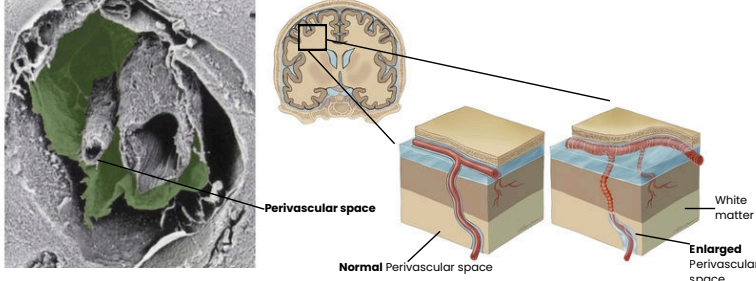


STARDUST - Deep segmentation of brain perivascular spaces in Parkinson's disease: informing decision support from clinical low-field MRI

Letizia Girardi¹, Selene Tomassini¹, Raffaella Di Giacompo², Carlo Cosimo Quattrocchi^{2,3}, Paolo Giorgini¹

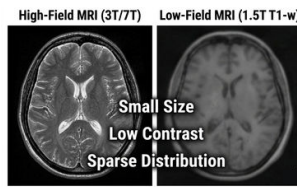
INTRODUCTION



- **Perivascular spaces (PVS)** play a critical role in the brain's glymphatic clearance system
- The **enlargement of PVS** is recognized as a **biomarker** in neurodegenerative diseases such as **Parkinson's disease (PD)**.

CHALLENGES

- **PVS detection** is strongly **limited by MRI spatial resolution**.
- PVS are extremely small, sparsely distributed structures.
- **Low contrast** in 1.5T T1-weighted MRI makes PVS difficult to distinguish from surrounding tissue.



AIM

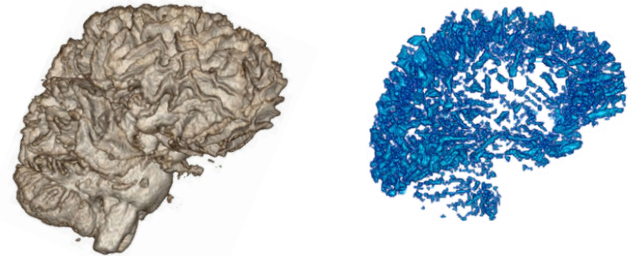
Support **PD biomarker analysis** through **STARDUST**: an automated framework for fine-grained **PVS segmentation** in clinical **1.5T T1-weighted MRI**, enabling an objective proxy for glymphatic function and disease progression.

OBJECTIVES

- Handle anatomical sparsity and label uncertainty.
- Validate reliability in real-world 1.5T protocols, **shifting from subjective visual rating scales to objective, topology-informed quantification**.

RESULTS

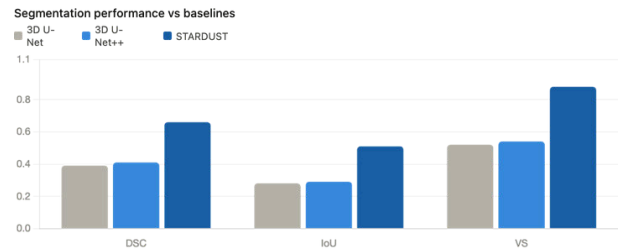
QUALITATIVE RESULTS



WHITE MATTER 3D RECONSTRUCTION

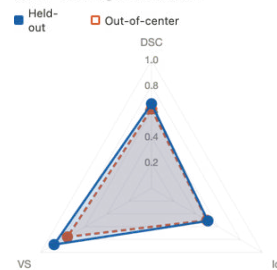
PVS 3D RECONSTRUCTION

QUANTITATIVE RESULTS

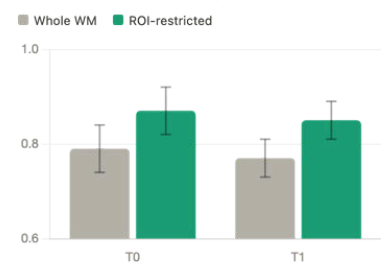


STARDUST outperforms baseline variants.

Cross-center generalization



Temporal stability & ROI analysis

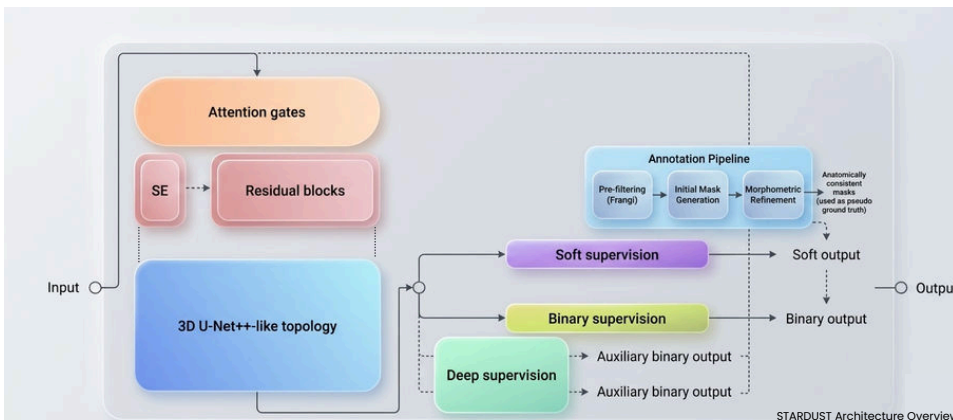


STARDUST demonstrates **robust** performance on the held-out test set, both at global and subject level.

Region-restricted evaluation **improves** anatomical precision and enhances clinical interpretability.

Ablation studies demonstrate that the integration of SE modules, attention gates and dual supervision strategies results in a cumulative **+61% DSC improvement** over the standard 3D U-Net++ backbone.

METHODS



STARDUST Architecture

- **3D U-Net++ backbone** with **SE-residual blocks** and **attention gates**.
- **Multi-output training** leveraging soft supervision and asymmetric losses.
- Specifically **designed for sparse anatomical structures and uncertain labels**.
- **Semi-automatic annotation** via multiscale Frangi filtering and morphometric refinement for high-fidelity pseudo-ground truth.