

¹ Macapype: An open multi-software framework for non-human primate brain anatomical MRI processing

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DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

Software

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Editor: Adam Tyson  

Submitted: 05 September 2025

Published: unpublished

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Non-human primates (NHPs) are increasingly used for neuroimaging studies due to the progress of MR acquisitions and the promises it holds in the field of neuroscience (Milham et al., 2018). Despite the standardization of MRI processing in humans with several well-known software options available, such as AFNI (Cox, 1996), FSL (Smith et al., 2004), SPM12 (Frackowiak et al., 1997), and ANTs (Avants et al., 2011), defining robust processing pipelines for NHP anatomical image segmentation remains difficult.

⁹ Summary

¹⁰ Although brain anatomical Magnetic Resonance Imaging (MRI) processing is largely
¹¹ standardized and functional in humans, it remains a challenge to define robust processing
¹² pipelines for the segmentation of non-human primate (NHP) images. To unify the processing
¹³ of NHP anatomical MRI, we propose Macapype, an open-source framework to create custom
¹⁴ pipelines for data preparation, brain extraction, and brain segmentation.

¹⁵ Statement of Need

¹⁶ Non-human primates (NHPs) are increasingly used for neuroimaging studies due to the progress
¹⁷ of MR acquisitions and the promises it holds in the field of neuroscience (Milham et al., 2018).
¹⁸ Despite the standardization of MRI processing in humans with several well-known software
¹⁹ options available, such as AFNI (Cox, 1996), FSL (Smith et al., 2004), SPM12 (Frackowiak
²⁰ et al., 1997), and ANTs (Avants et al., 2011), defining robust processing pipelines for NHP
²¹ anatomical image segmentation remains difficult.

²² State of the field

²³ Two categories of methods have been proposed to address the issue of NHP anatomical
²⁴ MR image segmentation. The first category corresponds to particular implementations for
²⁵ NHP images of existing human-MRI softwares such as **NHP-Freesurfer** and **CIVET-Macaque**,
²⁶ respectively relying on human-MRI softwares FreeSurfer (Fischl, 2012) and CIVET (Lepage
²⁷ et al., 2021). The second category relies on the use of deep-learning and machine learning
²⁸ techniques, such as **U-Nets**, for example **nBEST** to provide brain mask, segmentation of GM,
²⁹ WM and subcortical nuclei (Zhong et al., 2024). The latter requires the use of GPUs, most
³⁰ existing softwares performs relatively badly on small NHP species such as marmoset due to
³¹ the lack of flexibility in the processing steps and the variability of brain peculiarities among
³² NHP species.

³³ Software design

³⁴ In this context, we propose a general framework for the tissue segmentation of non-human
³⁵ primate brain MR images that can provide multiple pipelines to adapt to a variety of image
³⁶ qualities and species. This open-source framework, named Macapype, is built upon Nipype, a
³⁷ widely used Python framework for human MRI analysis.

³⁸ The Macapype package was specifically designed to provide wrappers of custom tools specific
³⁹ to NHP anatomical MRI preprocessing, as well as pipelines and workflows to achieve high-
⁴⁰ quality automated tissue segmentation of NHP anatomical images. In particular, the tuning of
⁴¹ parameters for different species, should be possible if needed via the use of parameter files.

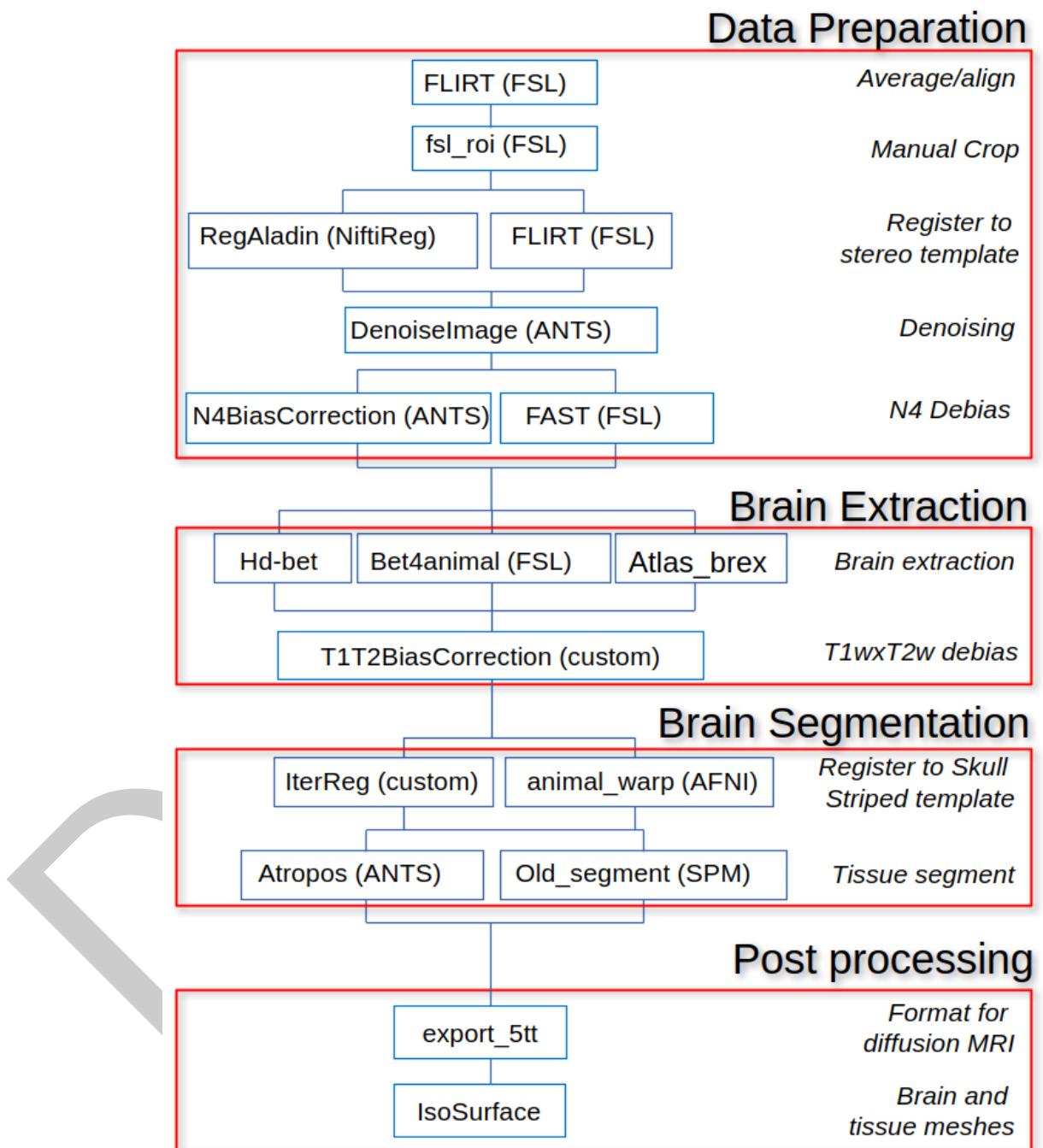


Figure 1: Different pipelines are chained

⁴² Pipelines

⁴³ Macapype provides configurable pipelines organized in three steps: data preparation, brain
⁴⁴ extraction, and brain segmentation. Post-processing allows for conversion to format for further

45 processing outside Macapype.

46 ▪ Data Preparation Pipeline

47 The data preparation pipeline is specified in a JSON parameters file and depends on individual
48 parameters. If cropping parameters are absent, Macapype performs an automated but low-
49 precision crop. The input volume is reoriented in a standard space, and denoising and debiasing
50 steps are performed.

51 ▪ Brain Extraction Pipeline

52 For the skull-stripping step, Macapype offers a choice between AtlasBREx (Lohmeier et al.,
53 2019) and bet4animal, an optimized version of the brain extraction tool (BET in FSL) for
54 NHP. HD-BET (Isensee et al., 2019) is also available for deep-learning-based brain extraction.

55 ▪ Tissue Segmentation Pipeline

56 Tissue segmentation is template-based and can be done in template or native space. Macapype
57 provides templates for macaque, marmoset, baboon, and chimpanzee. T1xT2 debias is applied,
58 followed by normalization and segmentation using ANTs-based Atropos or SPM12-based old
59 segment.

60 ▪ Post-Processing Pipeline

61 For compatibility with further processing, Macapype provides formatting options, such as the
62 5tt file from MRtrix (Tournier et al., 2019) for further processing of diffusion MRI and meshes
63 in STL format for 3D printing.

64 Research impact statement

65 Macapype is compatible with FAIR principles, storing all processing steps and parameters in
66 a JSON file. It allows evaluation of results at different preprocessing steps and is tested on
67 images from the PRIME-DE database (Milham et al., 2018) and is listed as a software solution
68 on PRIME-RE (Messinger et al., 2021).

69 AI usage disclosure

70 No generative AI tools were used in the development of this software, or the preparation
71 of supporting materials. This manuscript has been written with the help of Mistral AI for
72 formatting, as well as syntax and language checking.

73 Acknowledgements

74 We are grateful to Adrien Meguerditchian, Paul Apicella, and Guilhem Ibos for providing MRI
75 datasets for testing.

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