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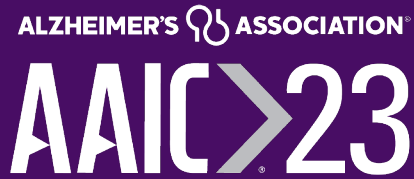
JULY 16-20 > AMSTERDAM, NETHERLANDS, AND ONLINE

ISTAART Neuroimaging PIA THE BASICS OF NEUROIMAGING SEMINAR SERIES



ISTAART Neuroimaging PIA

The Basics of Neuroimaging Series



BASICS OF NEUROIMAGING

STRUCTURAL MRI

DR DAVID CASH

UCL QUEEN SQUARE INSTITUTE OF NEUROLOGY, LONDON, UK



@davecash75

By the end of this session, you should be able to:

- Identify the common structural MRI sequences and how they are used in dementia research
- Outline the basic preprocessing steps needed for structural MRI data and how to look for issues in the data and processing
- Differentiate the numerous applications of registration using structural MRI

NEUROIMAGING DATA ANALYSIS: A BLUEPRINT FOR STRUCTURAL MRI

1. Data acquisition

Aim: obtain good quality and consistent data

Different sequences (T1w, T2w, FLAIR) provide complimentary information
Trade-offs often necessary (e.g. time vs resolution).
Optimize protocol for research aim.

2. Data preprocessing

Aim: Reduce noise and prepare data for further analyses

Bias correction: remove intensity inhomogeneity
Brain extraction: remove non-brain tissue
Registration: within-subject, align to template
Requires careful checking.

3. Single-subject analysis

Aim: Obtain measure of interest for each subject (often an image)

Brain/regional volumes, GM/WM concentration, cortical thickness, White Matter Hyperintensities, longitudinal change between scans

4. Group-level analysis

Aim: Compare single-subject results across groups

Comparisons between groups
Associations of volumes with other clinical data
Voxel/Vertex-wise analysis
Consider data harmonization if multiple cohorts

5. Statistical inference

Aim: test reliability of results and generalizability to the general population

Common step across modalities

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What can we measure with structural MRI?

1. Data acquisition

2. Data preprocessing

3. Single-subject analysis

4. Group-level analysis

5. Statistical inference

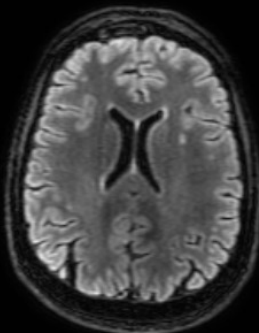
T1-Weighted



T2-Weighted



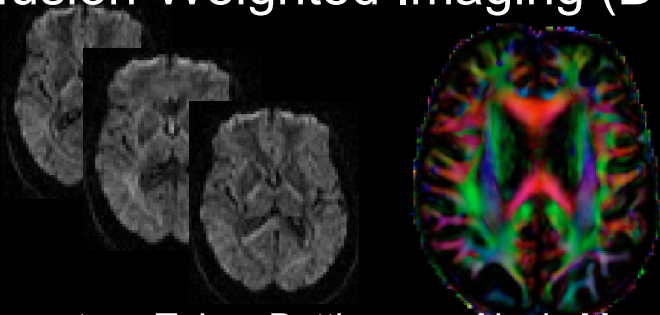
T2-FLAIR



Arterial Spin Labeling (ASL)



Diffusion-Weighted Imaging (DWI)



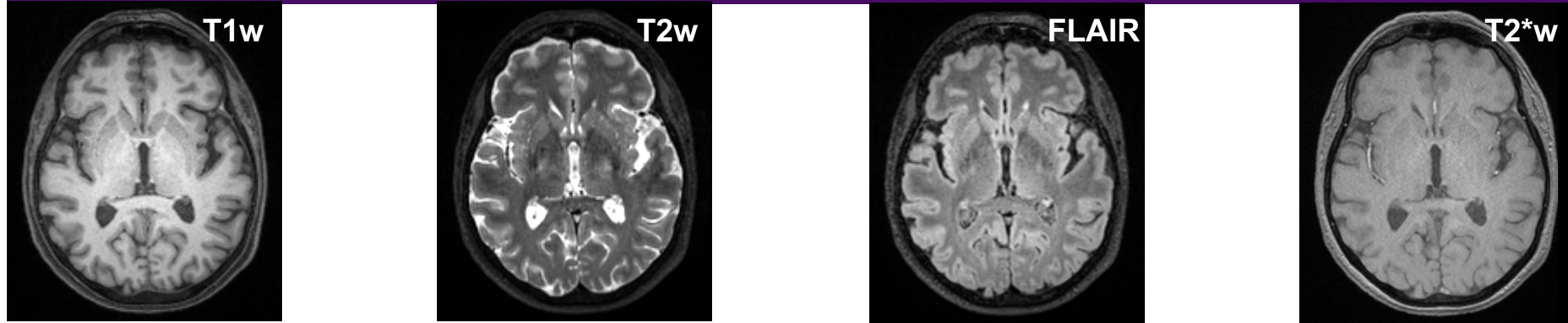
April 21
webinar

Functional-MRI (fMRI)



April 26
webinar

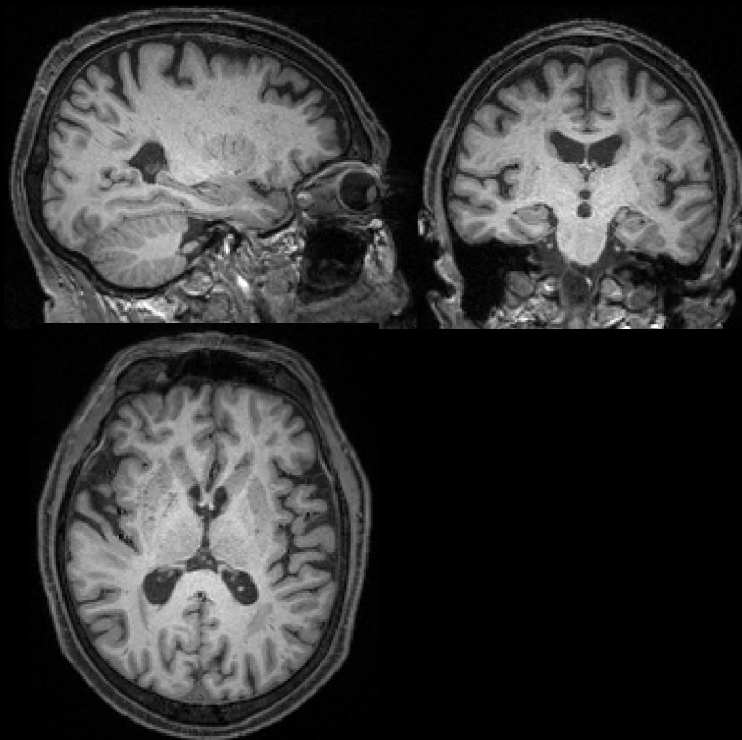
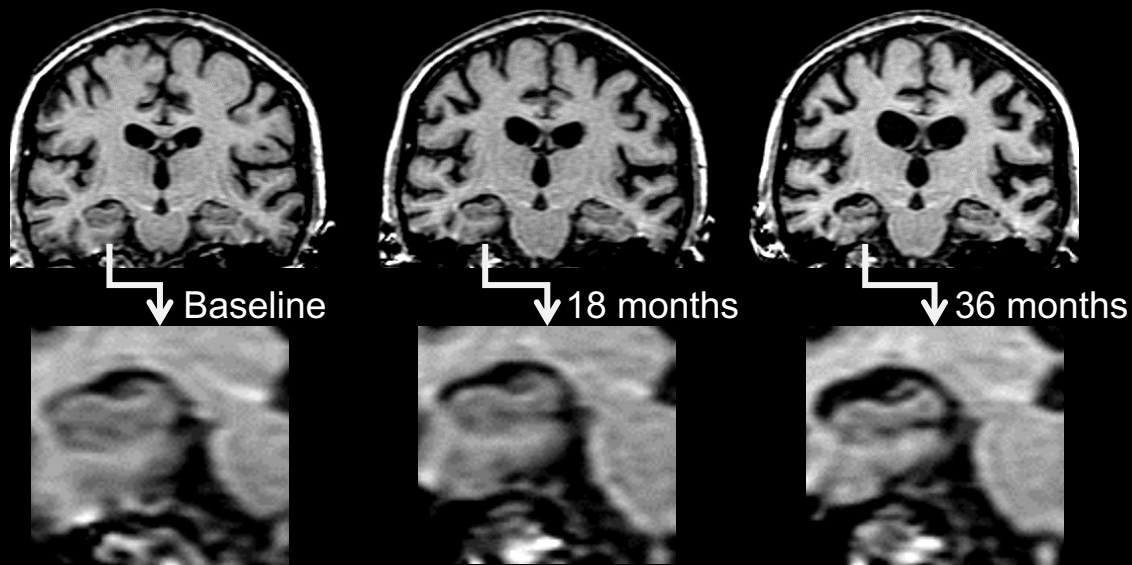
Structural MRI scans



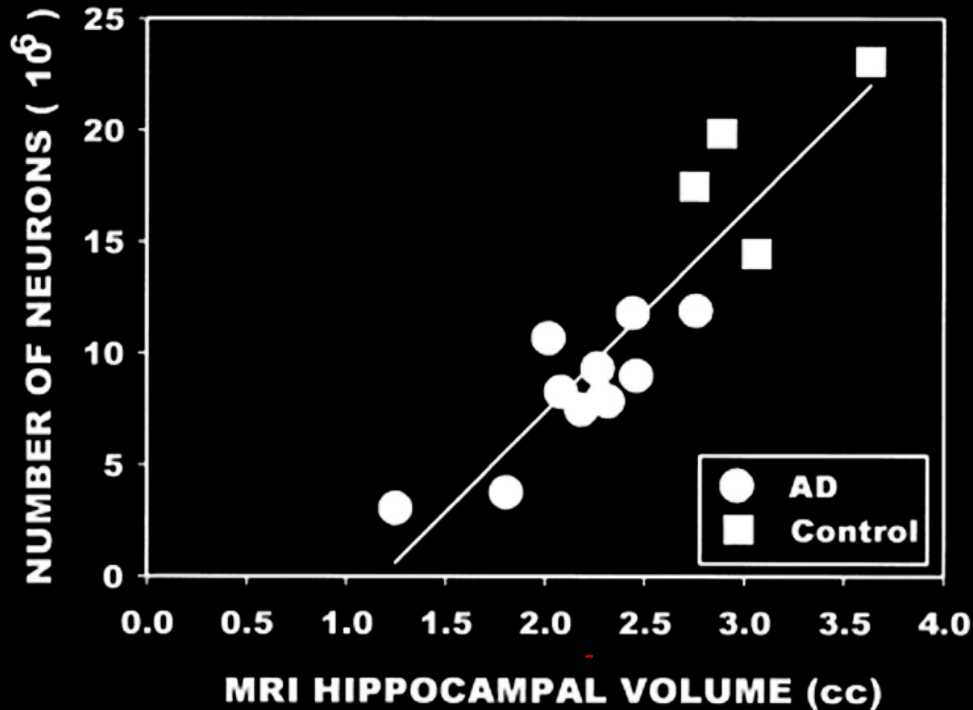
- The MRI scanner can be tuned in different ways (**pulse sequences**), resulting in different contrasts
- Structural MRI sequences are sensitive to (or **weighted** by) one of three fundamental properties: **T1, T2, and T2* relaxation times**.
 - *Intensities are NOT quantitative measures of T1, T2, T2*.*
- **FLAIR** (FLuid Attenuated Inversion Recovery) is a T2 weighted sequence where we suppress CSF signal, so bright CSF voxels become dark

T1-weighted MRI

- High-resolution (~1 mm) information about **neuroanatomy** and **neurodegeneration**
- Can be acquired in any orientation in ~4-6 minutes
- Good contrast between different tissues (GM, WM, CSF)

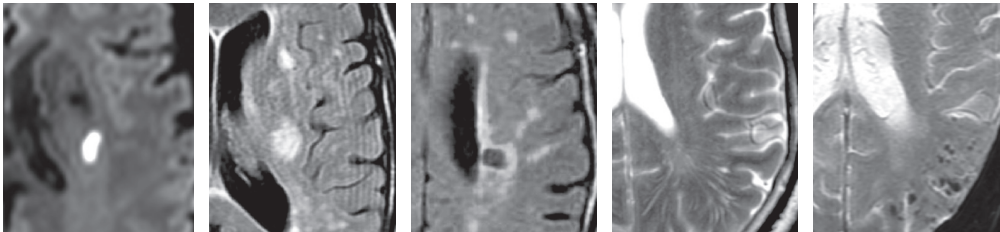


Atrophy on MRI relates to loss of neurons



Bobinski M et al. The histological validation of post mortem magnetic resonance imaging-determined hippocampal volume in Alzheimer's disease. *Neurosci* 95(3): 721-725. (1999)

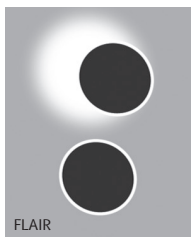
- Best sequence for detecting white matter hyperintensities (WMH) and other vascular-related damage
- Provide complimentary information to T1
- Clinical versions of these scans tend to have good in-plane resolution (≤ 1 mm) but thick slices (thick slabs of 3-5 mm)
- Newer versions of these scans have similar resolution to T1 scans



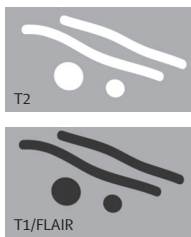
Recent small subcortical infarct



Subcortical hyperintensity



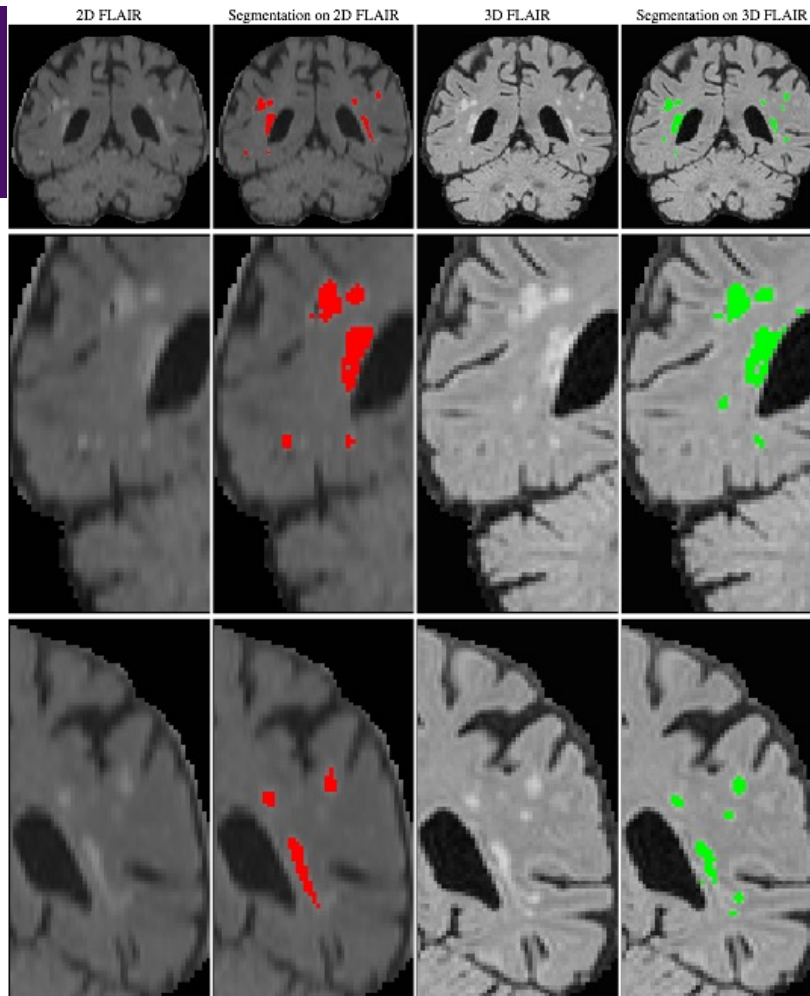
Lacune



Perivascular spaces



Cerebral microbleeds



Images courtesy Carole Sudre, UCL

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What are the key processing steps?

1. Data acquisition

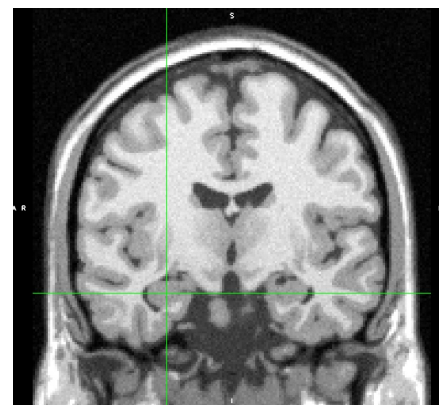
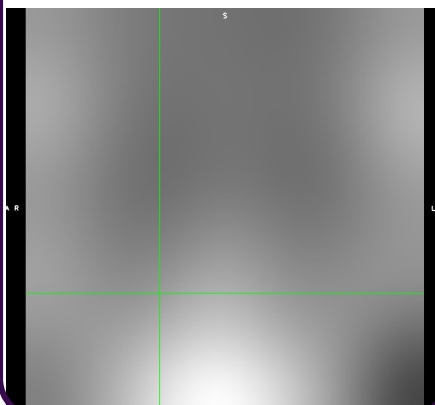
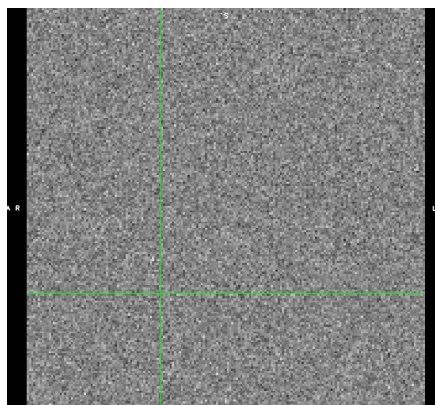
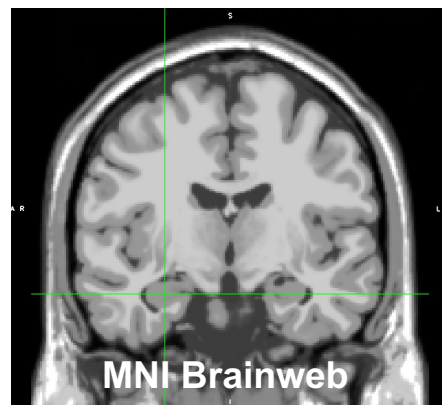
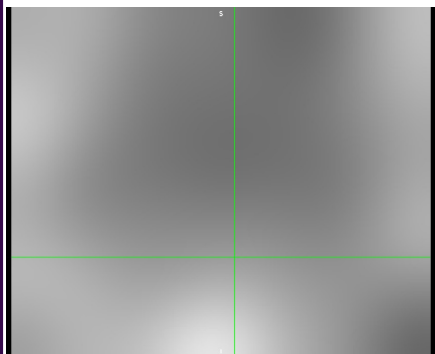
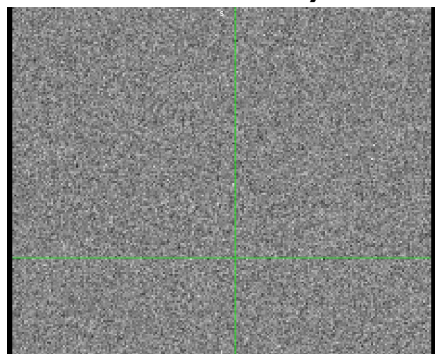
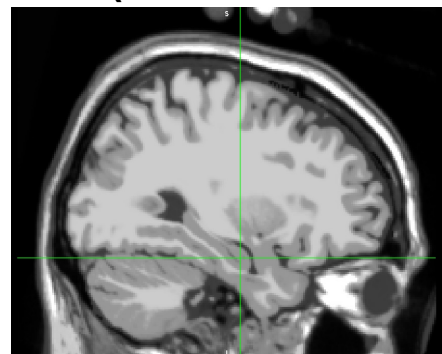
2. Data preprocessing

3. Single-subject analysis

4. Group-level analysis

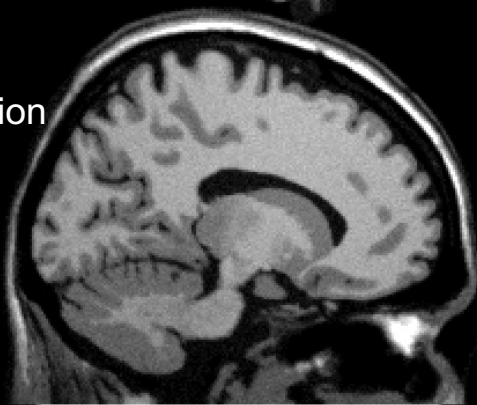
5. Statistical inference

(True data + Noise) x Bias = Observed data

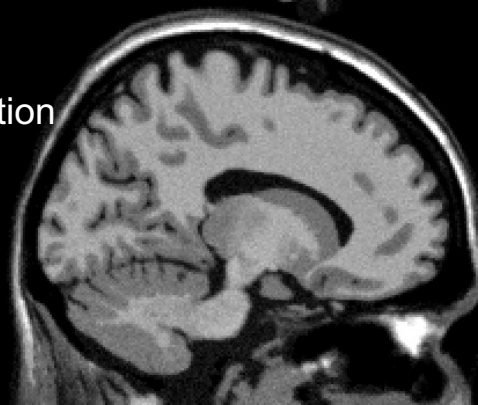


Bias correction: Has it worked?

Before
Bias
Correction



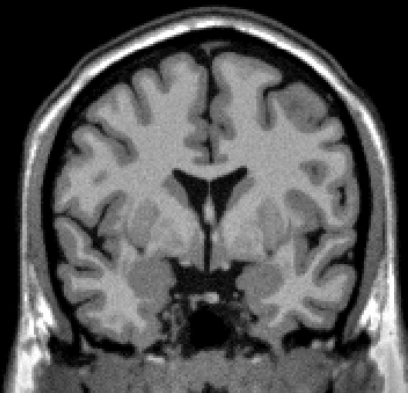
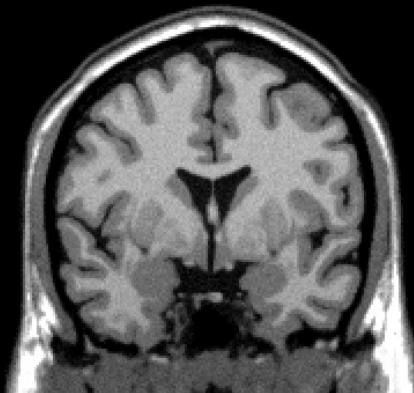
After
Bias
Correction



3E+03

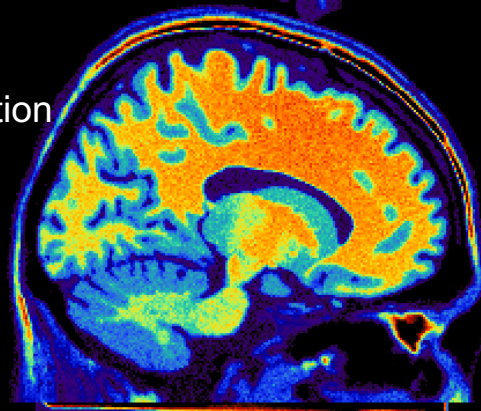
mbrainweb_observed

500

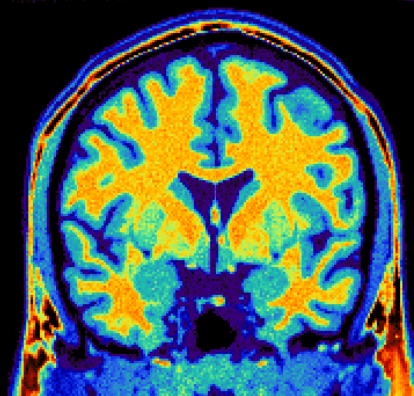
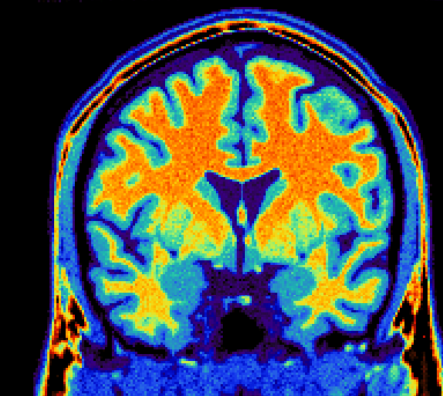
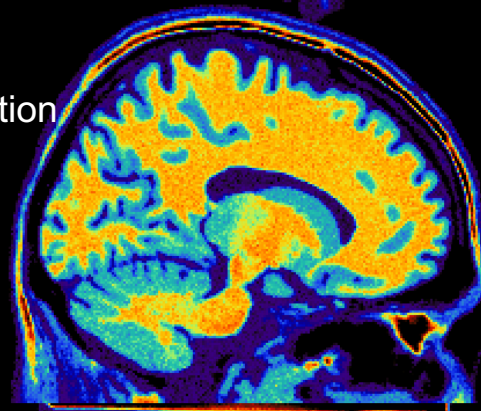


Bias correction: Did it work?

Before
Bias
Correction



After
Bias
Correction

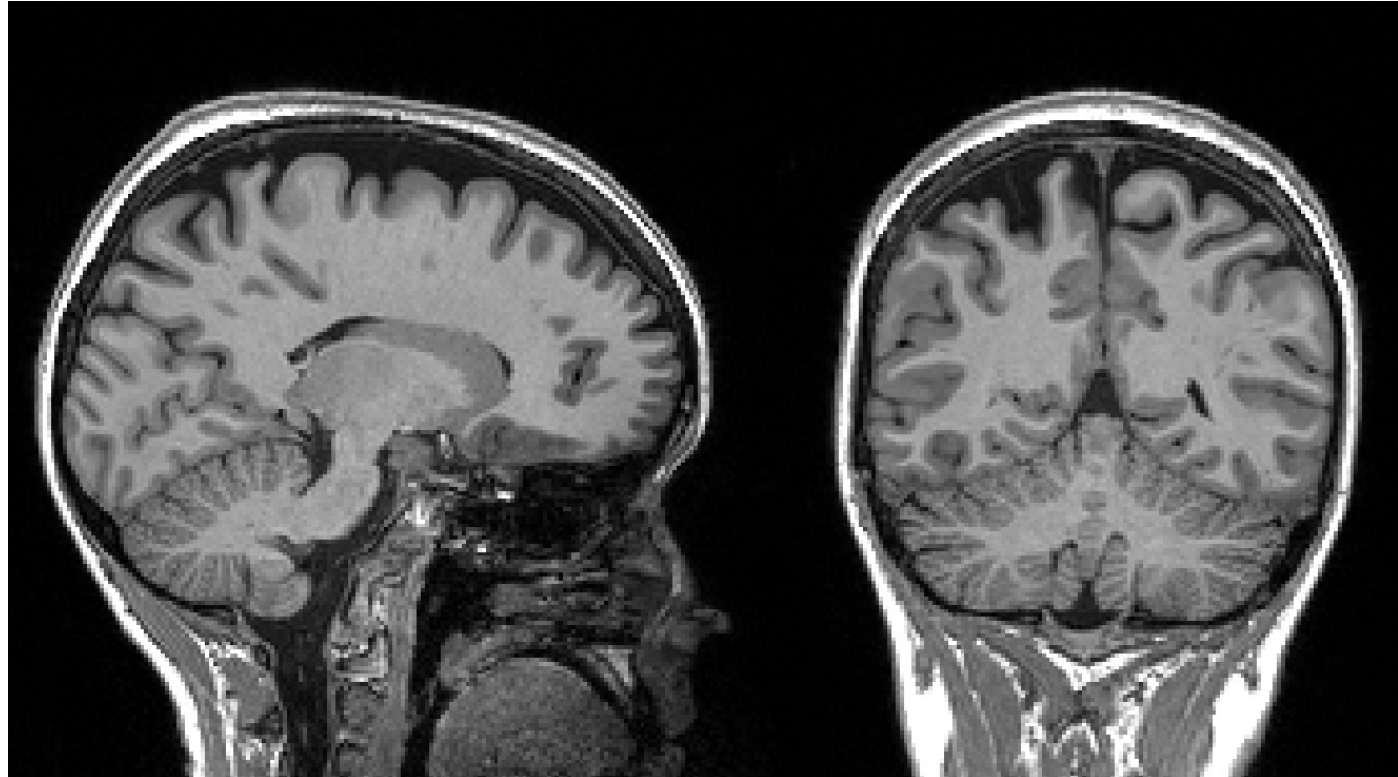


3E+03

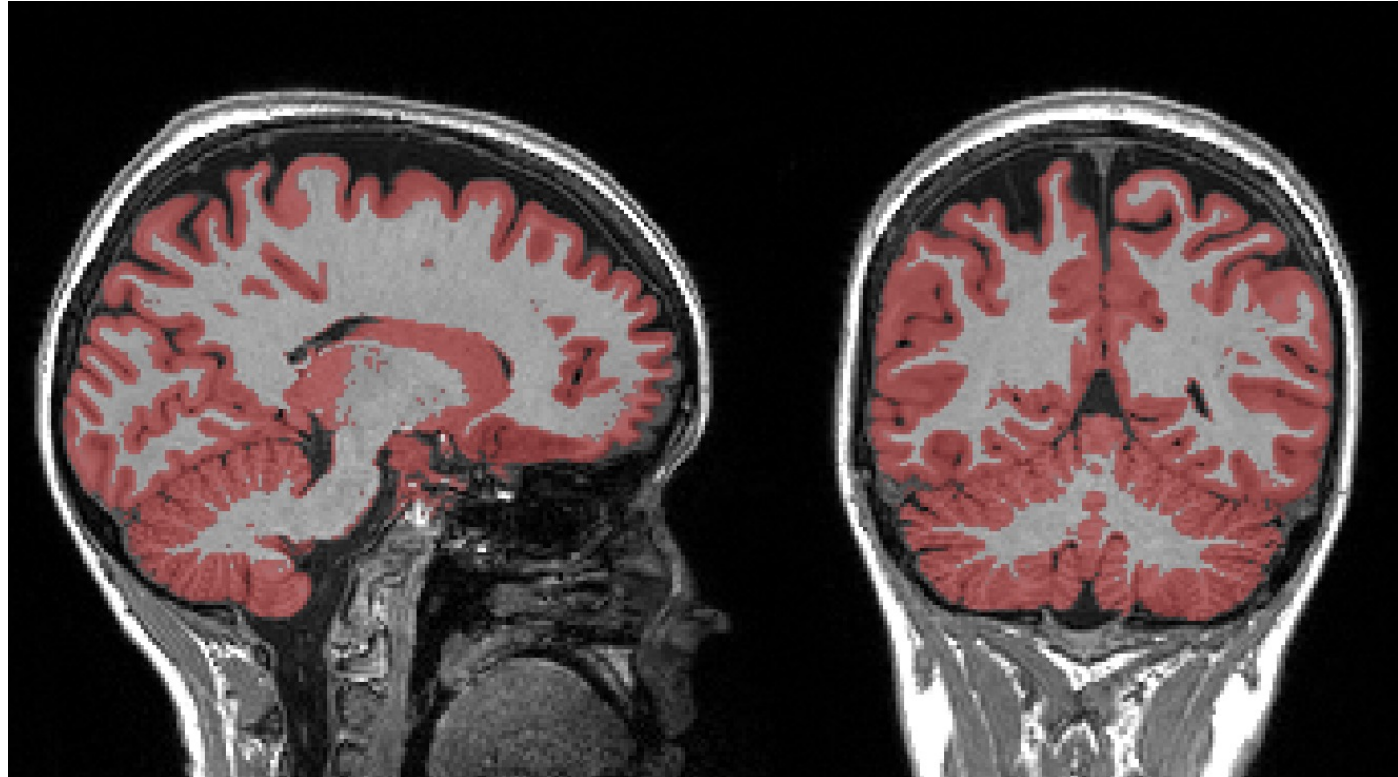
mbrainweb_observed

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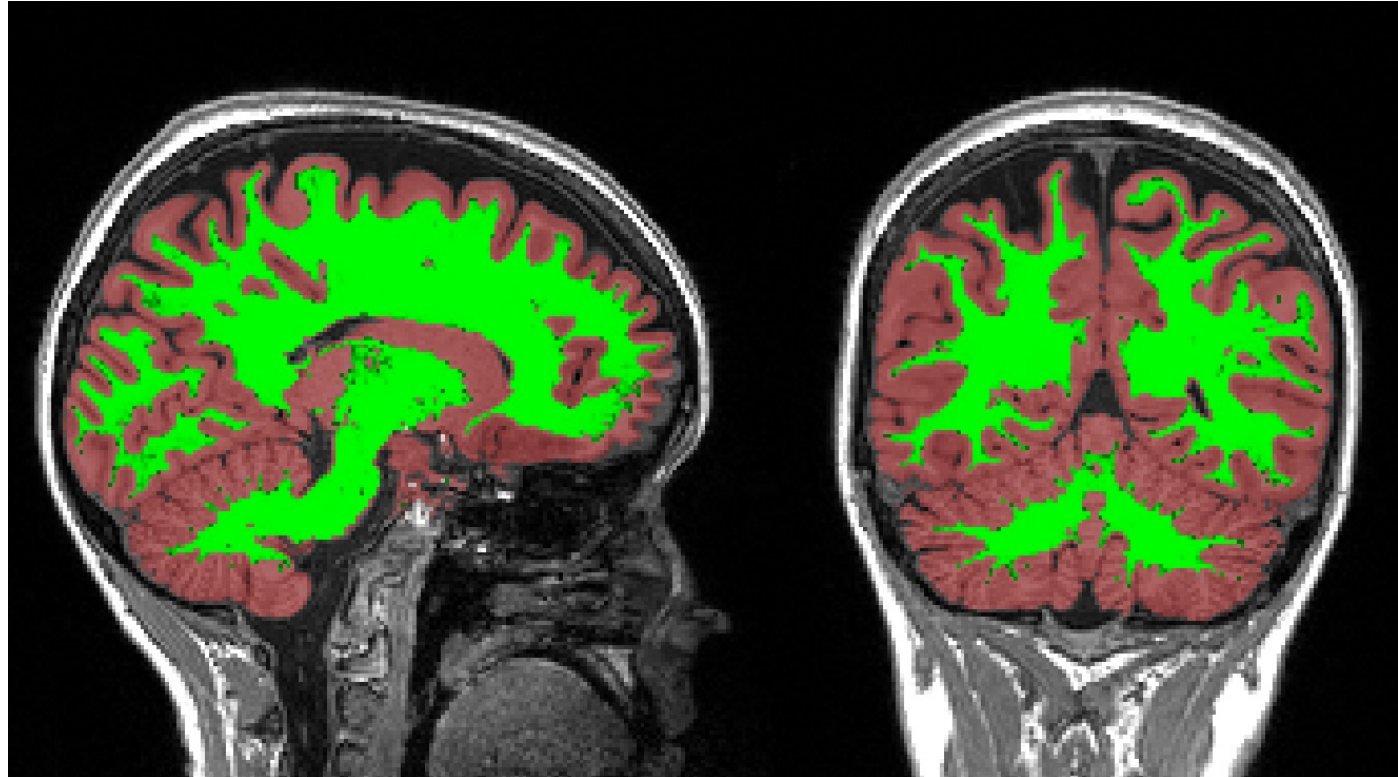
- Assigns voxels to one of three main tissue classes
 - Grey matter (GM)
 - White matter (WM)
 - Cerebrospinal fluid (CSF)
- “Soft” segmentation - each voxel contains the probability that it belongs to a class



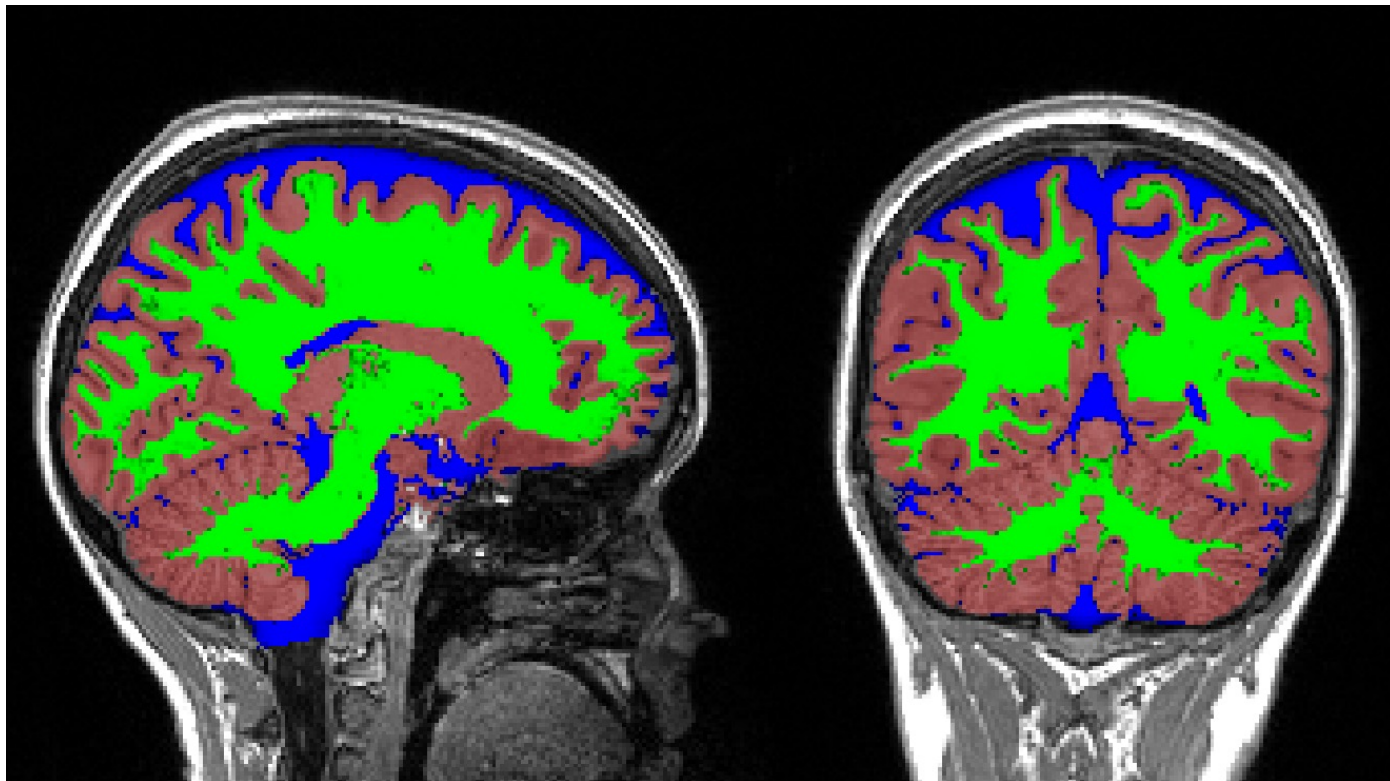
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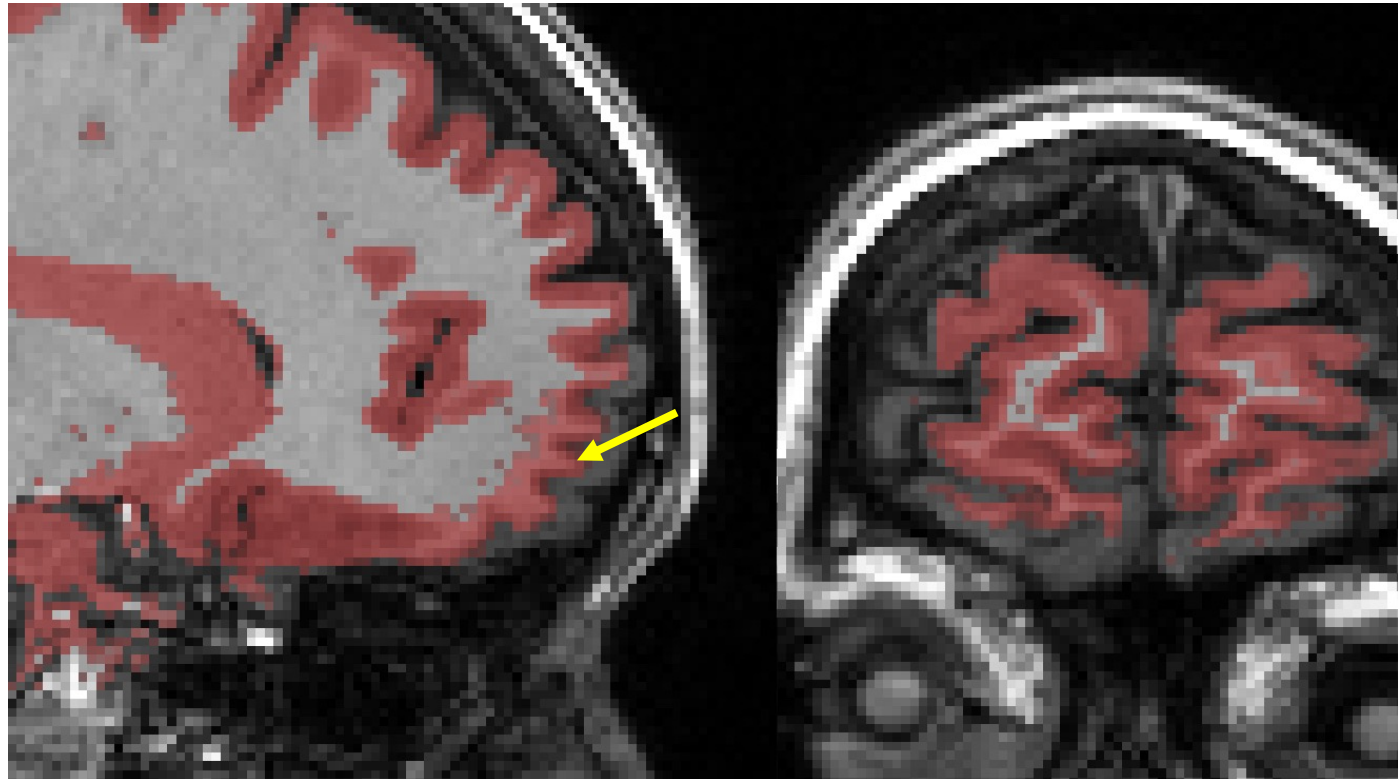


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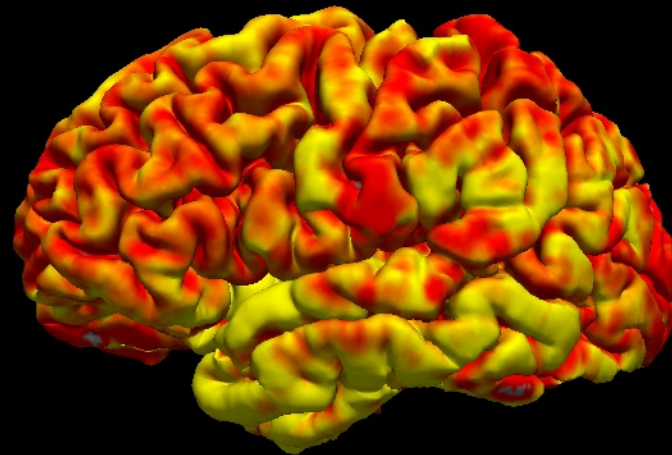
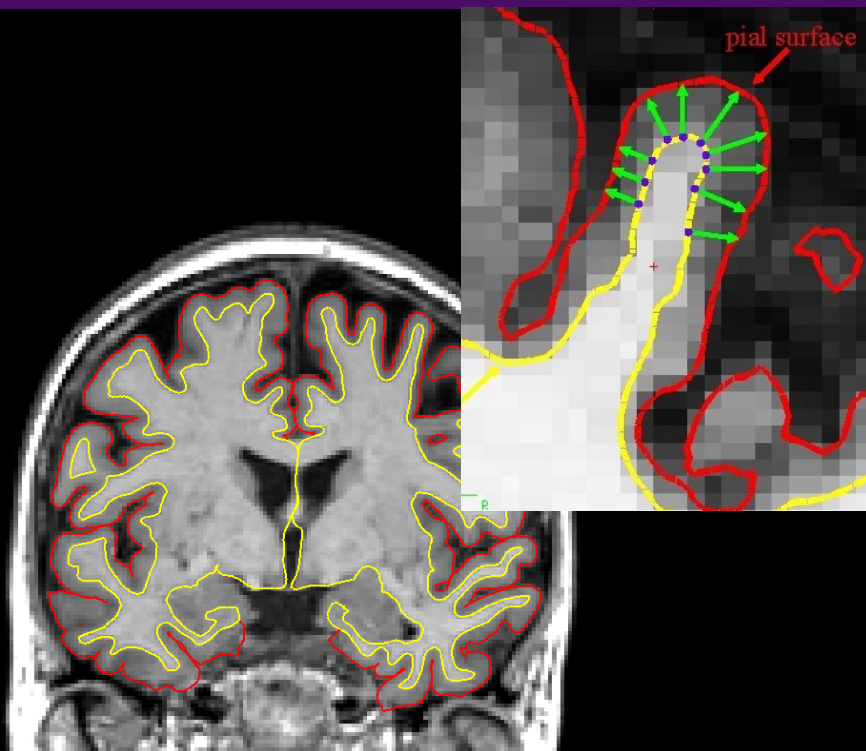


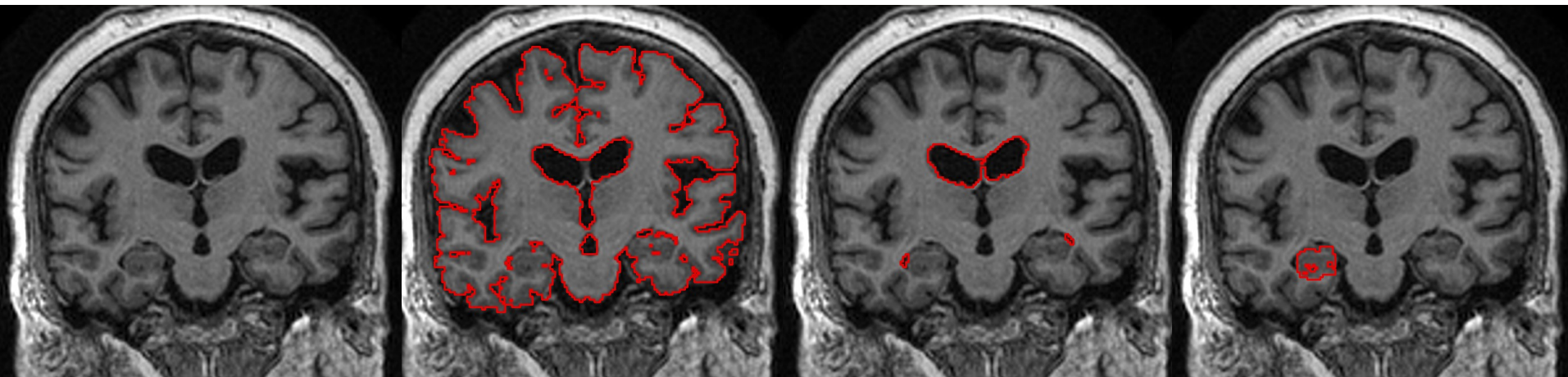
Tissue segmentation: Did it work?

- Assigns voxels to one of three main tissue classes
 - Grey matter (GM)
 - White matter (WM)
 - Cerebrospinal fluid (CSF)
- “Soft” segmentation - each voxel contains the probability that it belongs to a class

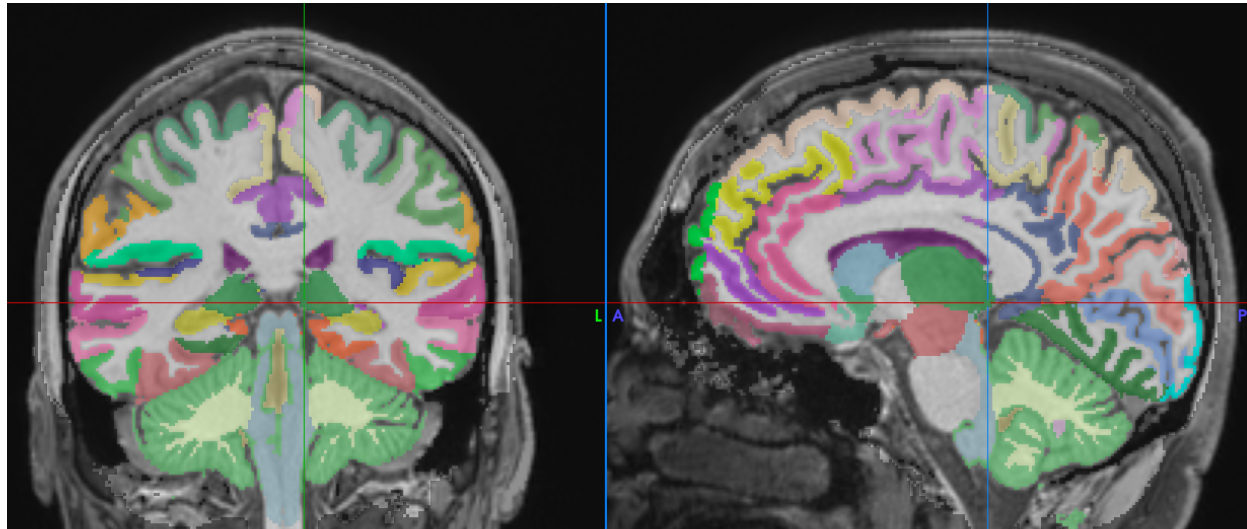


Cortical Thickness





- Rather than tissue types, assign voxels to anatomically defined structures in the brain
- Based on a single atlas or multiple subjects from a template database
- Helpful for regional statistics, both in T1 and multimodal studies
- Difficult to check each region in each image, look for obvious failures, volume outliers



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IMAGE REGISTRATION

1. Data
acquisition

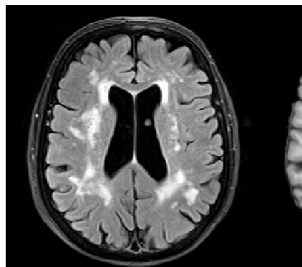
2. Data
preprocessing

3. Single-subject
analysis

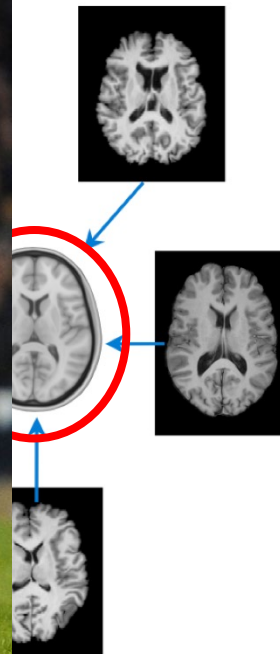
4. Group-level
analysis

5. Statistical
inference

Within-subject &

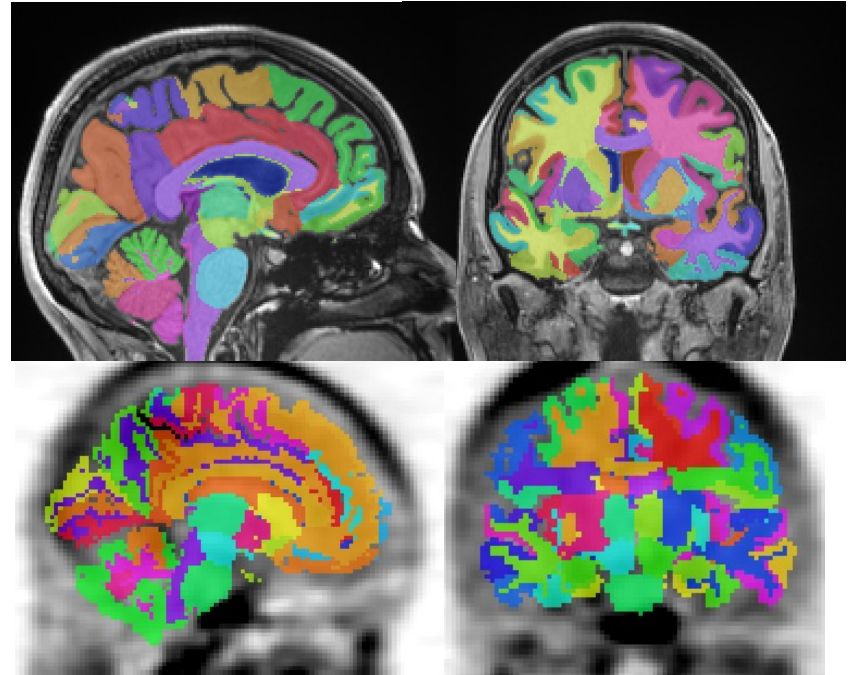


Inter-subjects



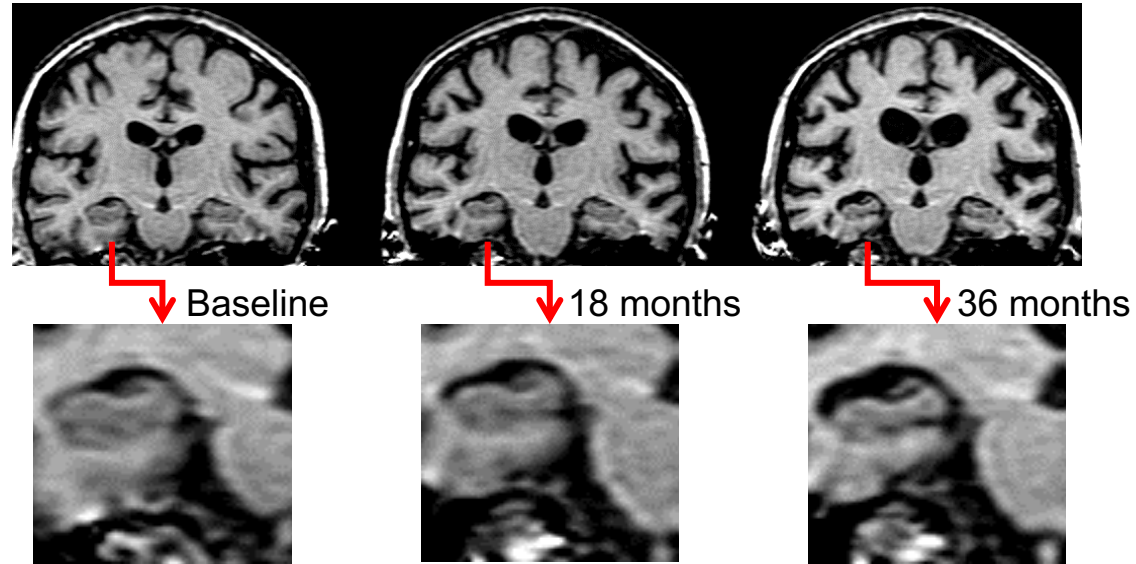
Standard space =
"average brain" used as reference

- Structural T1 provides high-resolution anatomical context for other lower resolution modalities (fMRI, DWI, PET)
- Regions of interest (ROIs) defined on the structural T1 scan can be transferred to co-registered images
- Tissue properties from segmentation can also provide some information on partial volume effect (mixture of different tissues)



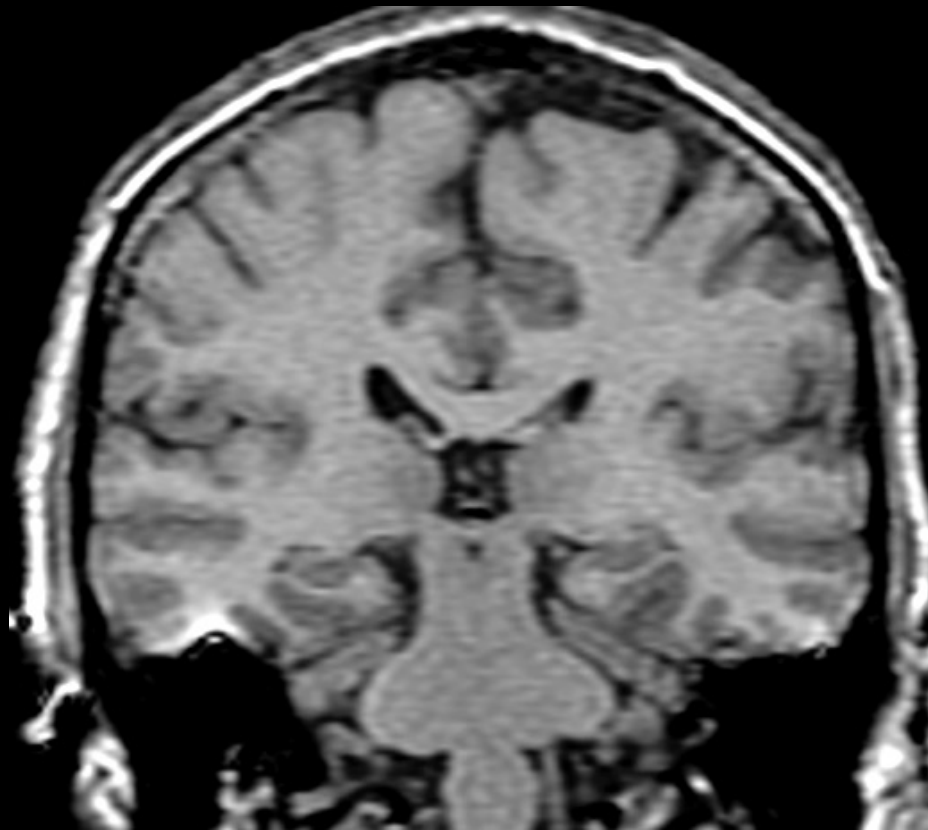
Longitudinal registration: within subject, between sessions

- Register baseline T1 with follow-up scans and measure differences
- More sensitive to disease-related atrophy than cross-sectional measures
- Abnormal rates of atrophy can be detected with scan intervals as small as six months apart (though longer intervals tend to be more reliable)
- Be careful! Check that changes in the images result from disease-related effects, not changes in acquisition (different parameters, movement)



MCI

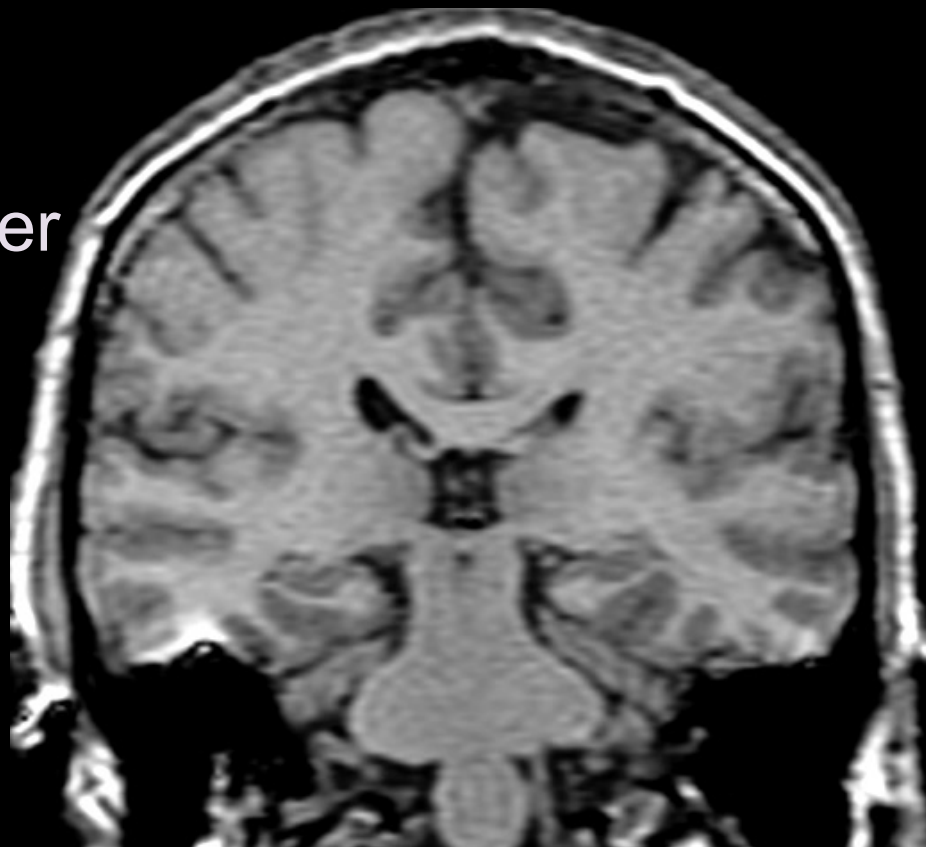
Scan 1



MCI

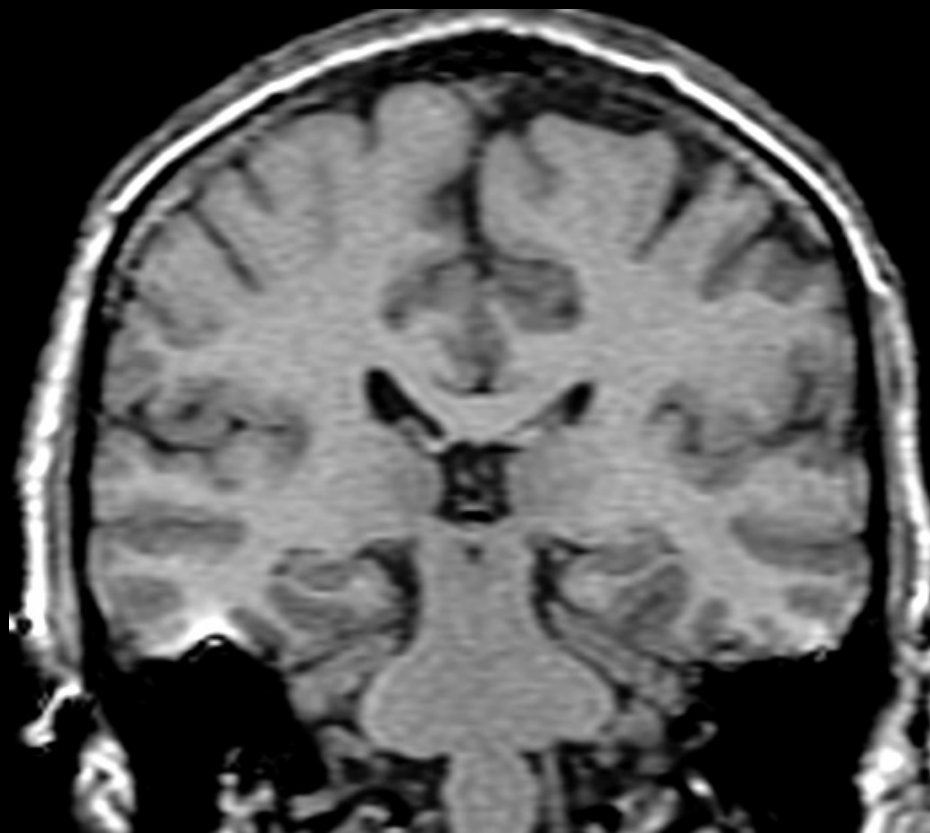
Scan 2

1 year later



MCI

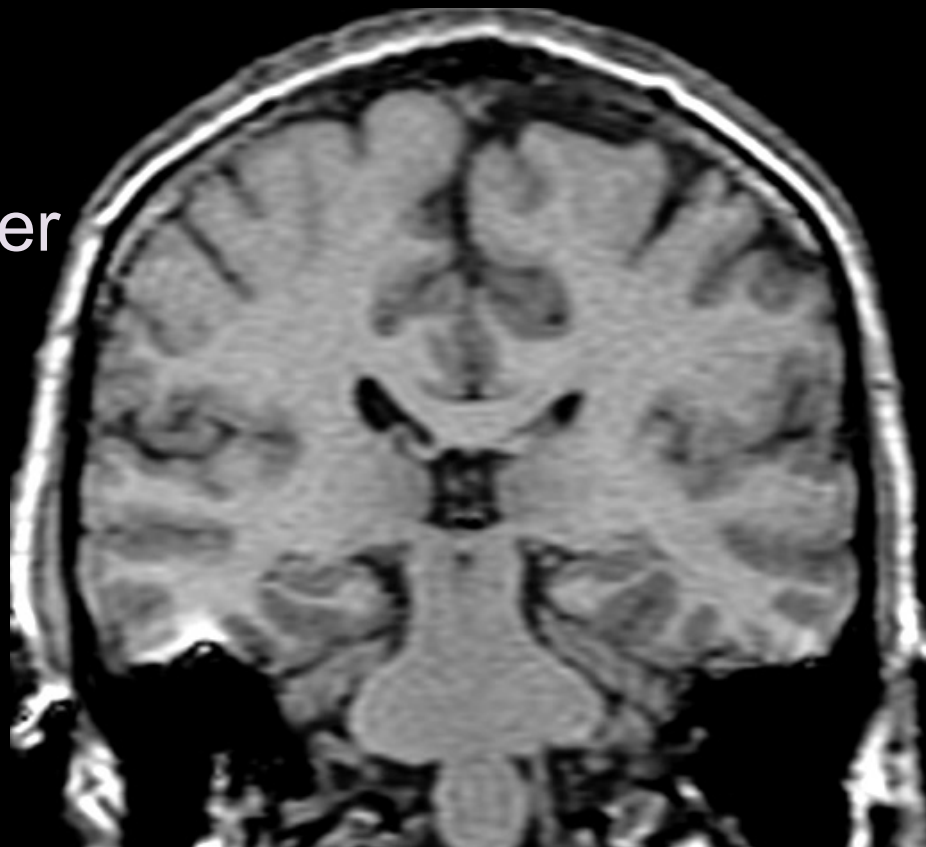
Scan 1



MCI

Scan 2

1 year later



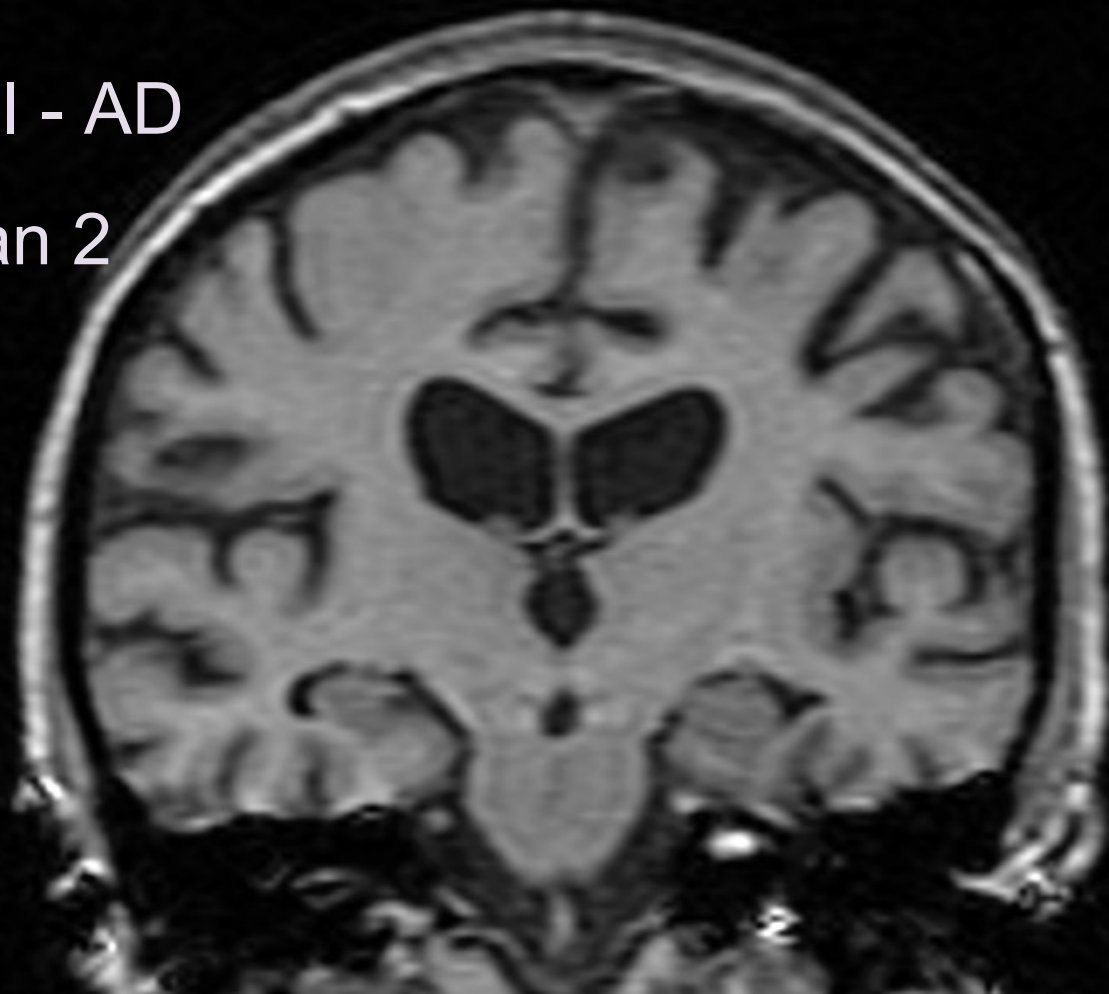
MCI

Scan 1



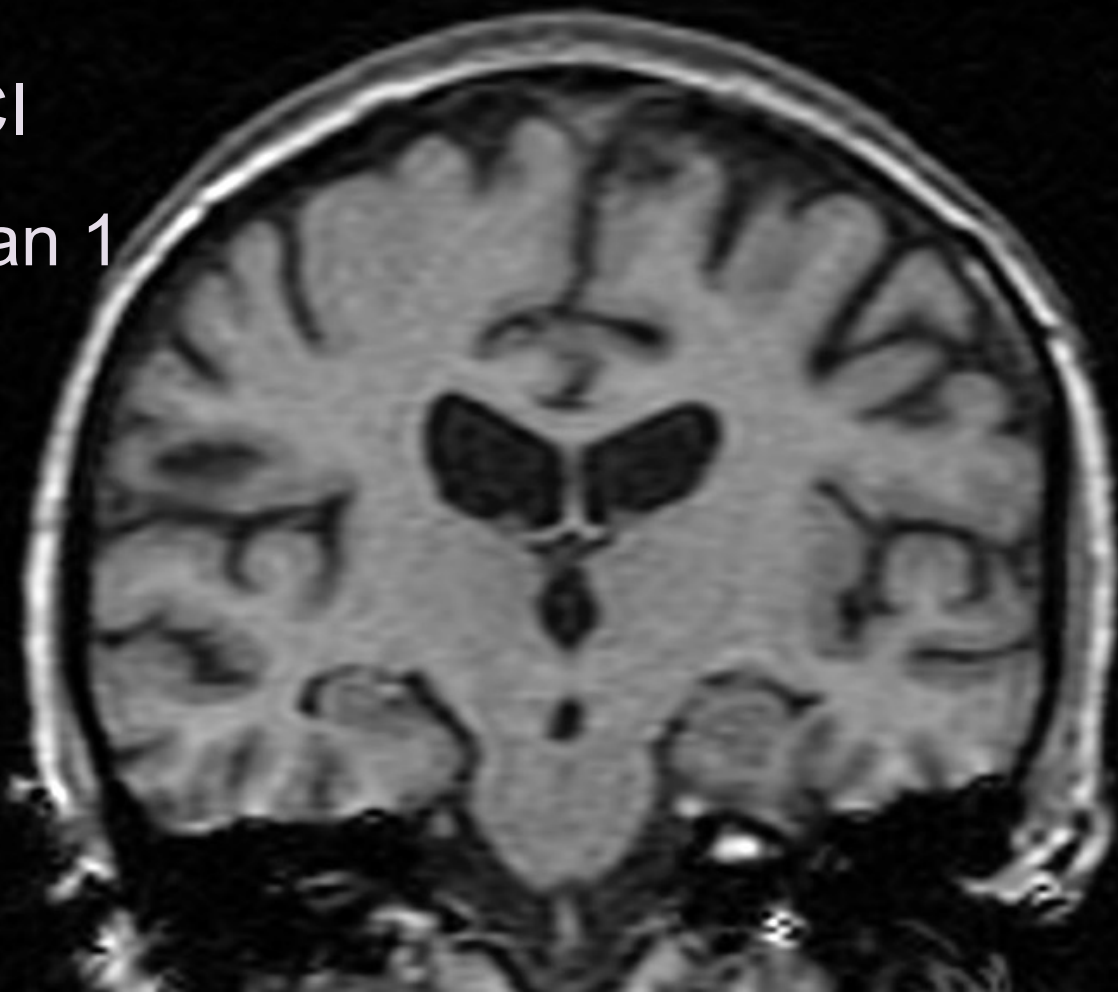
MCI - AD

Scan 2



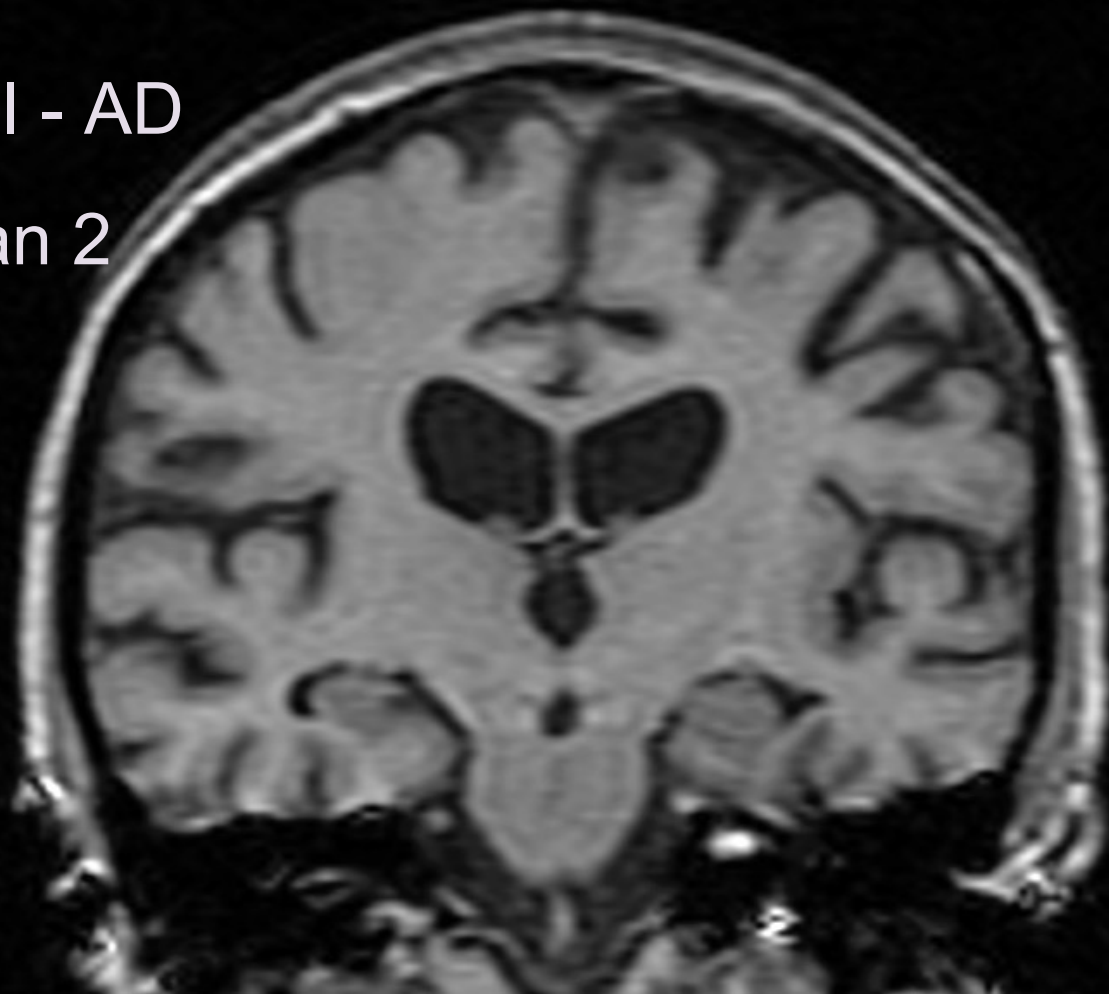
MCI

Scan 1



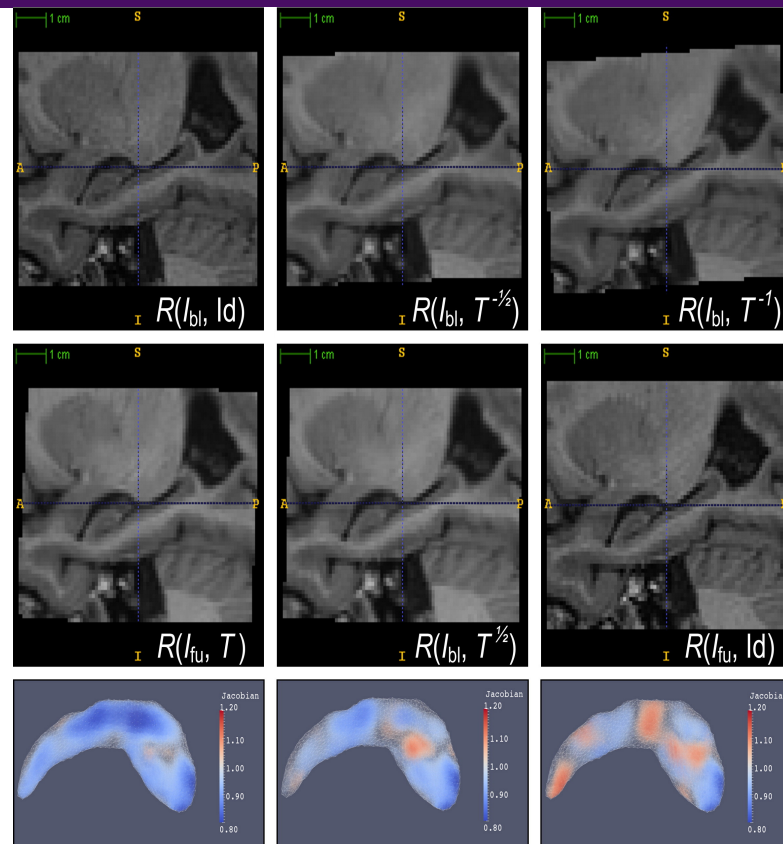
MCI - AD

Scan 2



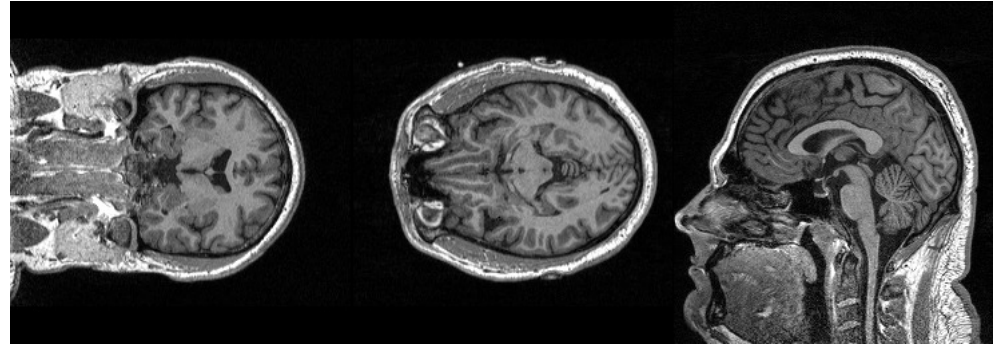
Longitudinal Registration: Treat all images equally

- Registering all follow-up images to baseline means that these images undergo more processing than the baseline scan
- This creates an asymmetry that results in biased measurements of atrophy
- Be sure to use a pipeline that treats all images equally, often creating a “halfway” or “midpoint” space between all timepoints so that all images are treated equally

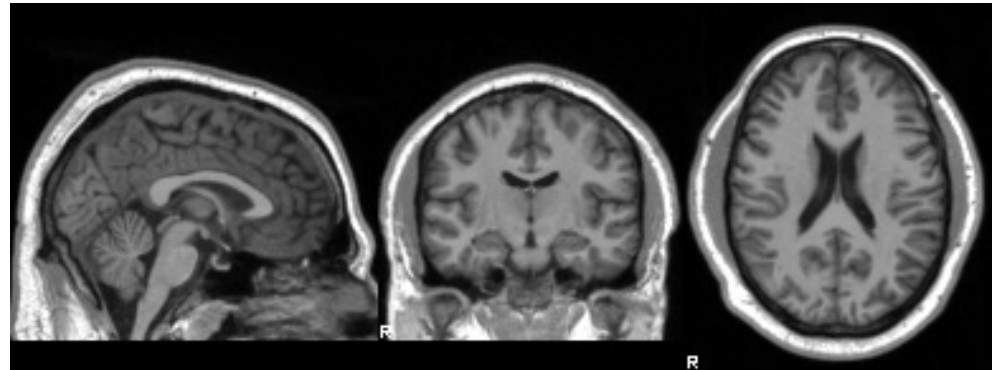


Spatial Normalization: Between-subject

342 Participants - Original

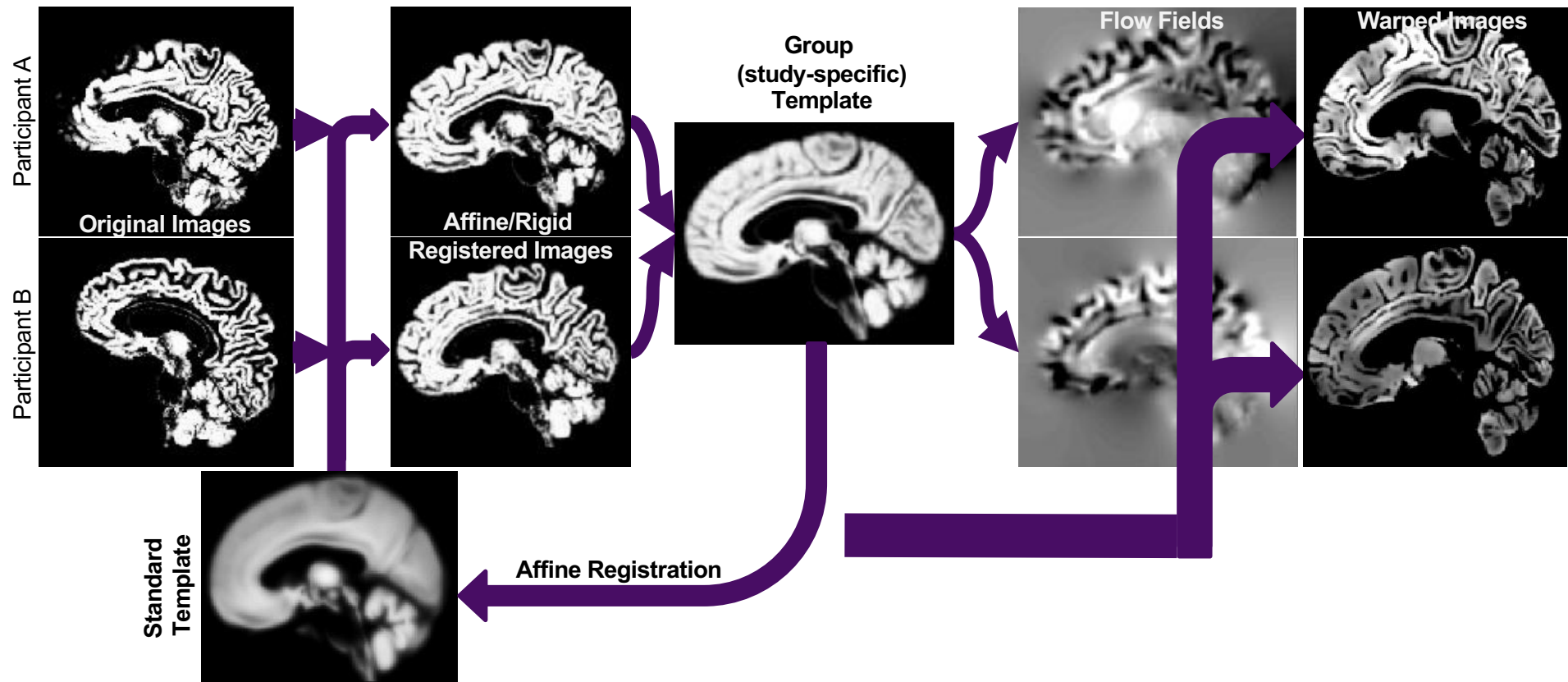


342 Participants - Registered

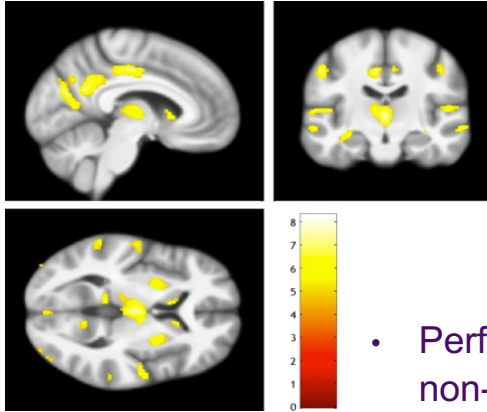


- Warping an individual scan(s) to a population atlas or template.
- After spatial normalization, the same anatomy is in the same area of the image
- The anatomy of every individual is unique, so it is not possible to exactly align images
- Spatially normalizing atlases are based on young, healthy adults to different populations (AD, Down's syndrome, older individuals) can result in greater error and bias

Creating a study-specific template to reduce errors in spatial normalisation

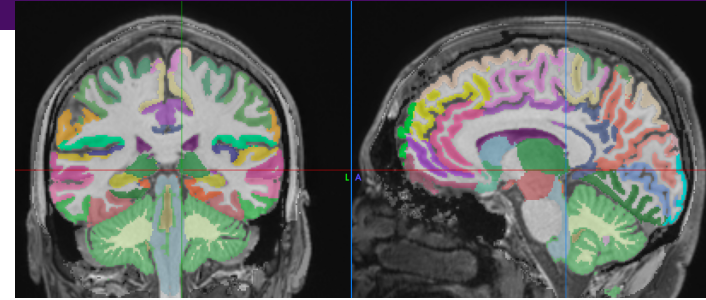


Regional versus voxel/vertex analysis



VOXEL/VERTEX LEVEL

- Perform statistical tests (t-test, F-test, non-parametric) on each voxel/vertex
- Localizes changes to a very high resolution
- Not constrained by anatomical definitions
- Corrections needed for multiple comparisons to control for false positives – reduces sensitivity



REGIONAL LEVEL

- Conventional stats on volume
- Hard segmentation
 - Volume = voxel volume * (# voxels)
- Soft segmentation
 - Volume = voxel volume * $\sum p(x)$
- Averaging over region reduces noise, increase SNR
- Volume differences more interpretable
- Can be performed in native or standard space



What have we learned about dementia from structural MRI?

Please see “Structural and Vascular Imaging” webinars at <https://training.alz.org/Research-Webinars>

1. Data acquisition

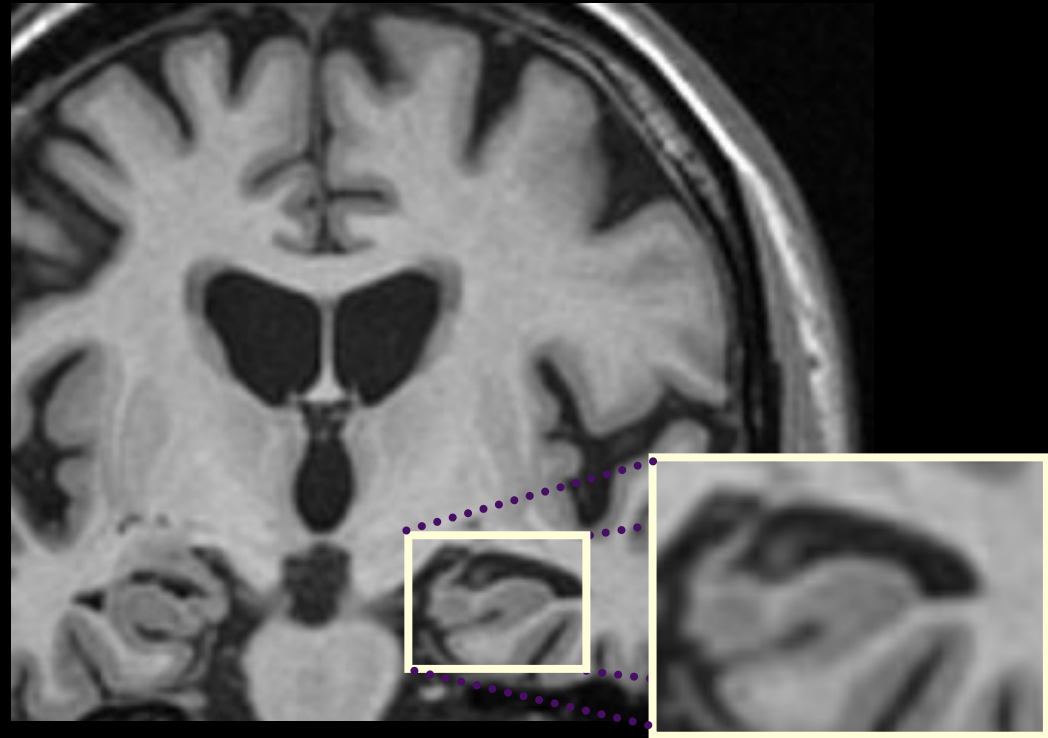
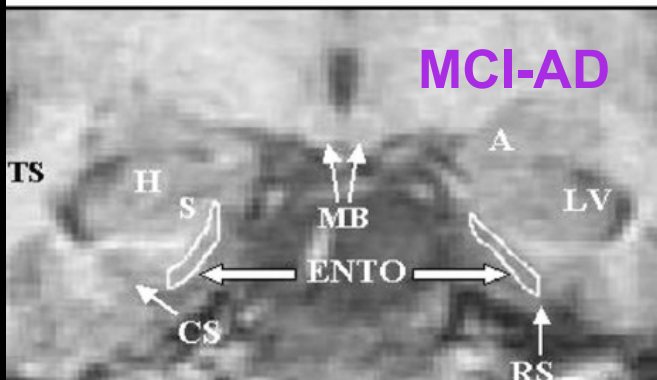
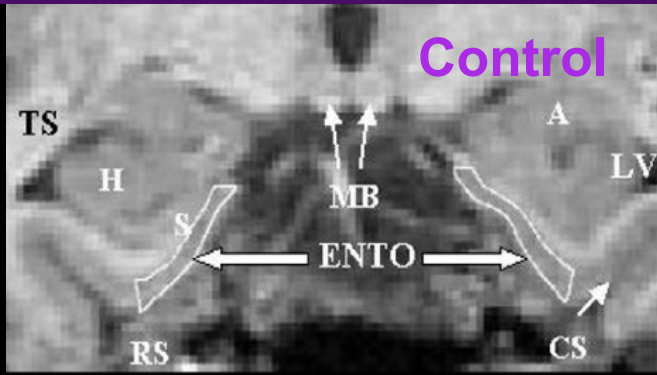
2. Data preprocessing

3. Single-subject analysis

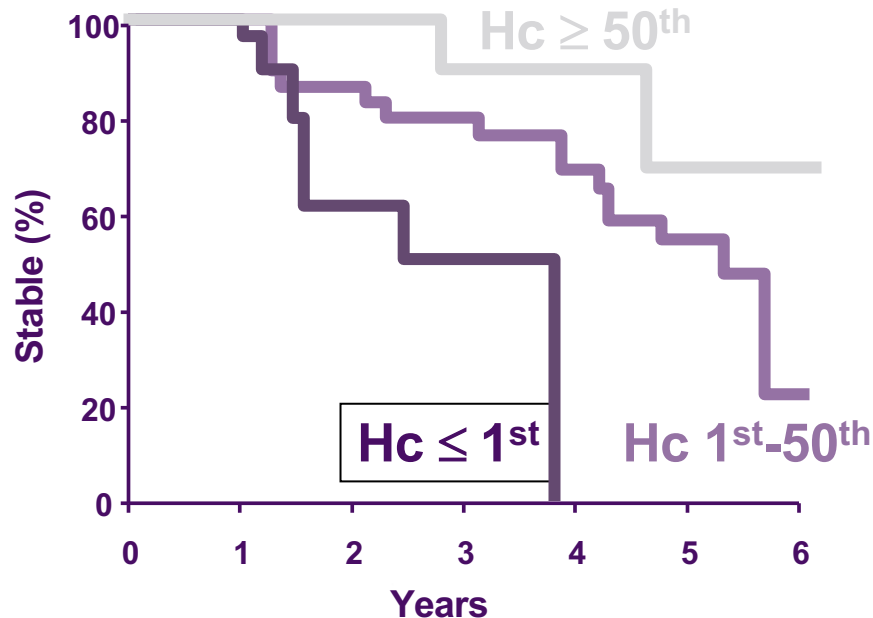
4. Group-level analysis

5. Statistical inference

Entorhinal Cortex and hippocampus are some of the earliest sites of atrophy in AD



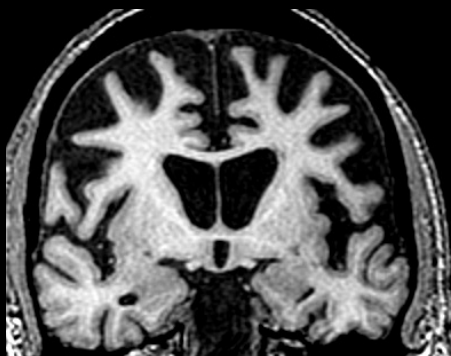
Hippocampal atrophy as a predictor for future decline



- A marker for AD
- A risk factor for progression MCI \rightarrow AD

Jack *et al* **Neurology**: '97, '99; De Leon **AJNR** '93 etc;
Apostolova **Arch Neurol** '06; De Carli **Arch Neurol** '07

Focal atrophy patterns helpful for differential diagnosis



Frontotemporal dementia



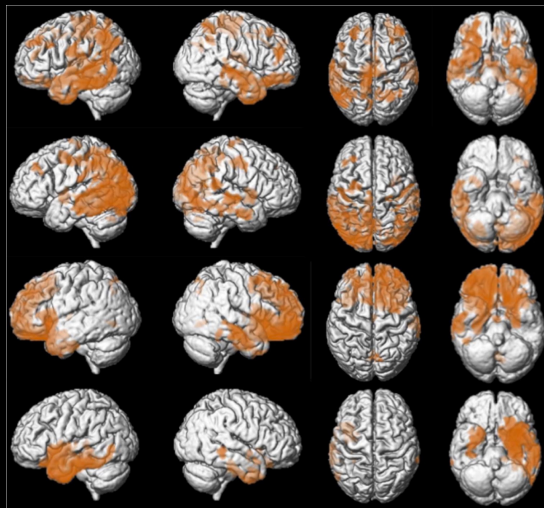
Semantic dementia:
Ubi +ve, tau -ve

Alzheimer's Disease

Posterior Cortical Atrophy

Frontotemporal Dementia

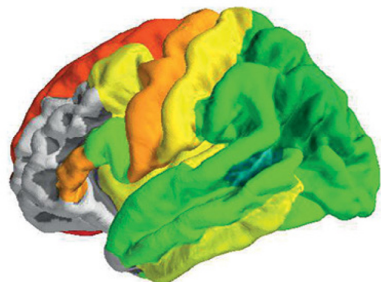
Semantic Dementia



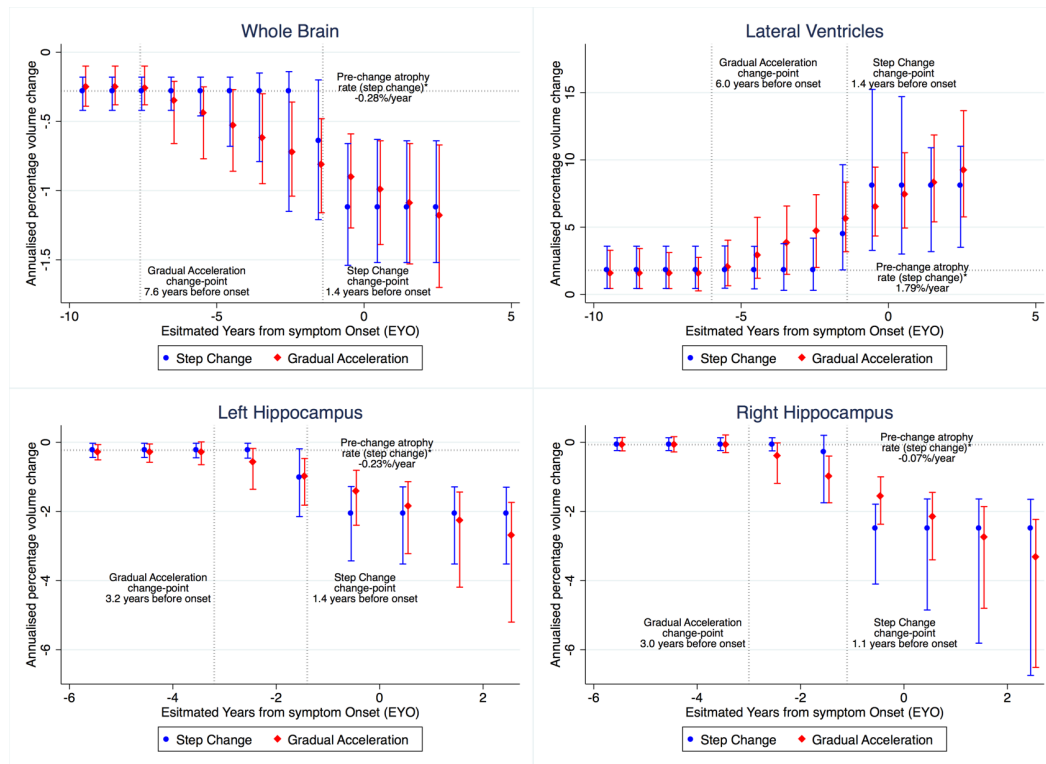
Adapted from Hedderish, et al., *European Radiology*, 2020

Longitudinal MRI detects atrophy before onset of symptoms

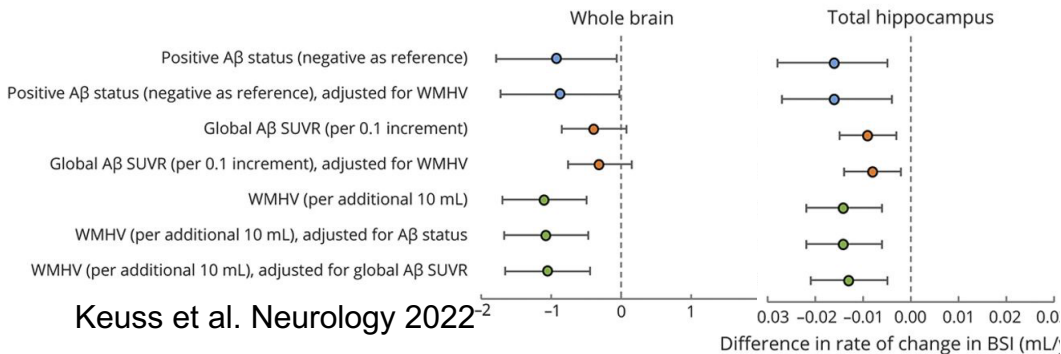
Thickness



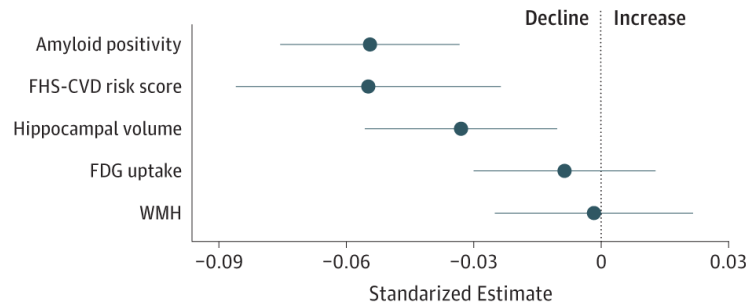
Estimated years to onset



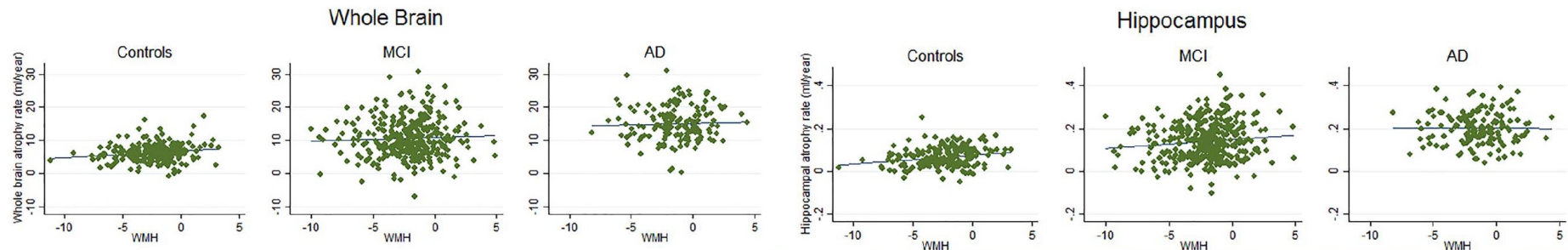
Vascular contributions to AD



Keuss et al. *Neurology* 2022



Rabin et al., *JAMA Neurol*, 2018



Barnes Neurobiology Aging 2013

Fiford Hippocampus 2017

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POP QUIZ!

The earliest structure in the brain where atrophy can be detected is:

- a) Cerebellum
- b) Hippocampus
- c) Posterior Cingulate
- d) Entorhinal Cortex
- e) Fusiform Gyrus

The process of bias correction involves:

- a) Removing a slowly varying intensity inhomogeneity caused by small imperfections in the magnetic field
- b) Removes noise and differences across scanners
- c) Correct for systematic changes when comparing scans between different scanners

Which one of these statements is true about spatial normalization?

- a) It is important to get all of the image to line up exactly
- b) Study-specific templates that are representative of the participants in the study reduces potential errors
- c) Affine registrations are sufficient for spatial normalization.

THANK YOU!



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Don't miss:

Basics of Neuroimaging: Data structure and formats by Ludovica Griffanti

On demand at <https://training.alz.org/>

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GETTING STARTED WITH NEUROIMAGING WORKSHOP Friday July 14 8:00-12:00 Amsterdam

