### Brain mapping tools for neuroscience research

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# The goal of brain mapping



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# 1. Registration

### Align images into a standard coordinate system



- Enrich information by fusing modalities
- Analyze different specimens statistically
- Build databases of information indexed to spatial coordinates

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### 2. Interpretation

#### Leverage information stored in atlas coordinates.<sup>1</sup> MBA ARA



- Label images with standard ontologies
- Index to gene expression, cell types, tractography, etc.

 $<sup>^1 \</sup>rm MBA:$  Mouse brain architecture brainarchitecture.org, ARA: Allen reference atlas connectivity.brain-map.org/

# 3. Transformation

Studying transformations quantifies growth or atrophy



- Here thickness change in transentorhinal region measured from longitudinal MRI<sup>2</sup>
- Previously only observed at autopsy

<sup>&</sup>lt;sup>2</sup>Tward, Daniel J., et al. "Entorhinal and transentorhinal atrophy in mild cognitive impairment using longitudinal diffeomorphometry." Alzheimer's & Dementia: Diagnosis, Assessment & Disease Monitoring 9 (2017): 41-50. Daniel Tward (dtward@cis.jhu.edu) Johns Hopkins University Brain mapping tools for neuroscience research

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#### Significant thickness atrophy (%/yr)

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### The ingredients of a brain mapping tool

Transformation model: What types of mappings do we consider?



### Similarity: How good is an alignment?



### Regularization: How likely is a given transformation?



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### Challenges and solutions

Most brain mapping techniques were developed for medical imaging, but neuroscience data faces unique challenges:

- Incomplete or sliced data
- Artifacts or damaged tissue
- Multiple different modalities or appearance



We use machine learning techniques to predict one image from another, while jointly performing registration<sup>3</sup>



<sup>&</sup>lt;sup>3</sup>Tward, Daniel Jacob, et al. "Diffeomorphic registration with intensity transformation and missing data: Application to 3D digital pathology of Alzheimer's disease." BioRxiv (2019): 494005.

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# ARDENT<sup>4</sup>: NeuroData's open source brain mapping tool

Publications and code available online from neurodata.io/reg

Ingredient	Choice	Benefit
Transform	Diffeomorphism	Smooth invertible fluid transform
Similarity	Log likelihood	Enables statistical approaches to
		artifacts and multi-modality
Regularization	Kinetic energy	Enables sparse representations ef-
		fective in high dimensional bias
		variance tradeoff <sup>5,6</sup>



<sup>4</sup>Affine and Regularized Diffeomorphic Numeric Transform. <sup>5</sup>Tward, Daniel, et al. "Parametric surface diffeomorphometry for low dimensional embeddings of dense segmentations and imagery. IEEE transactions on pattern analysis and machine intelligence (2016) <sup>6</sup>Tward, Daniel, et al. "Estimating diffeomorphic mappings between templates and noisy data: Variance bounds on the estimated canonical volume form. Quarterly of Applied Mathematics (2019).

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