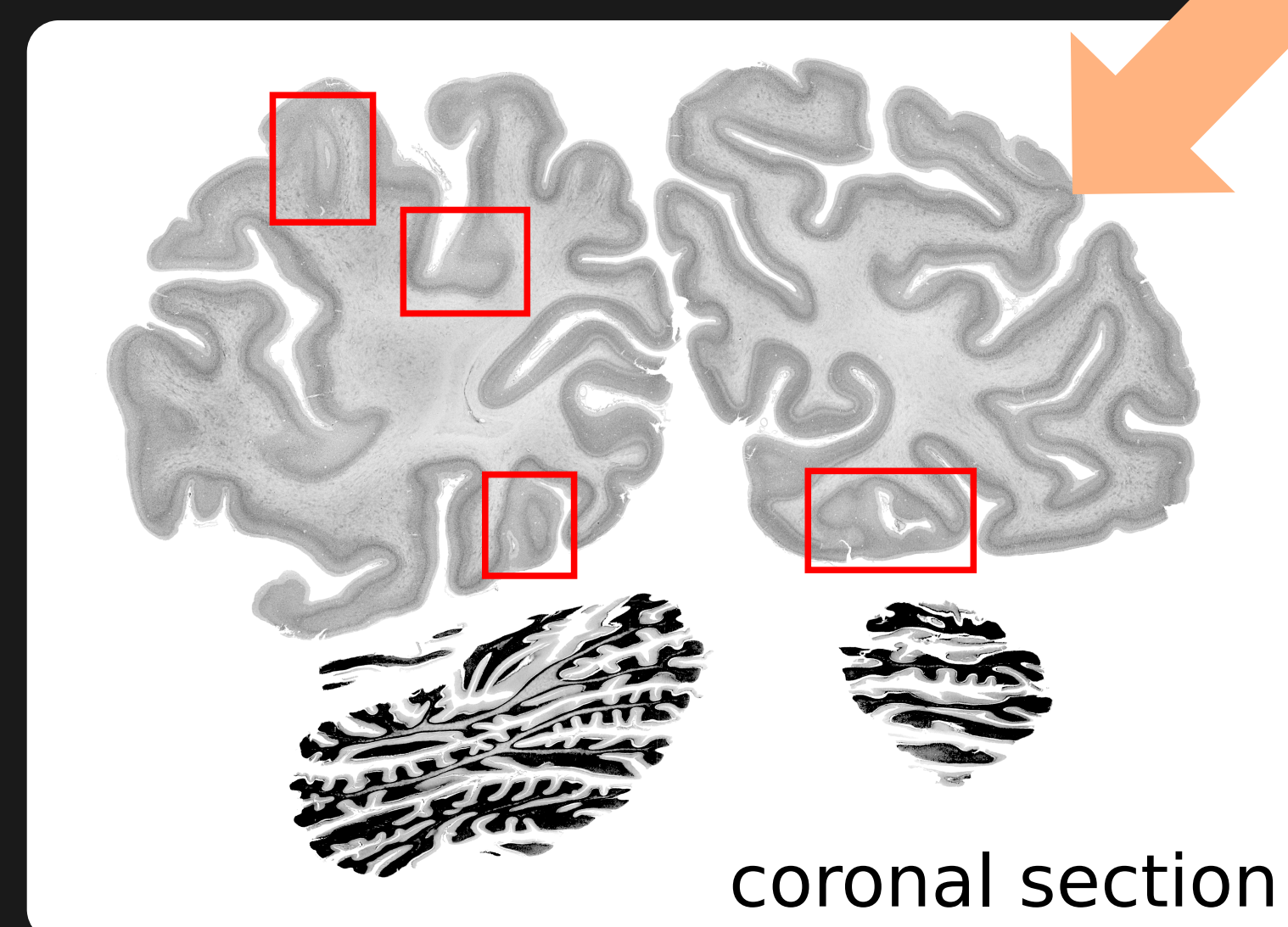
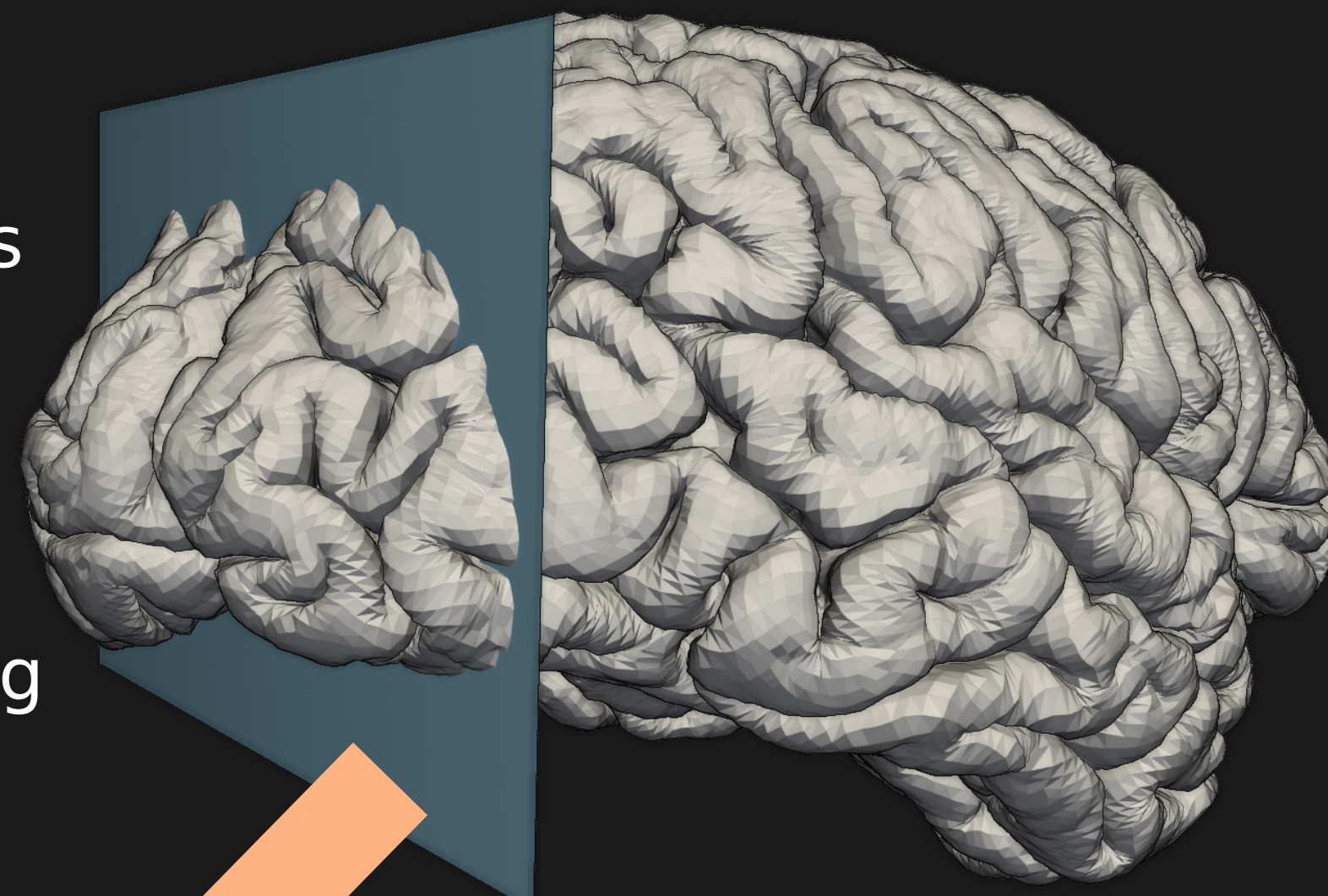
Problem

Identification of obliquely cut regions in histological human brain sections

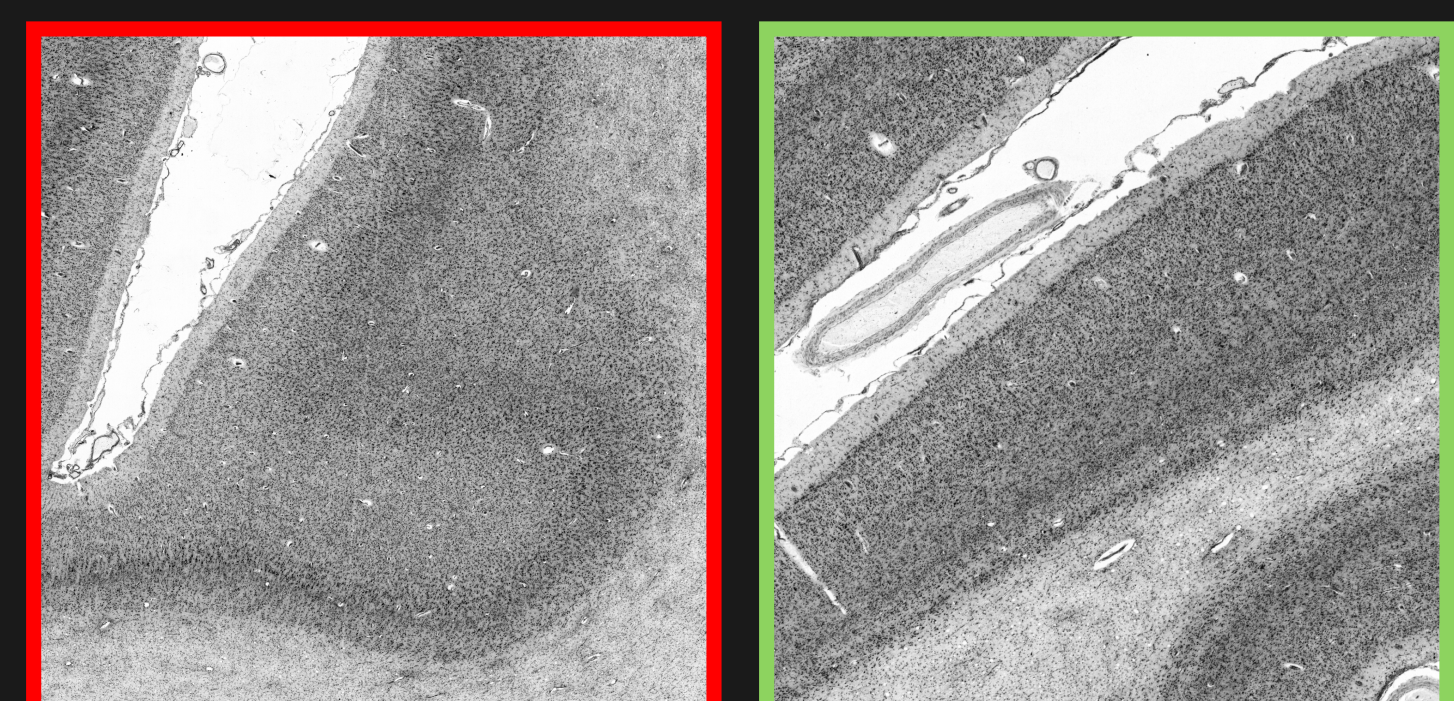


Histological processing

- Postmortem human brains
- ~7000 coronal sections
- Cell body staining
- Light microscopic scanning
- Resolution: 1µm / pixel



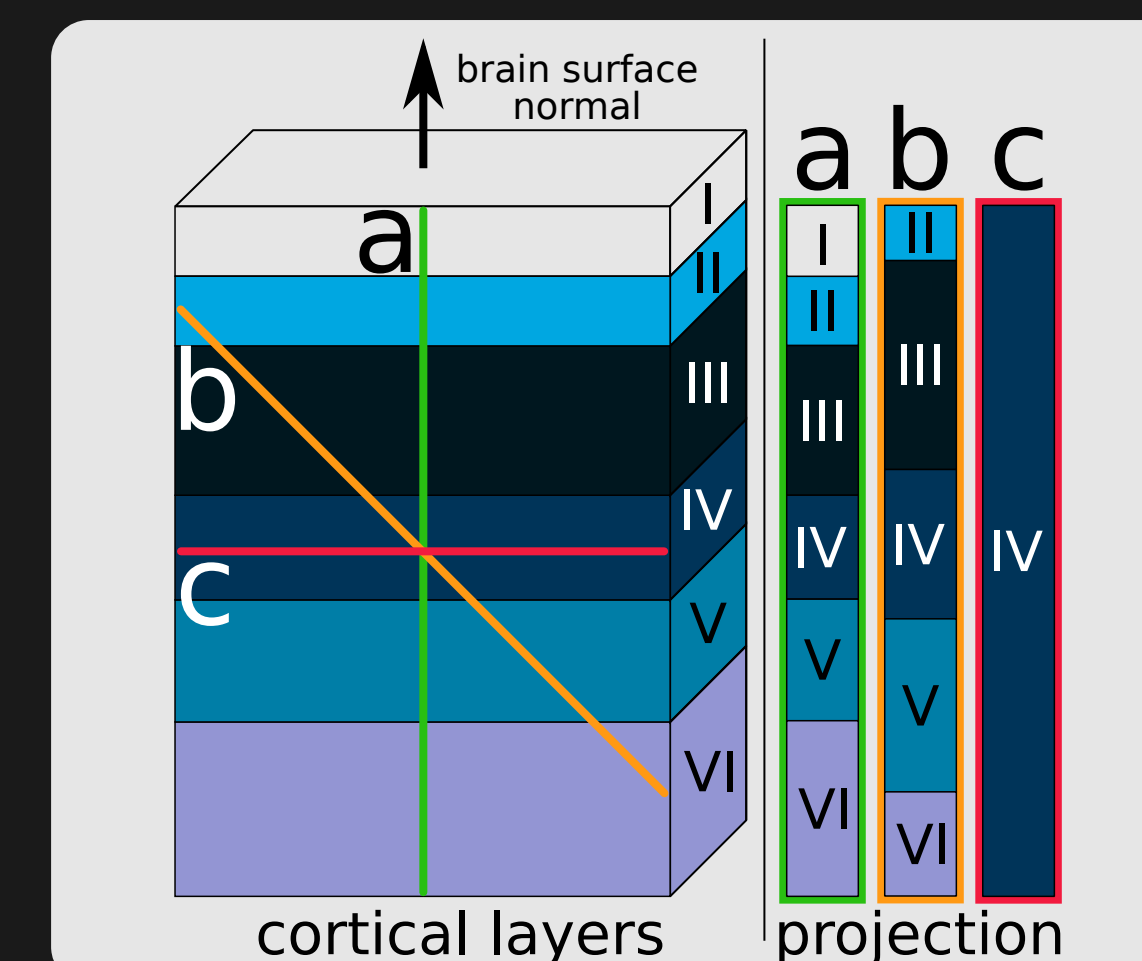
coronal section



oblique (bad) straight (good)

Challenge: Oblique cuts

- Cytoarchitecture analysis relies on cortical layers
- Layers not visible if cutting angle is too oblique

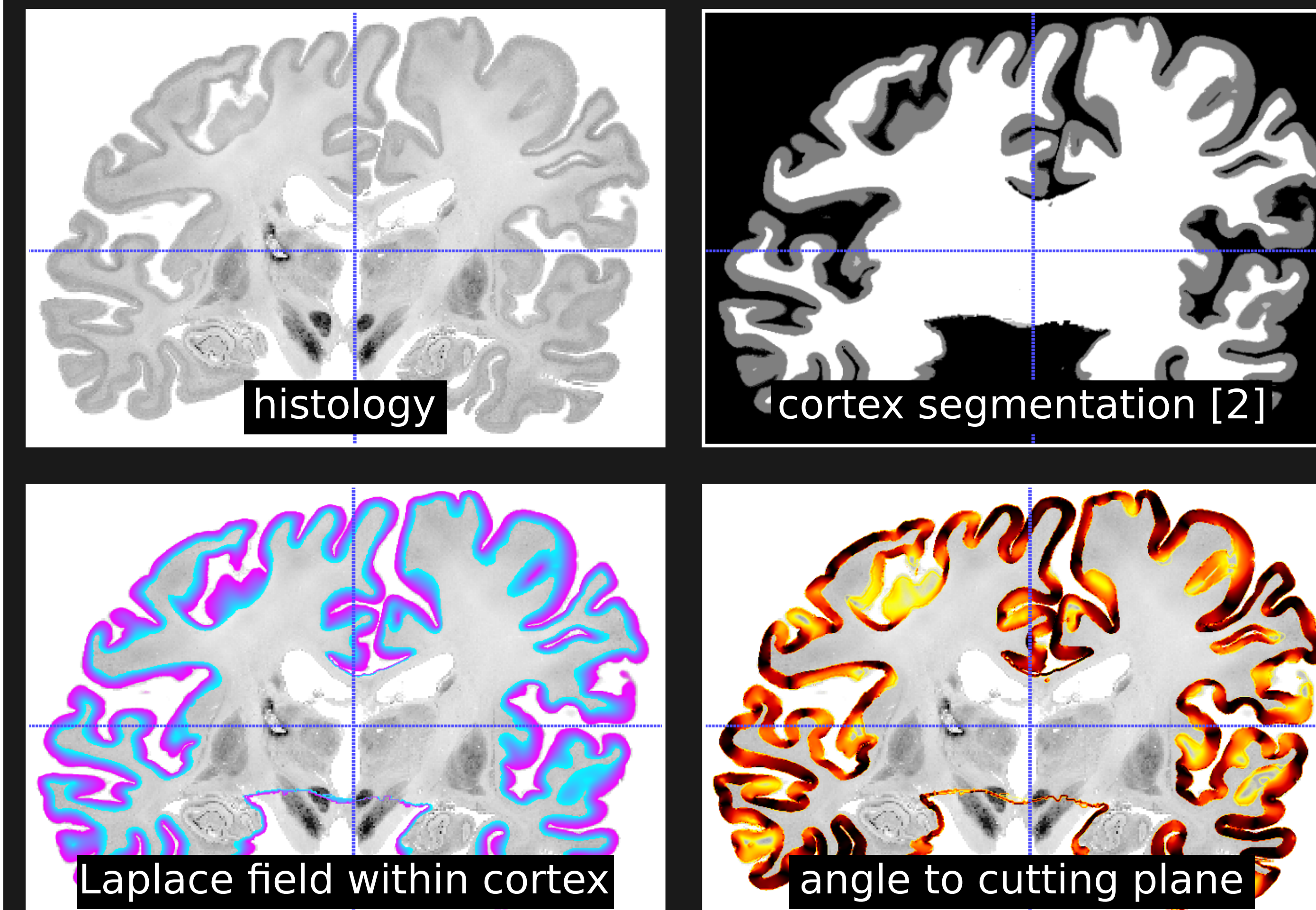


Method

Convolutional Neural Networks for automatic cutting angle prediction

Cutting angle computation in BigBrain [1]

- BigBrain: High-resolution 3D brain model reconstructed from 7404 sections (20µm isotropic resolution)
- Exact cutting angle computation based on 3D context

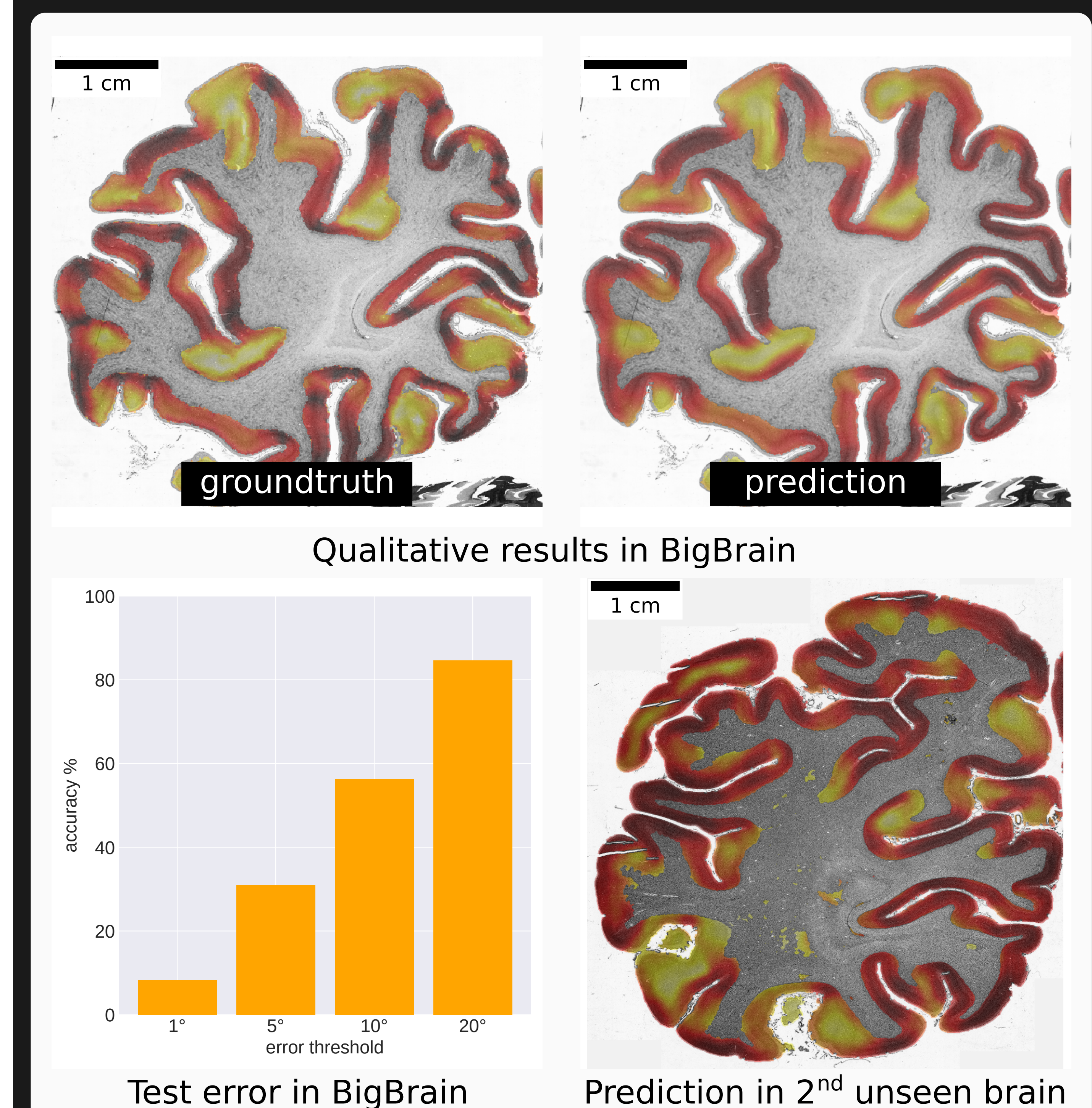


Convolutional Neural Network for angle prediction

- Project cutting angles in BigBrain to unregistered sections
- Train U-Net [3] for regression [4] of angles from histology
- Patchwise processing: 512px at 64µm resolution
- Training: 384'000 patches from 55 sections
- Testing: 15 sections not seen during training
- L2 loss restricted to cortical ribbon
- Intensity+geometric augmentation promotes transferability

Result

Automatic identification of oblique cuts based on cutting angle prediction



Conclusion

- Cutting angle prediction helps to identify oblique cuts
- Predictions on unseen brains demonstrate transferability
- Next: Excluding oblique cuts from downstream analysis tasks

References

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 2. Lewis, L. et al. (2014). "BigBrain: Initial Tissue Classification and Surface Extraction". In: OHBM Annual Meeting.
 3. Ronneberger, O. et al. (2015). "U-Net: Convolutional Networks for Biomedical Image Segmentation". In: MICCAI. Springer, pp. 234–241.
 4. Meyee, I. et al. (2018). "A pixel-wise distance regression approach for joint retinal optical disc and fovea detection". In: MICCAI. Springer, pp. 39–47.
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