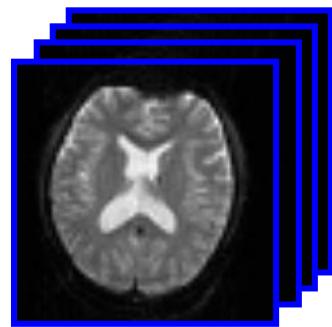


# Preprocessing II: Between Subjects

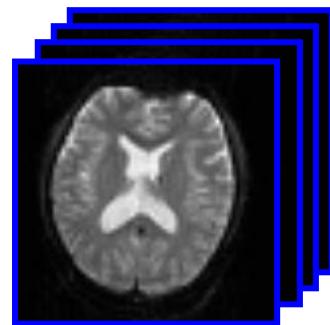
*John Ashburner*

# Pre-processing Overview

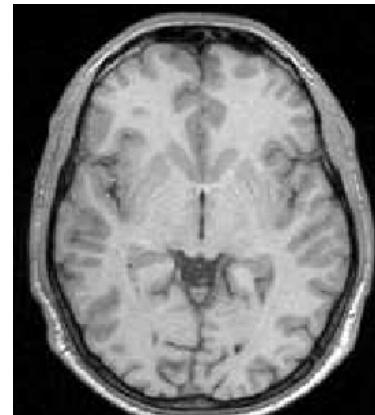
fMRI time-series



Motion Correct



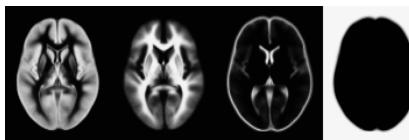
Anatomical MRI



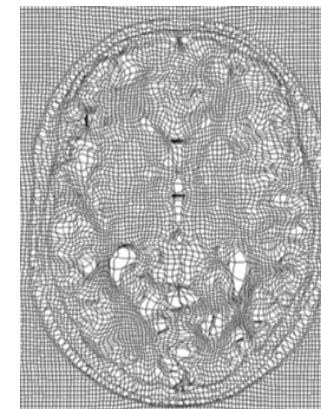
Coregister

$$\begin{pmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ m_{21} & m_{22} & m_{23} & m_{24} \\ m_{31} & m_{32} & m_{33} & m_{34} \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

Template



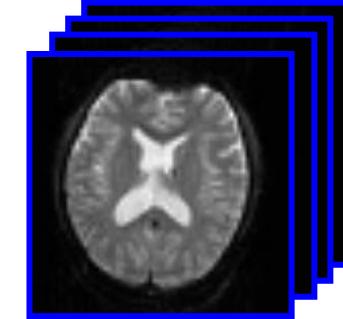
Estimate Spatial Norm



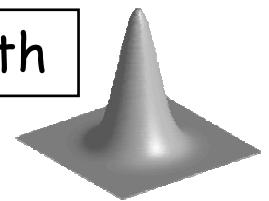
Deformation

Statistics or whatever

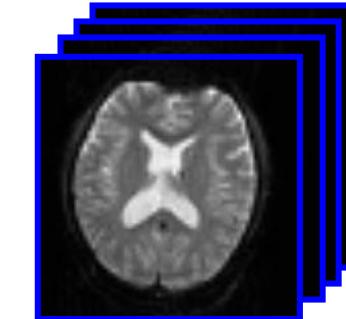
Smoothed



Smooth



Spatially normalised



# Spatial Normalisation

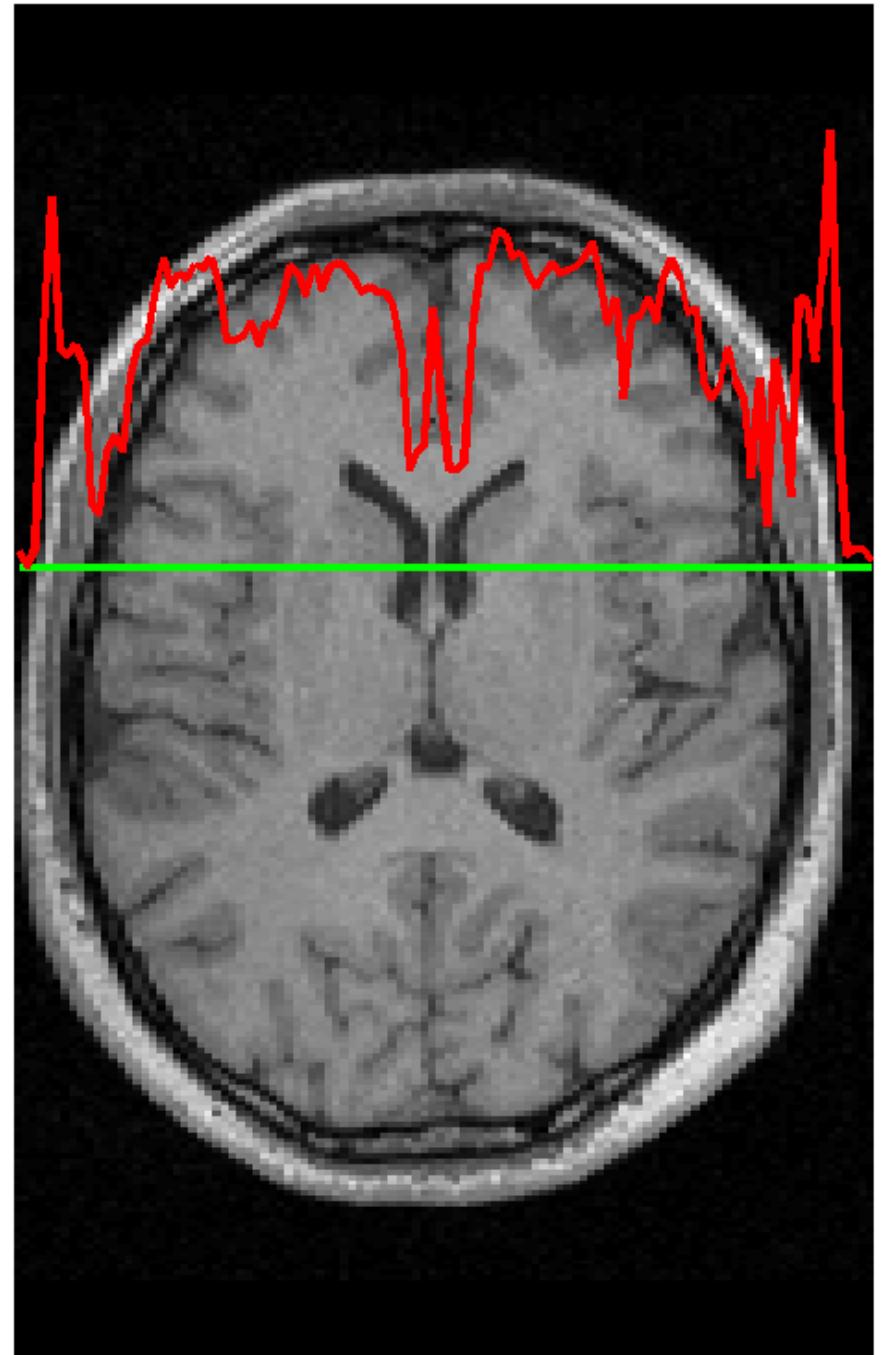
- \* Brains of different subjects vary in shape and size.
- \* Need to bring them all into a common anatomical space.
  - \* Examine homologous regions across subjects
    - \* Improve anatomical specificity
    - \* Improve sensitivity
  - \* Report findings in a common anatomical space (eg MNI space)
- \* In SPM, alignment is achieved by matching grey matter with grey matter and white matter with white matter.
  - \* Need to segment.

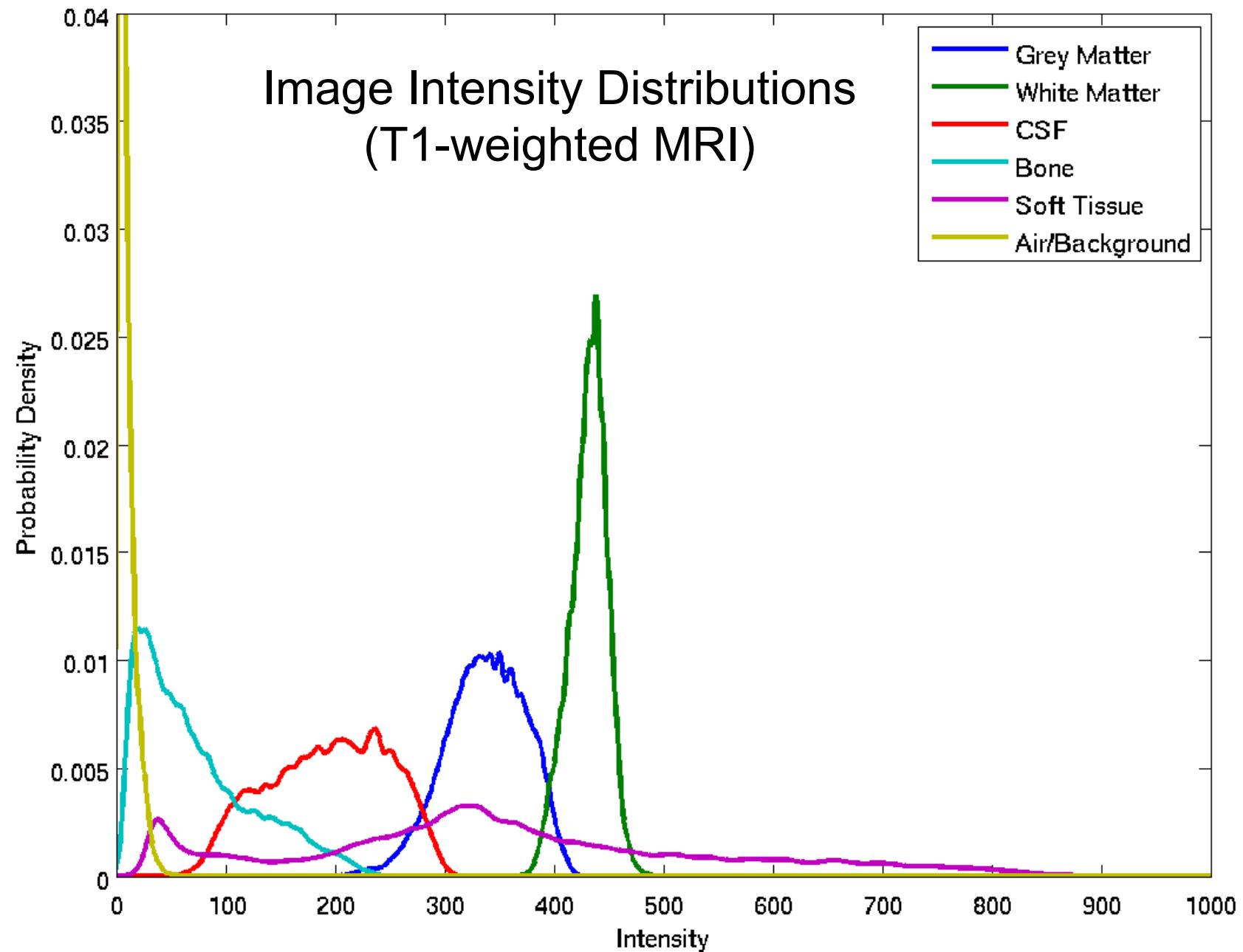
# Contents

- \* **Segment**
  - Use segmentation routine for spatial normalisation**
    - \* **Gaussian mixture model**
    - \* **Intensity non-uniformity correction**
    - \* **Deformed tissue probability maps**
- \* Dartel
- \* Smoothing

# Segmentation

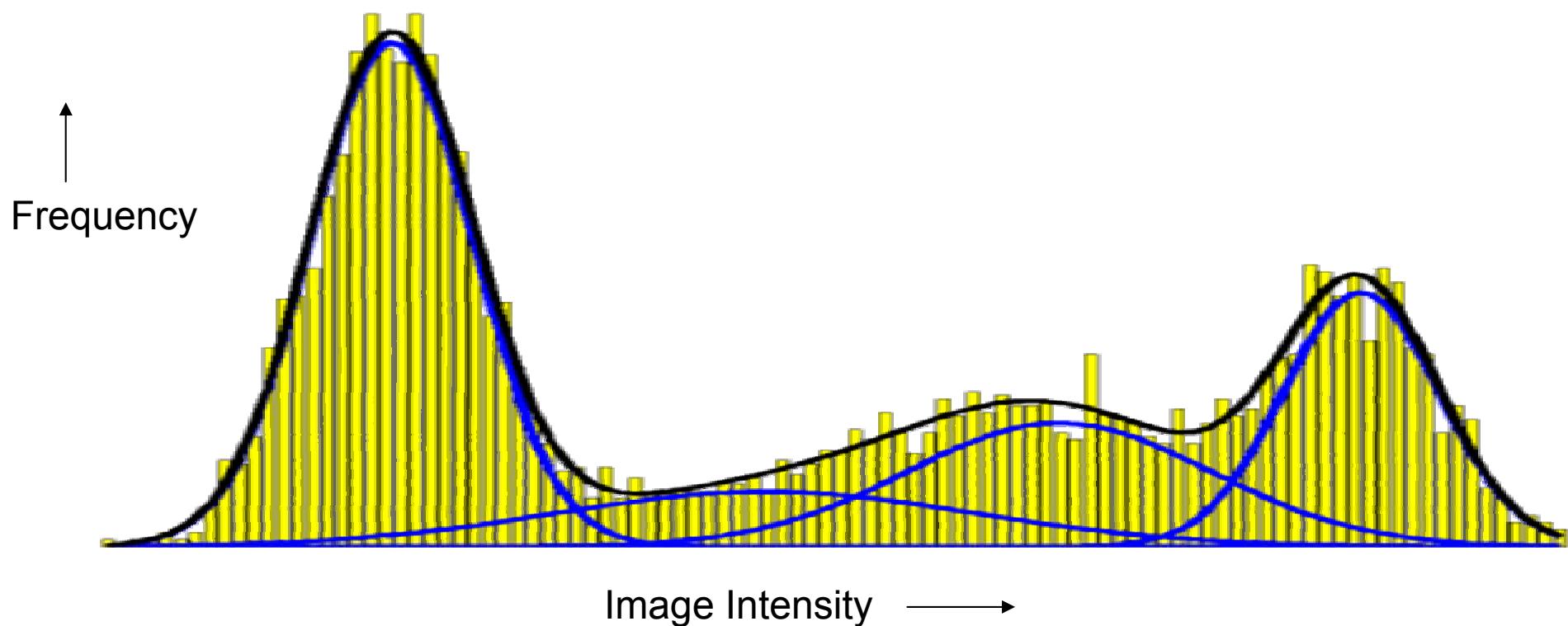
- \* Segmentation in SPM8 also estimates a spatial transformation that can be used for spatially normalising images.
- \* It uses a **generative model**, which involves:
  - \* Mixture of Gaussians (MOG)
  - \* Bias Correction Component
  - \* Warping (Non-linear Registration) Component





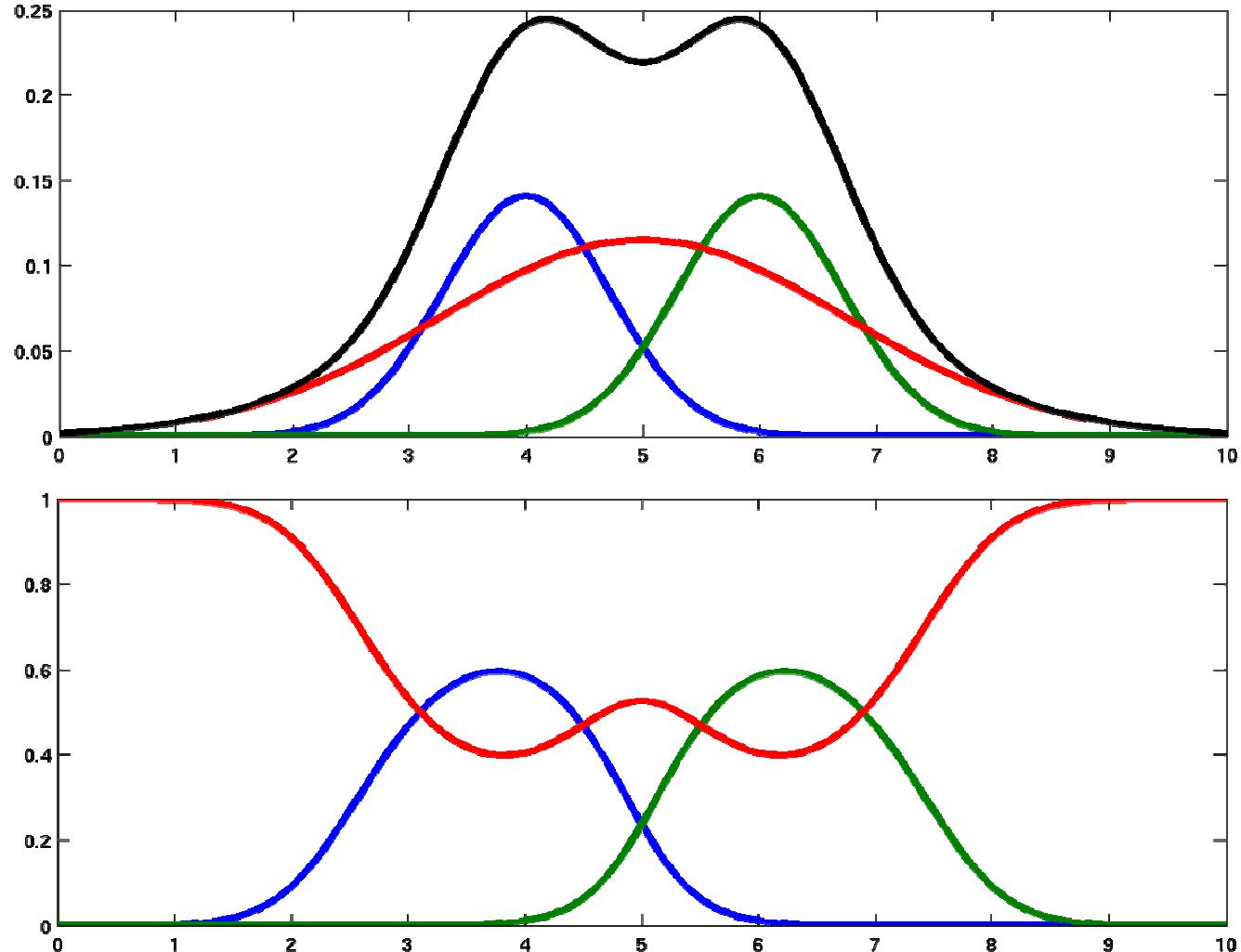
# Mixture of Gaussians (MOG)

- \* Classification is based on a Mixture of Gaussians model (MOG), which represents the intensity probability density by a number of Gaussian distributions.



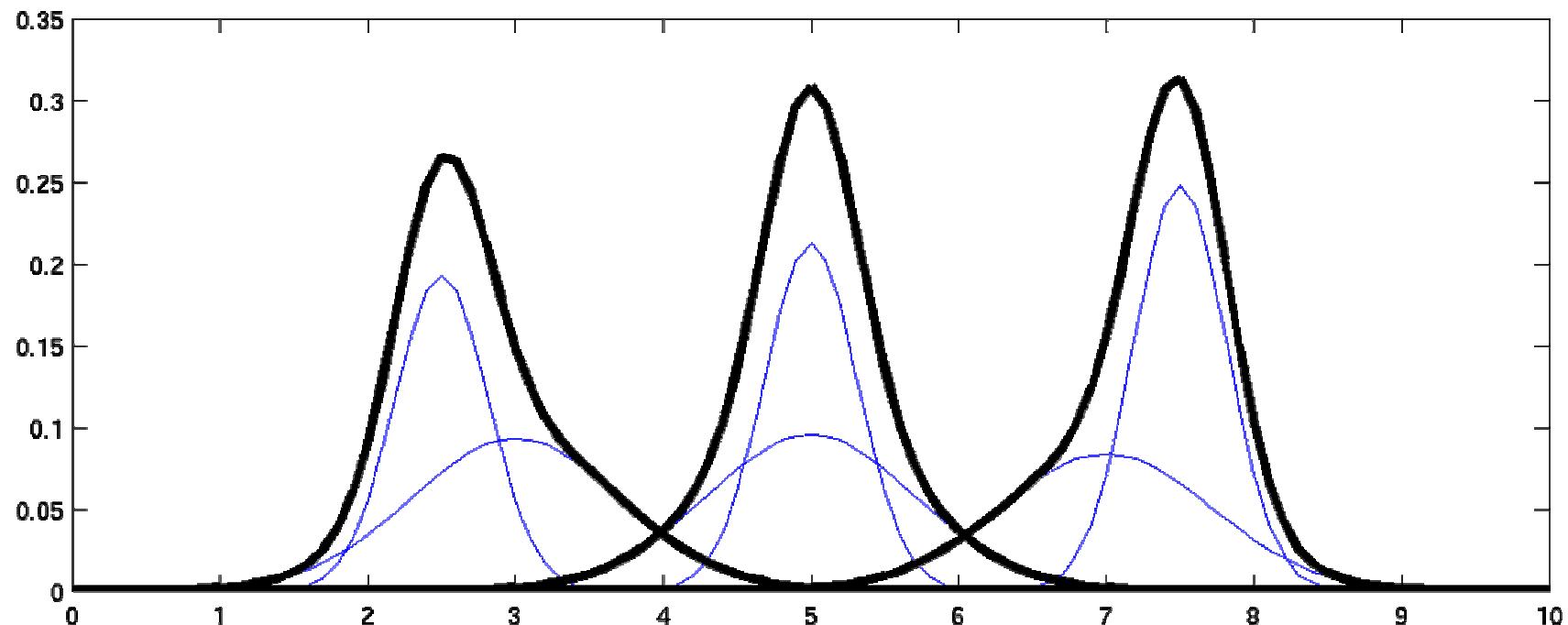
# Belonging Probabilities

Belonging probabilities are assigned by normalising to one.



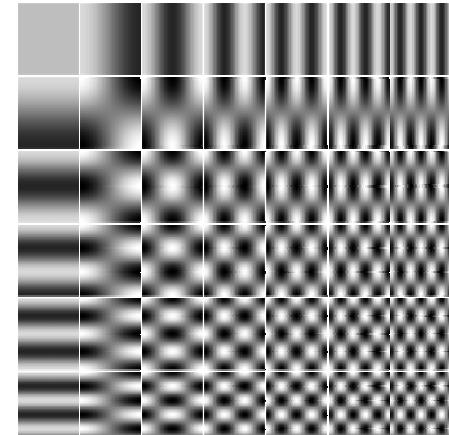
# Non-Gaussian Intensity Distributions

- \* Multiple Gaussians per tissue class allow non-Gaussian intensity distributions to be modelled.
  - \* E.g. accounting for partial volume effects

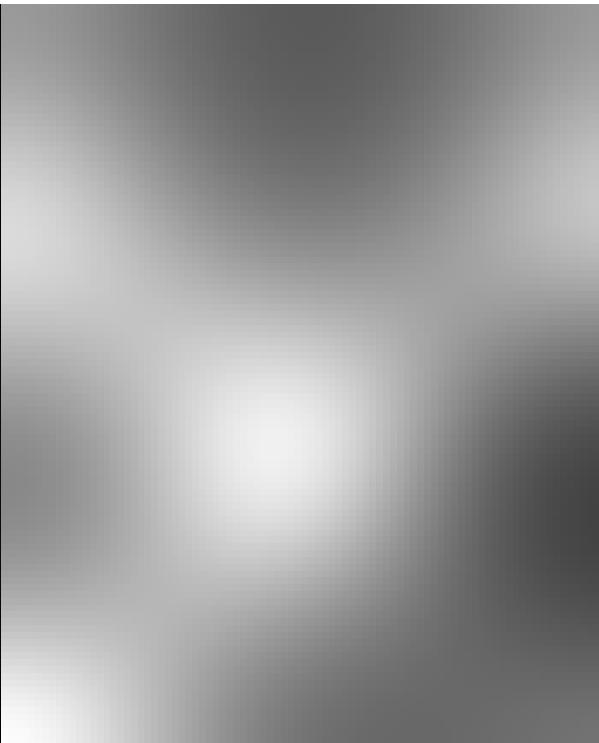


# Modelling a Bias Field

- \* A bias field is modelled as a linear combination of basis functions.



Corrupted image



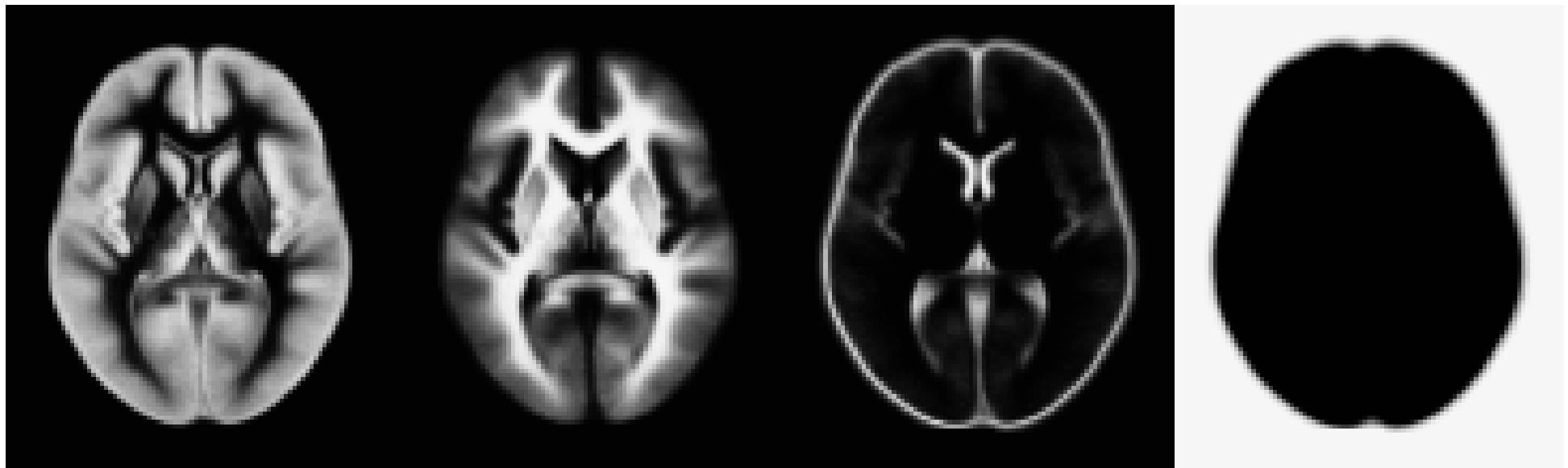
Bias Field



Corrected image

# Tissue Probability Maps

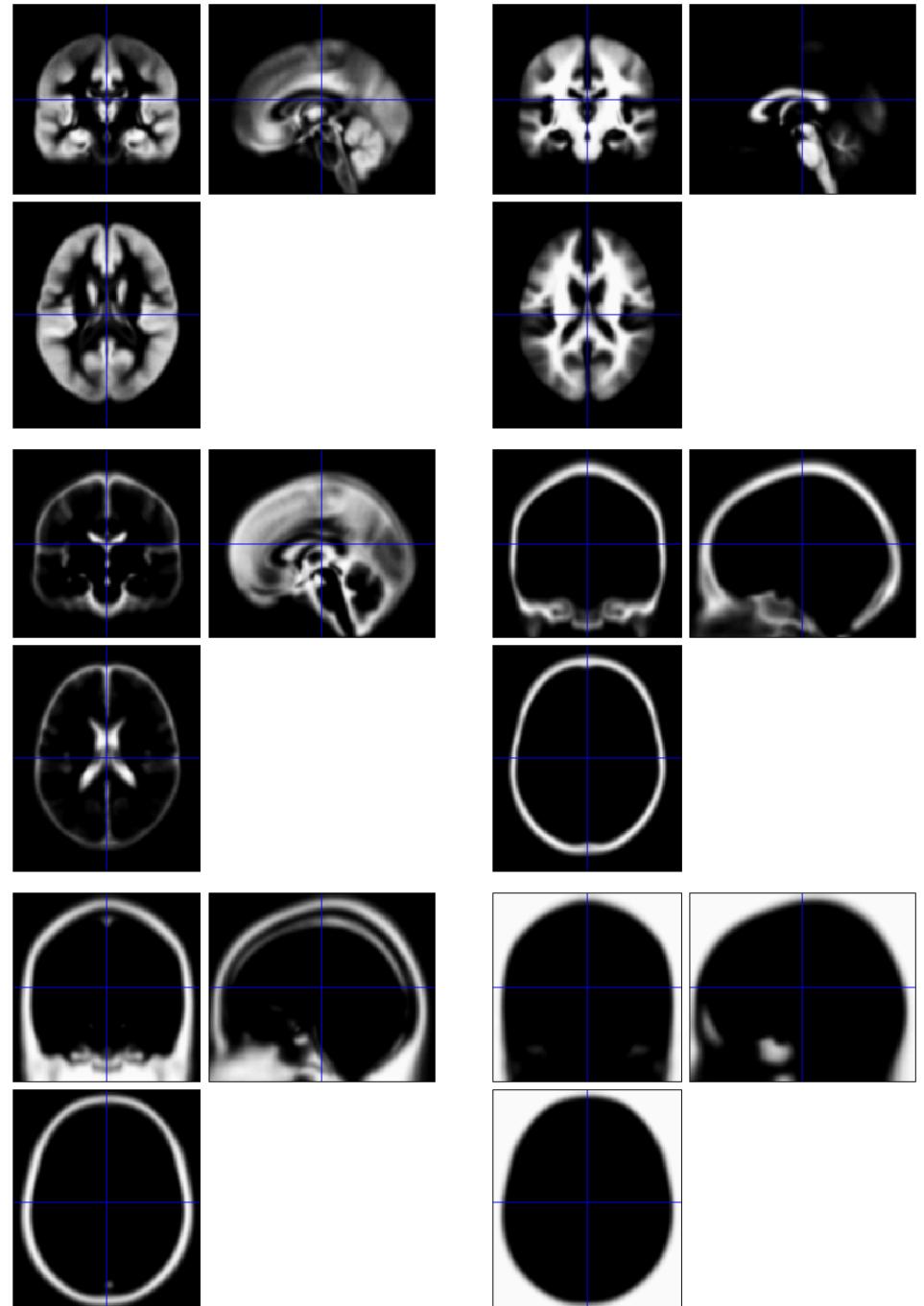
- \* Tissue probability maps (TPMs) are used instead of the proportion of voxels in each Gaussian as the prior.

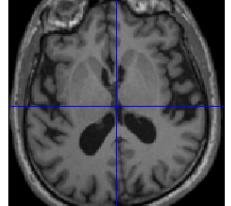
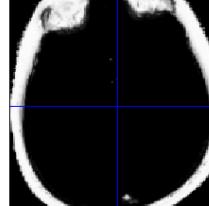
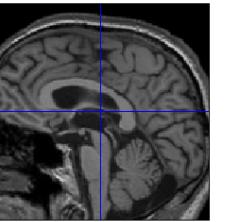
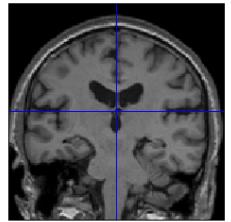
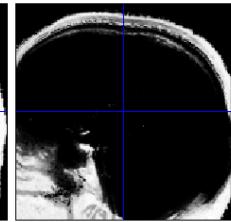
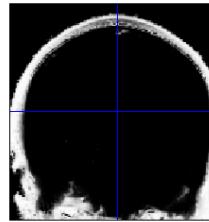
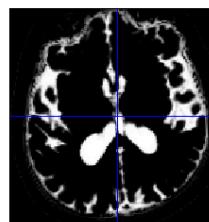
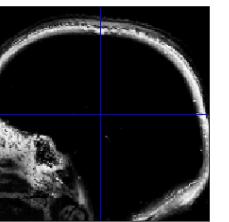
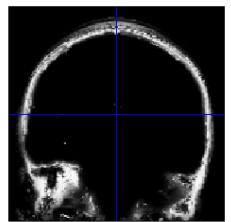
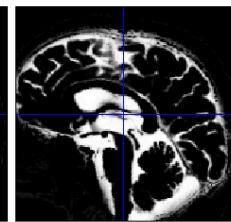
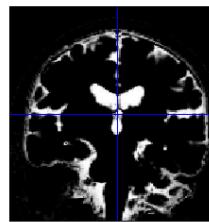
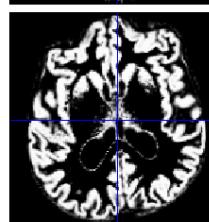
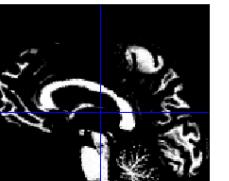
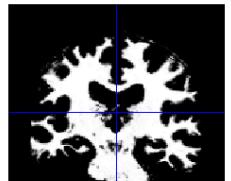
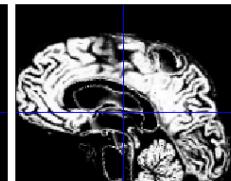
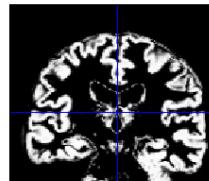
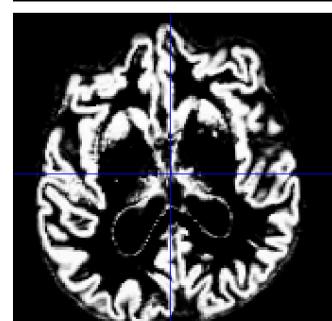
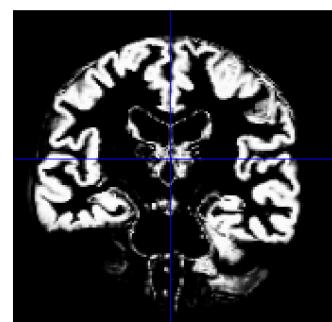
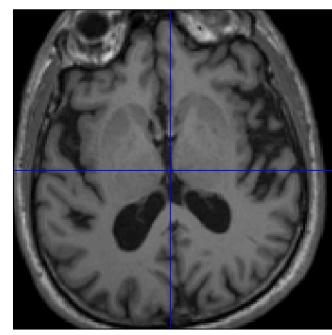
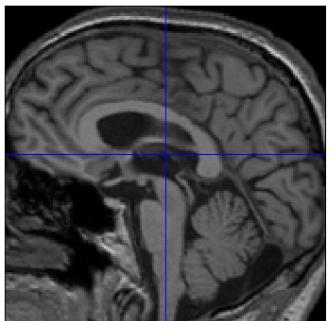
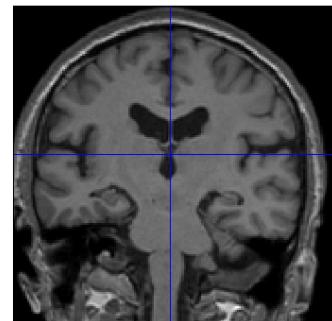


**ICBM Tissue Probabilistic Atlases.** These tissue probability maps are kindly provided by the International Consortium for Brain Mapping, John C. Mazziotta and Arthur W. Toga.

# Tissue Probability Maps for “New Segment”

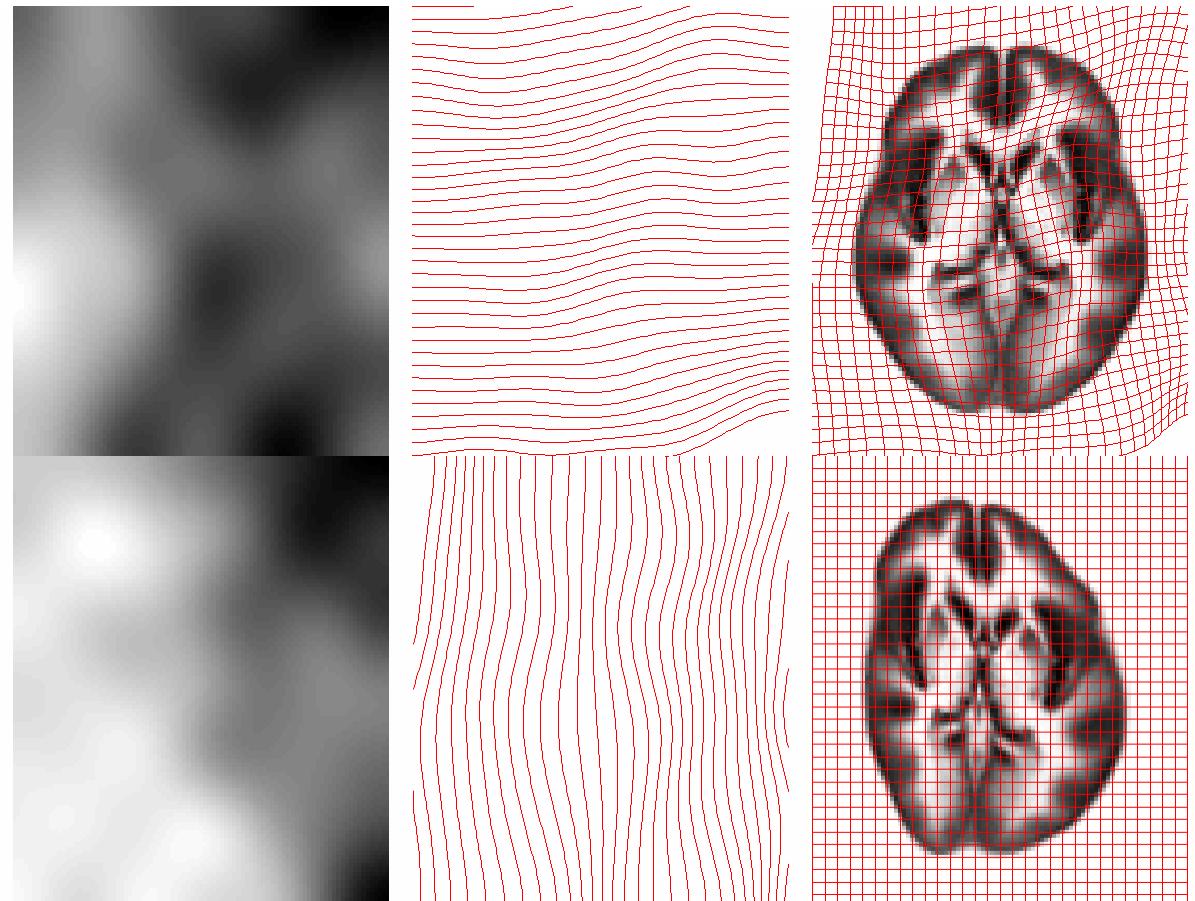
Includes additional non-brain tissue  
classes (bone, and soft tissue)





# Deforming the Tissue Probability Maps

- \* Tissue probability images are deformed so that they can be overlaid on top of the image to segment.

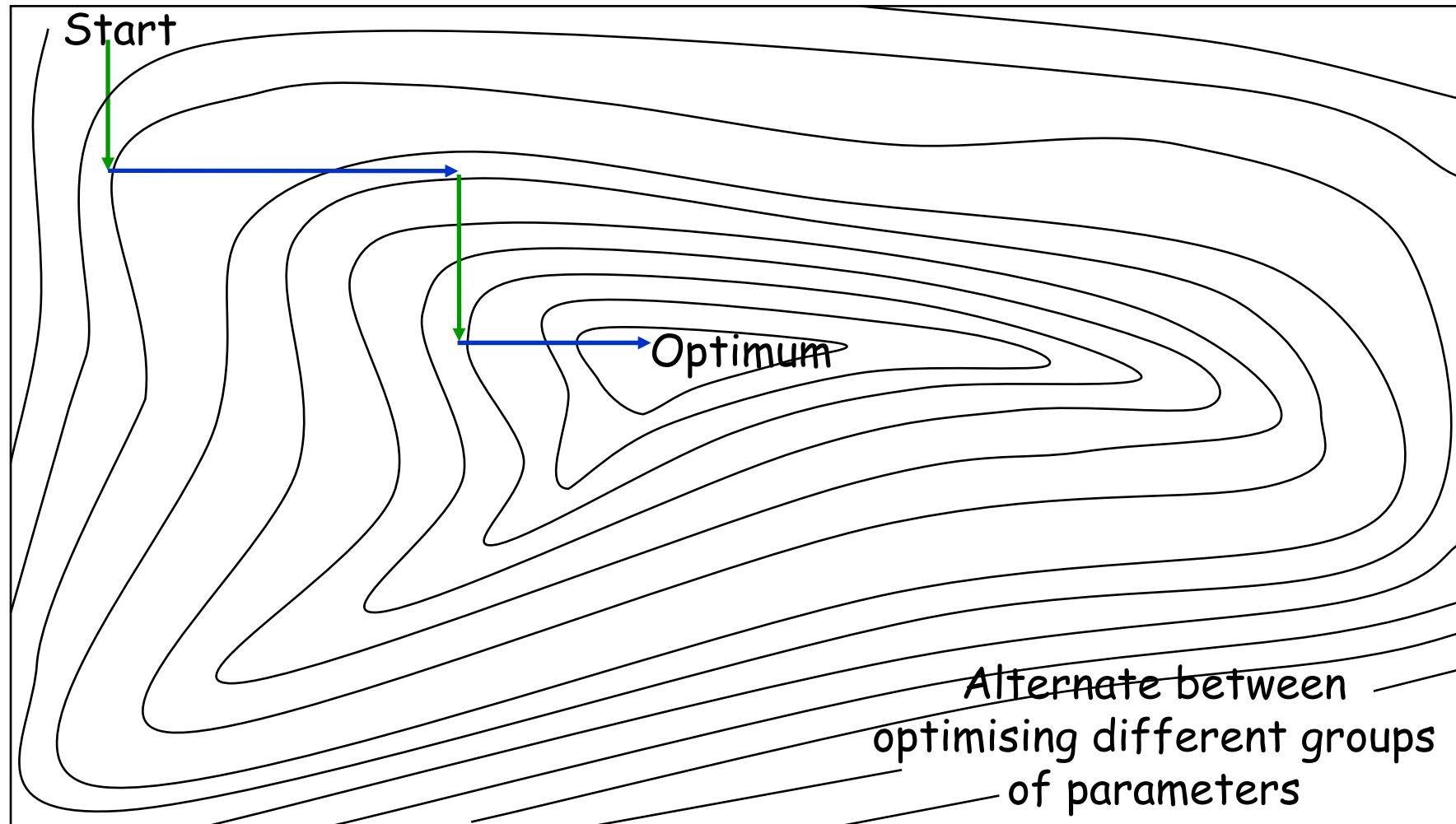


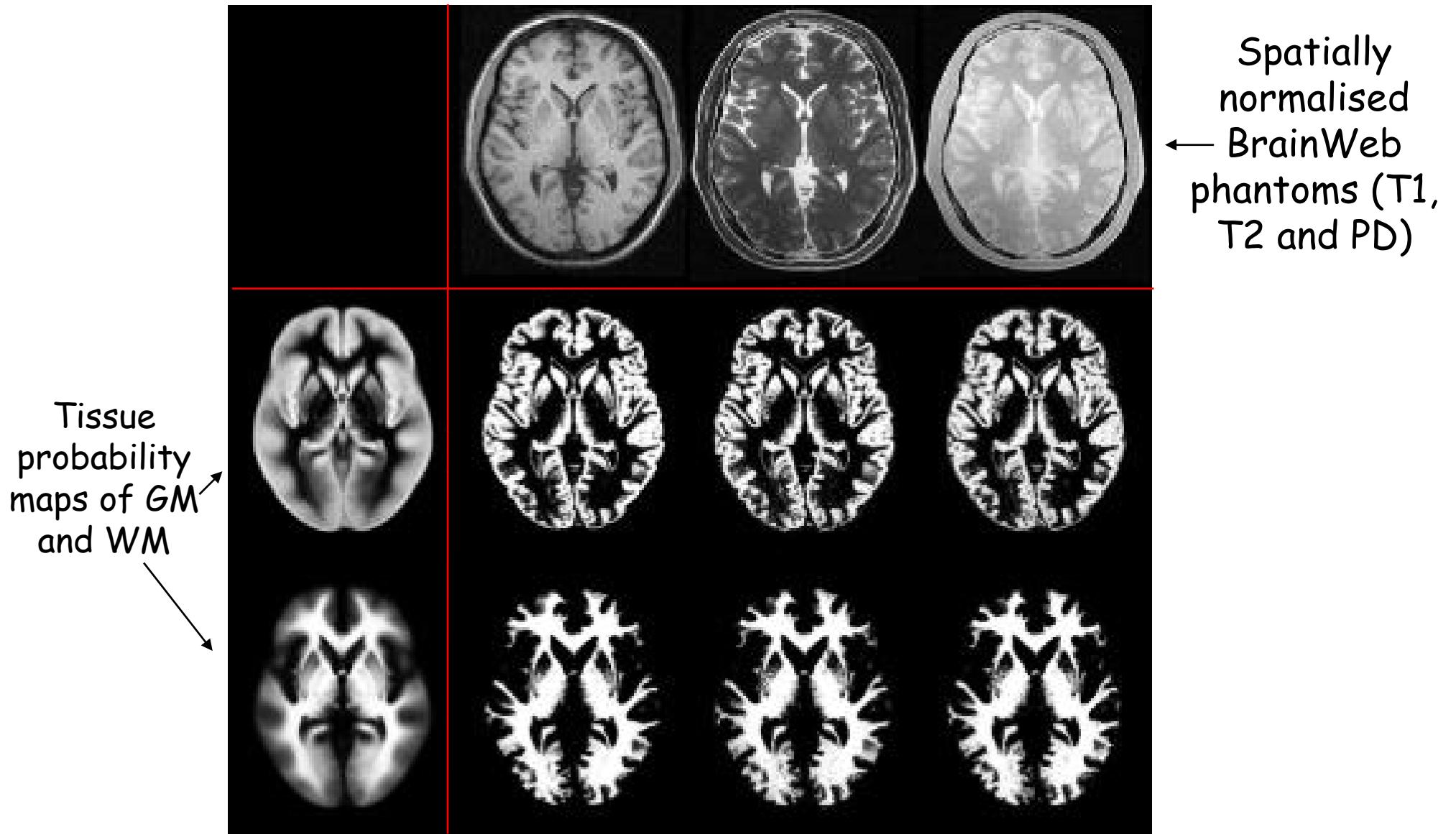
# Optimisation

- \* The “best” parameters are those that minimise this objective function.
- \* Optimisation involves finding them.
- \* Begin with starting estimates, and repeatedly change them so that the objective function decreases each time.

$$E = -\sum_{i=1}^I \log \left[ p_i \beta \sum_{k=1}^K \frac{\gamma_k b_{ik}(\alpha)}{\sum_{j=1}^K \gamma_j b_{ij}(\alpha)} \frac{1}{\sqrt{2\pi\sigma_k^2}} \exp \left( -\frac{(p_i \beta y_i - \mu_k)^2}{2\sigma_k^2} \right) \right]$$

# Steepest Descent





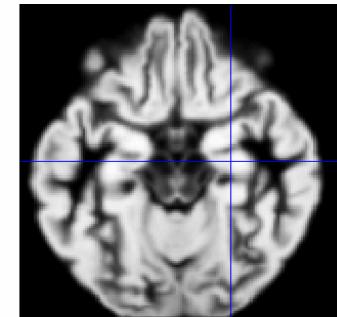
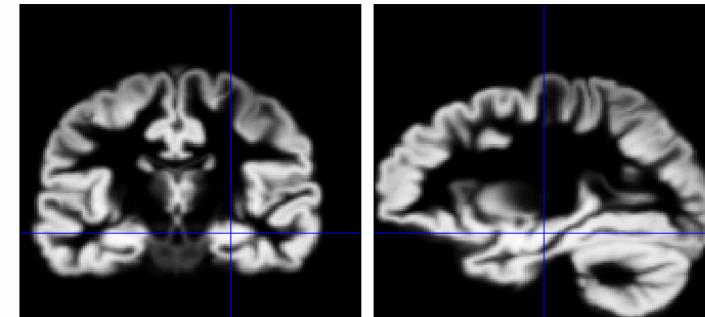
Cocosco, Kollokian, Kwan & Evans. "BrainWeb: Online Interface to a 3D MRI Simulated Brain Database". NeuroImage 5(4):S425 (1997)

# Contents

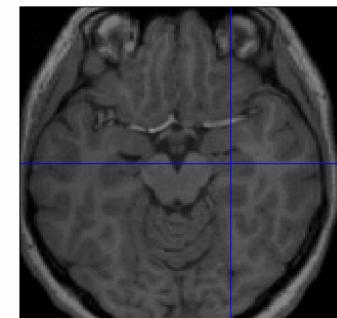
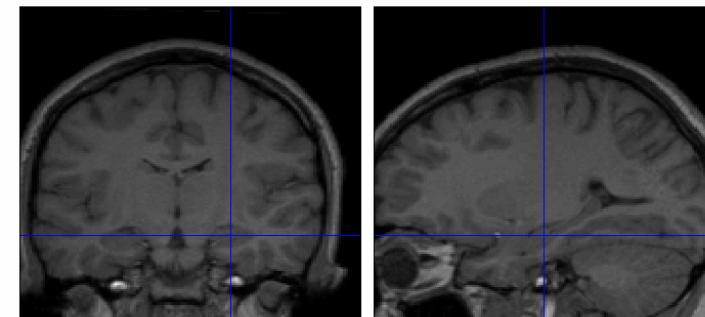
- \* Segment
- \* **DARTEL**
  - \* **Flow field parameterisation**
  - \* **Objective function**
  - \* **Template creation**
  - \* **Examples**
- \* Smooth

# DARTEL Image Registration

- \* Uses fast approximations
  - \* Deformation integrated using scaling and squaring
- \* Uses Levenberg-Marquardt optimiser
  - \* Multi-grid matrix solver
- \* Matches GM with GM, WM with WM etc
- \* Diffeomorphic registration takes about 30 mins per image pair (121×145×121 images).

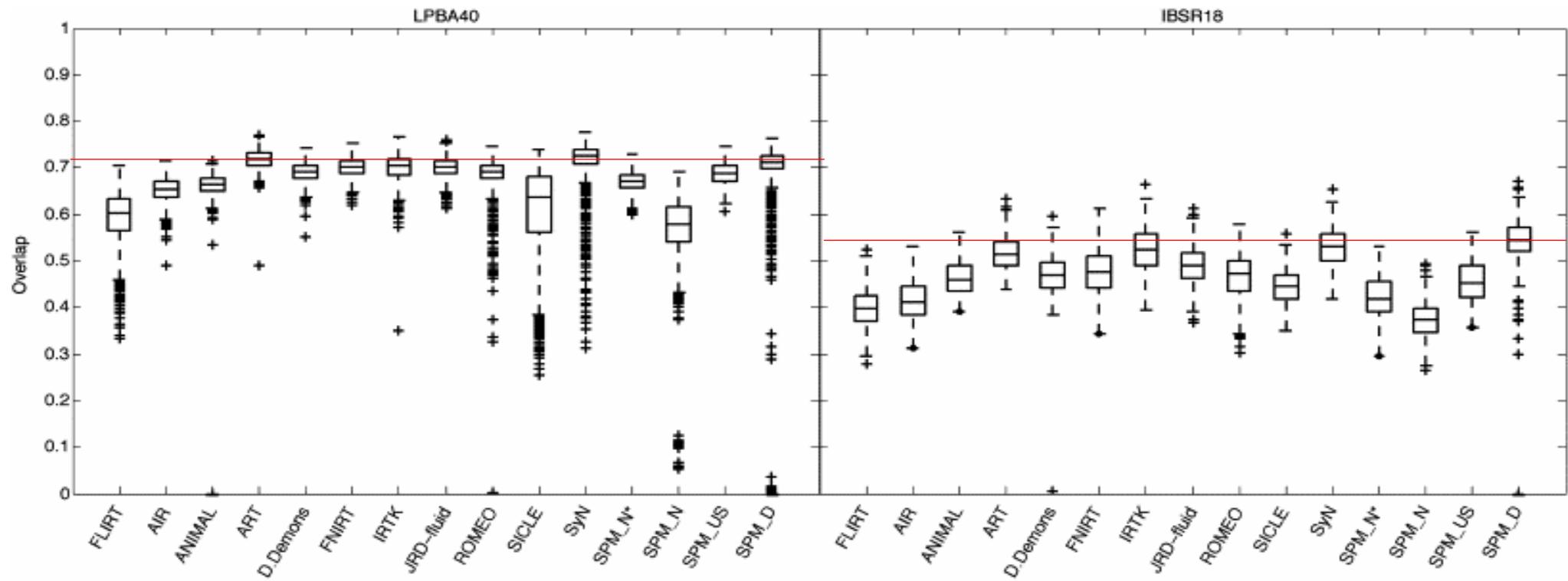
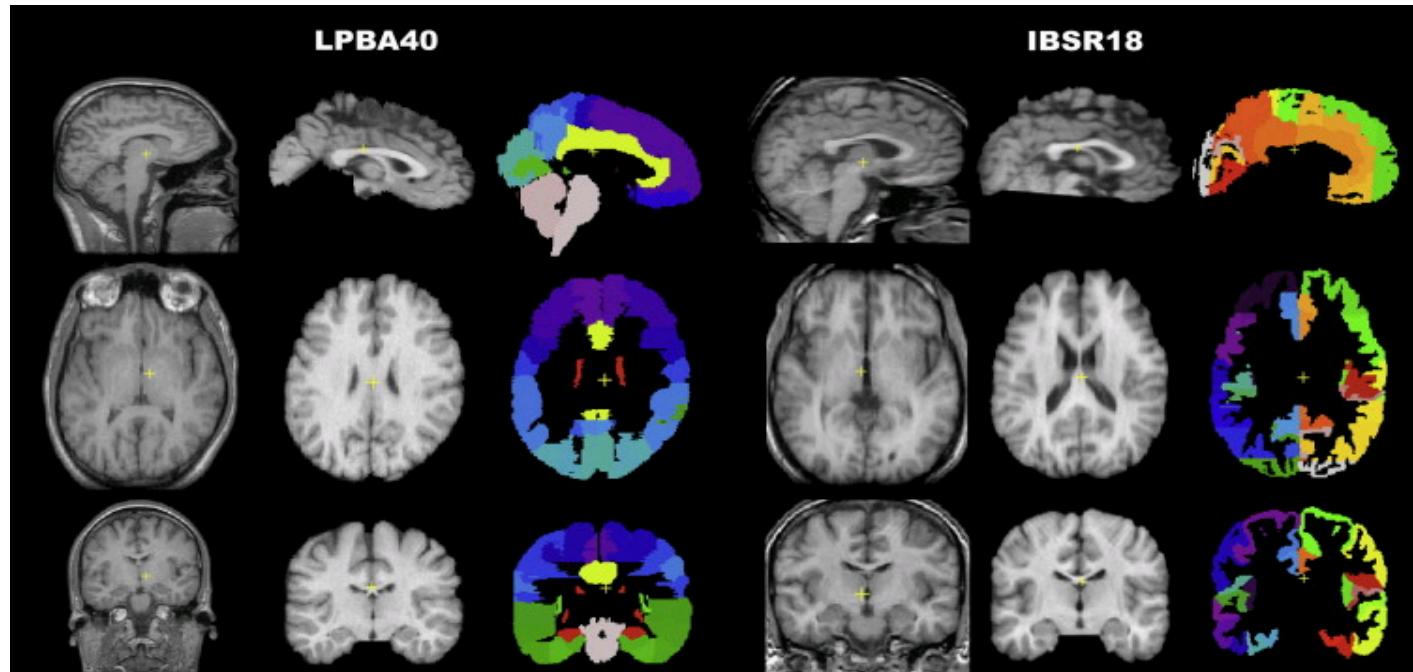


Grey matter template warped to individual



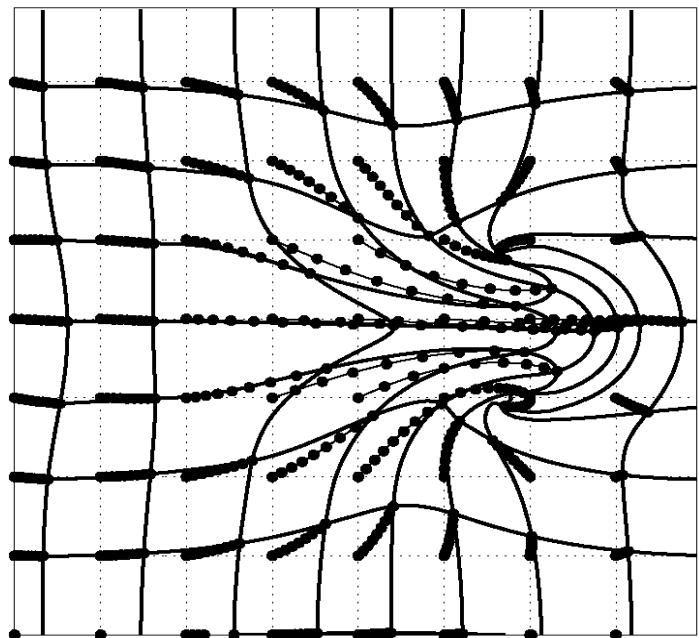
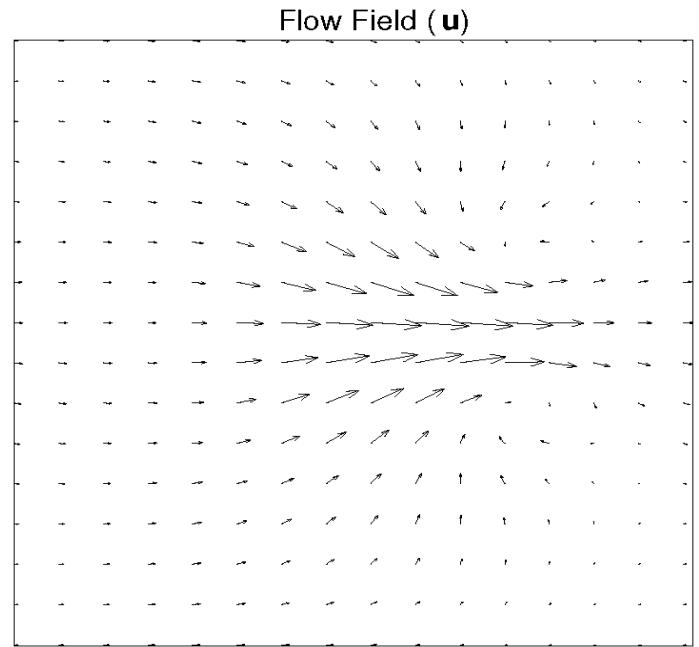
Individual scan

# Evaluations of nonlinear registration algorithms



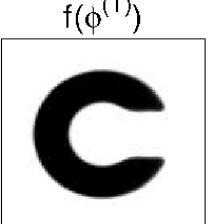
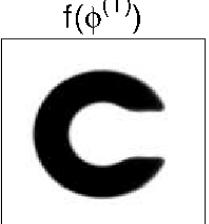
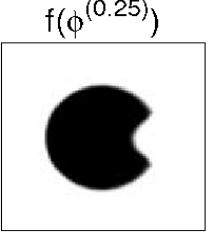
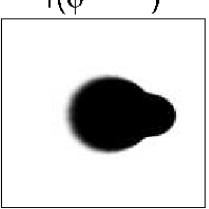
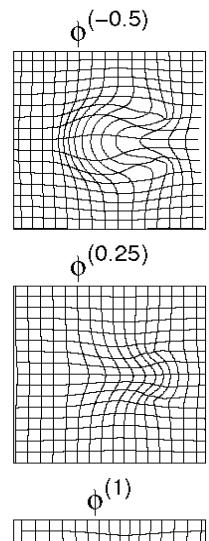
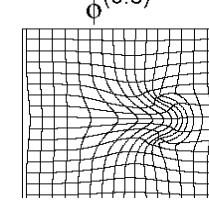
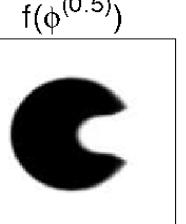
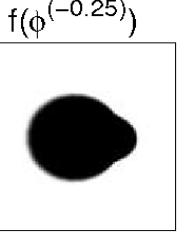
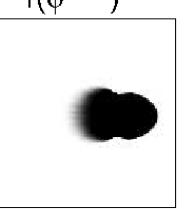
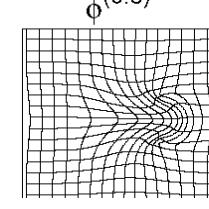
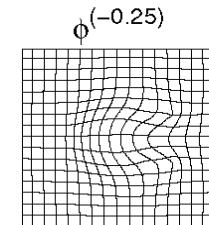
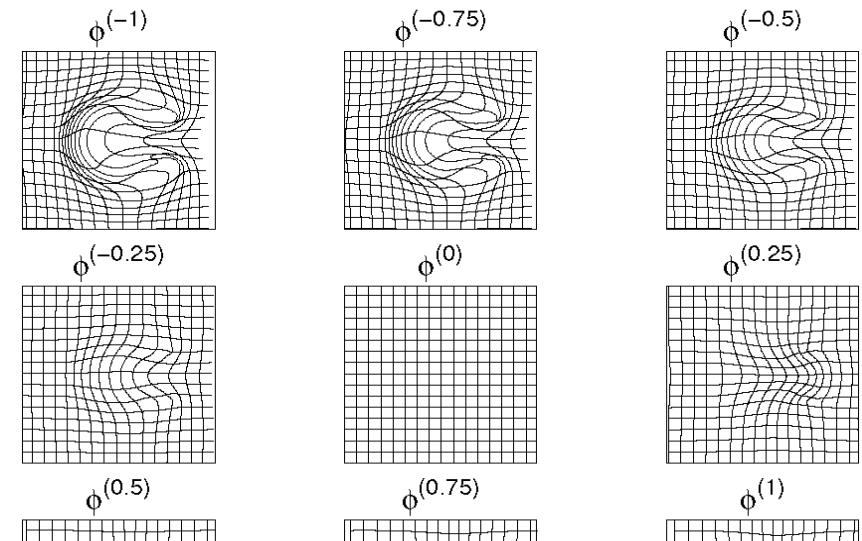
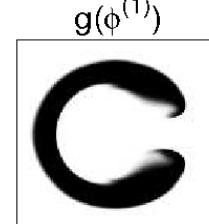
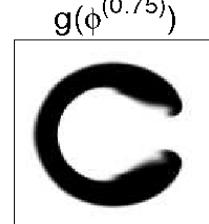
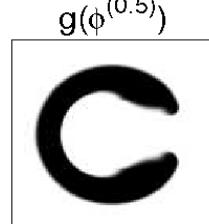
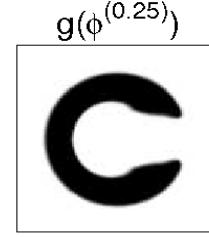
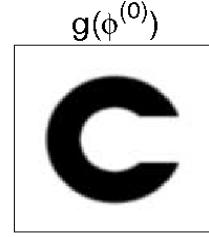
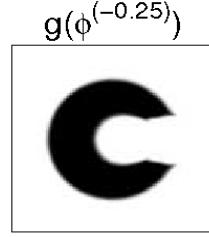
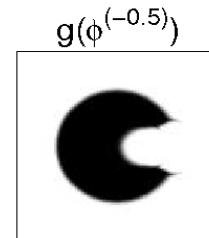
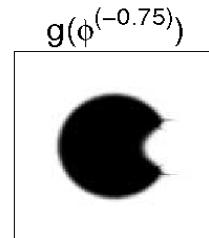
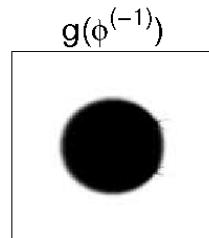
# DARTEL

- \* Parameterising the deformation
- \*  $\varphi^{(0)} = \text{Identity}$
- \*  $\varphi^{(1)} = \int_{t=0}^1 u(\varphi^{(t)}) dt$
- \*  $u$  is a flow field to be estimated
- \* Scaling and squaring is used to generate deformations.





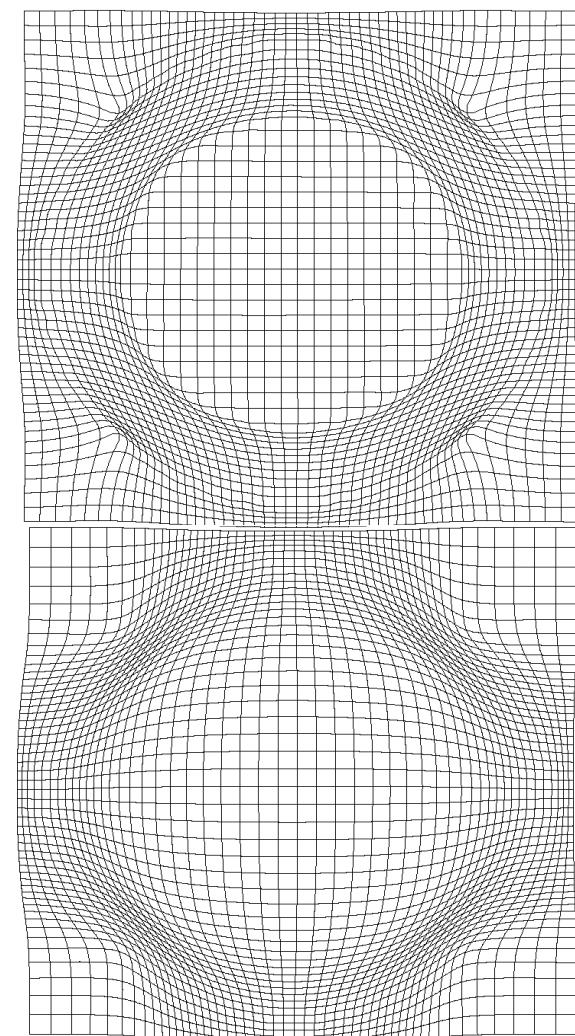
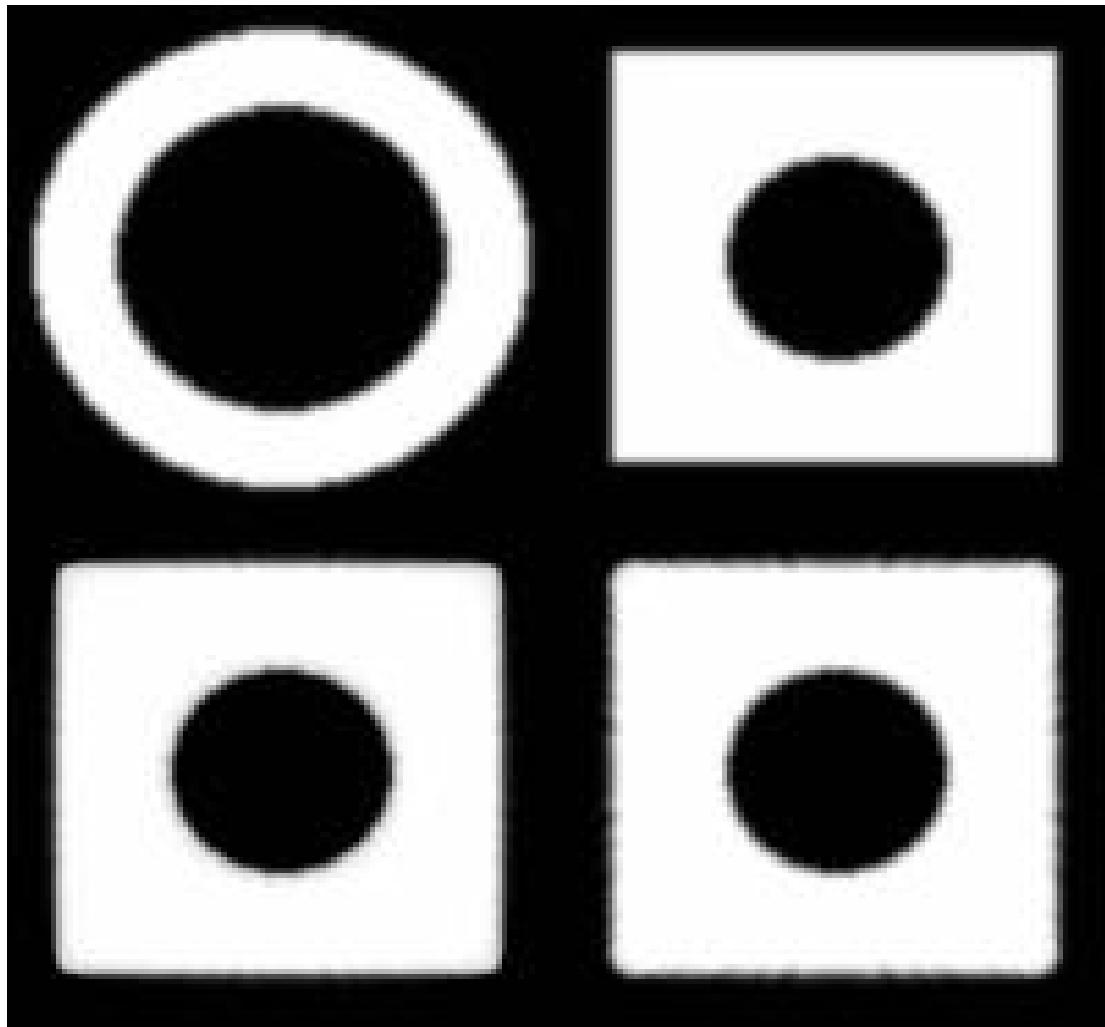
# Forward and backward transforms



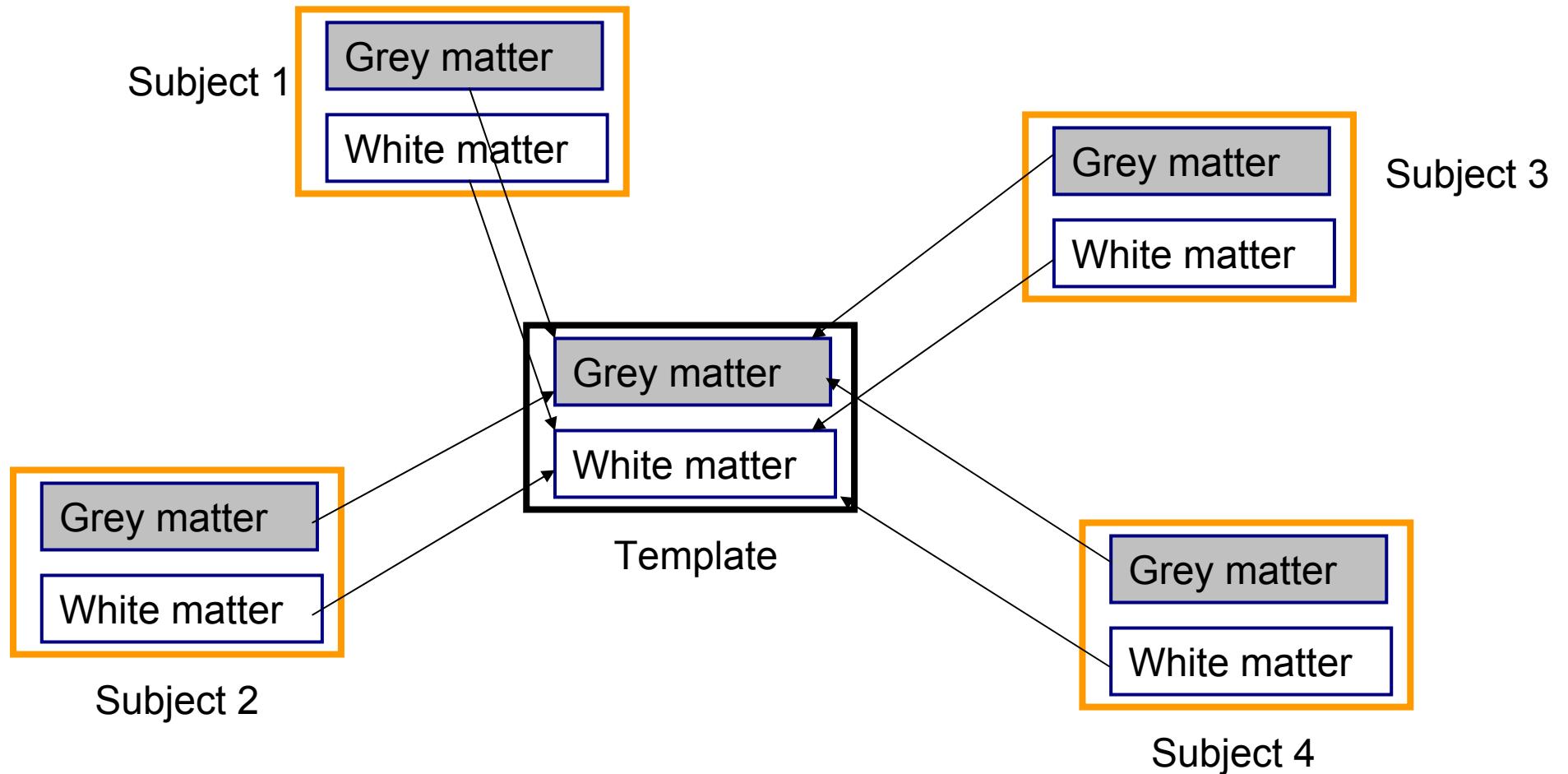
# Registration objective function

- \* Simultaneously minimize the sum of:
  - \* Matching Term
    - \* Drives the matching of the images.
    - \* Multinomial assumption
  - \* Regularisation term
    - \* A measure of deformation roughness
    - \* Regularises the registration.
- \* A balance between the two terms.

# Effect of Different Regularisation Terms



# Simultaneous registration of GM to GM and WM to WM



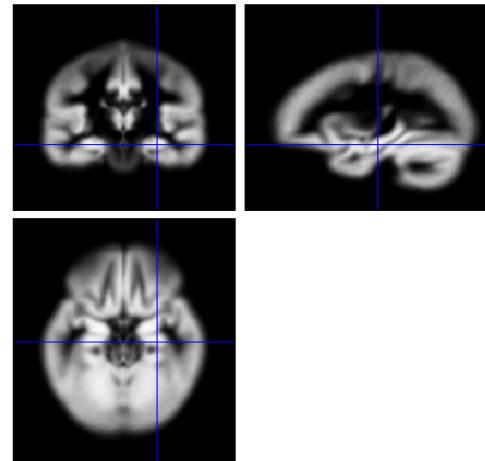
# Template

Iteratively generated  
from 471 subjects

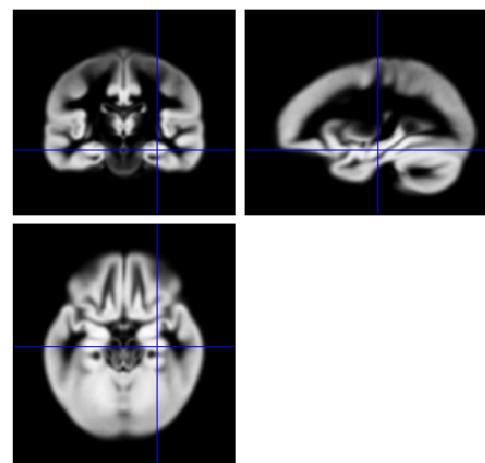
Began with rigidly  
aligned tissue  
probability maps

Used an inverse  
consistent  
formulation

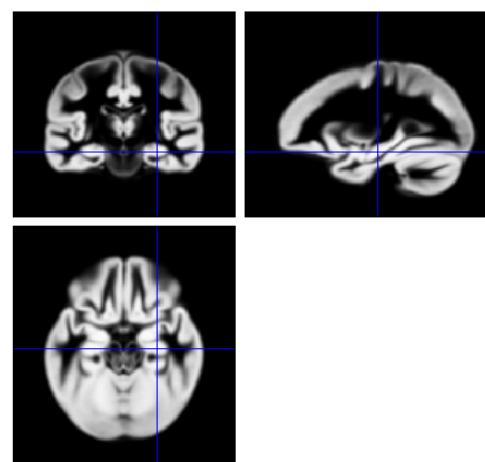
Initial  
Average

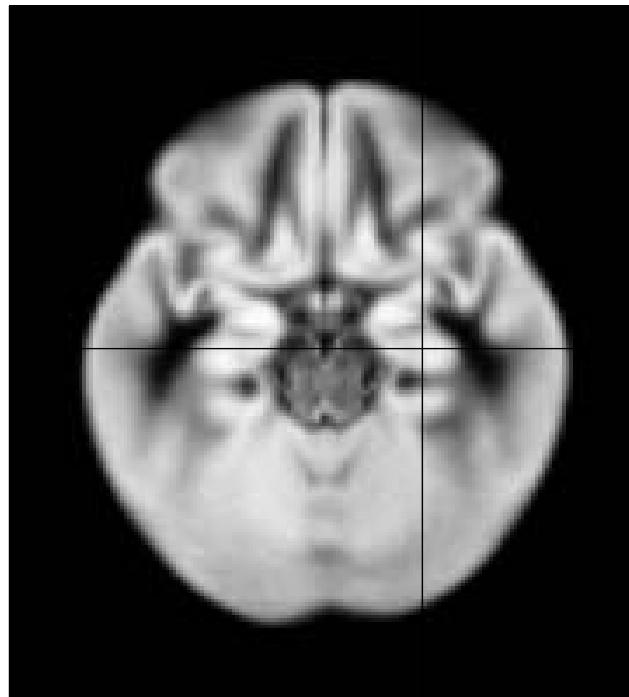
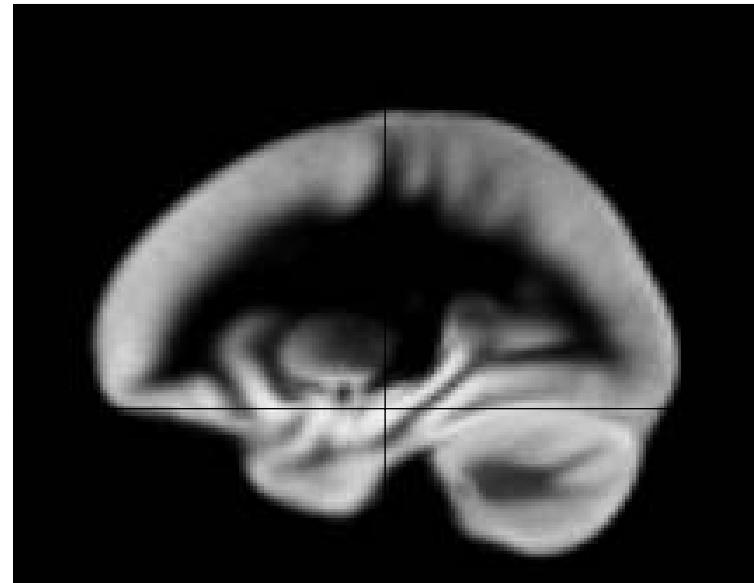
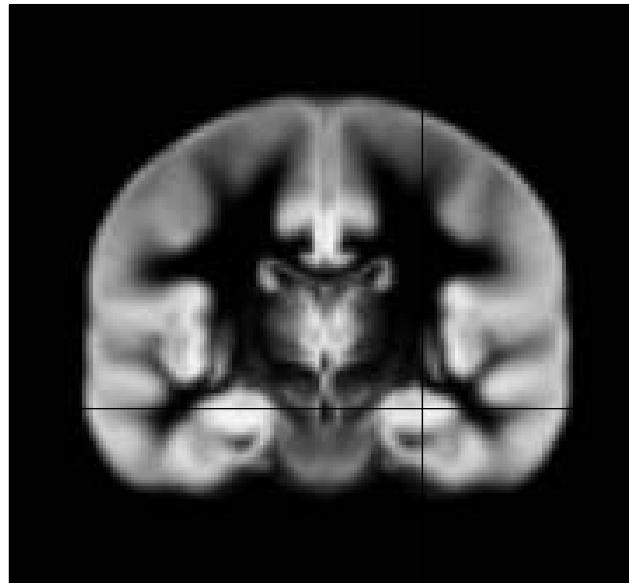


After a few  
iterations

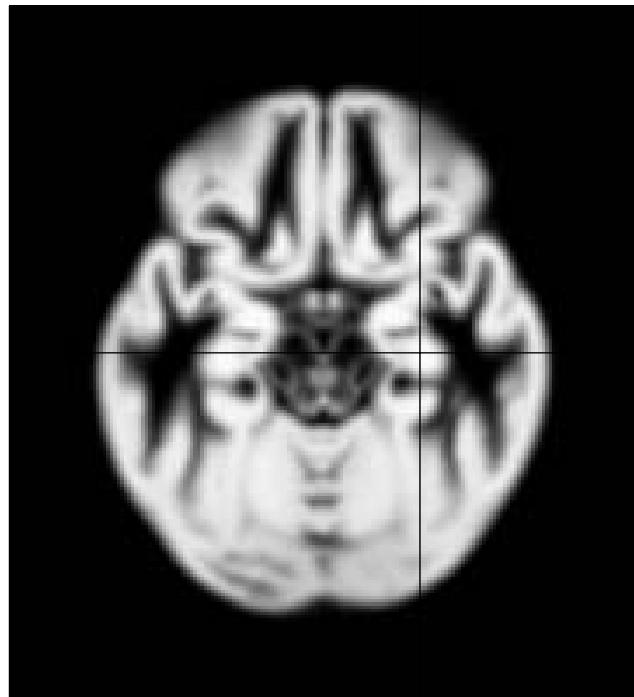
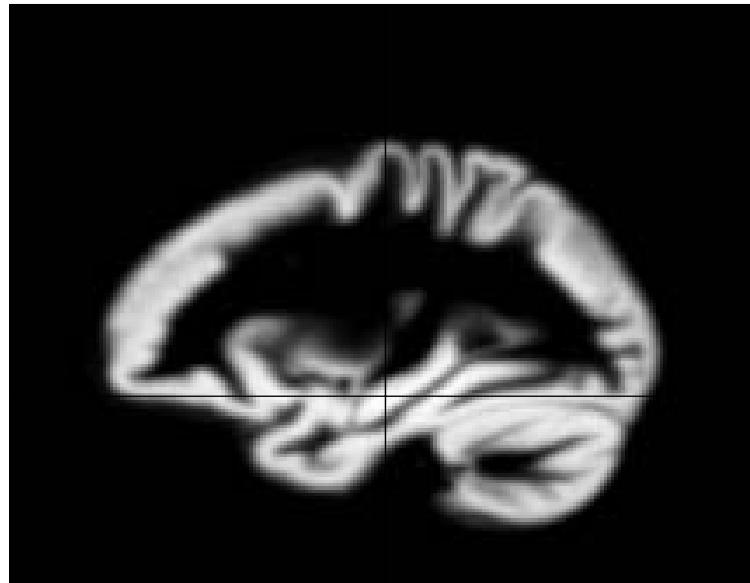
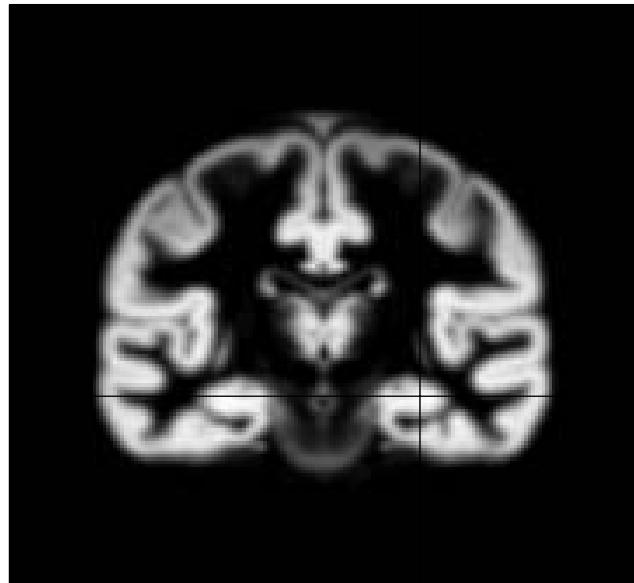


Final  
template



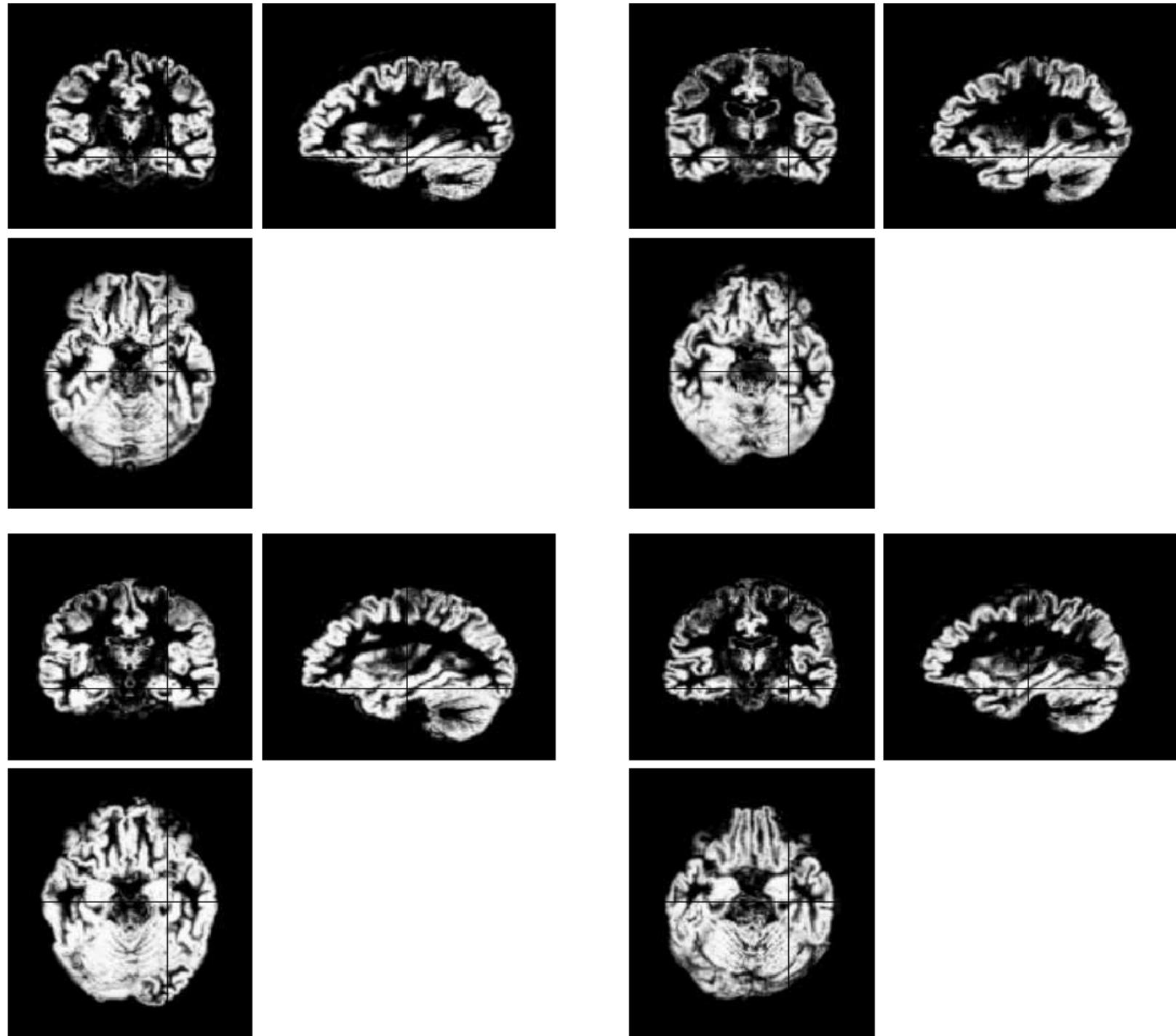


Grey matter  
average of 452  
subjects – affine

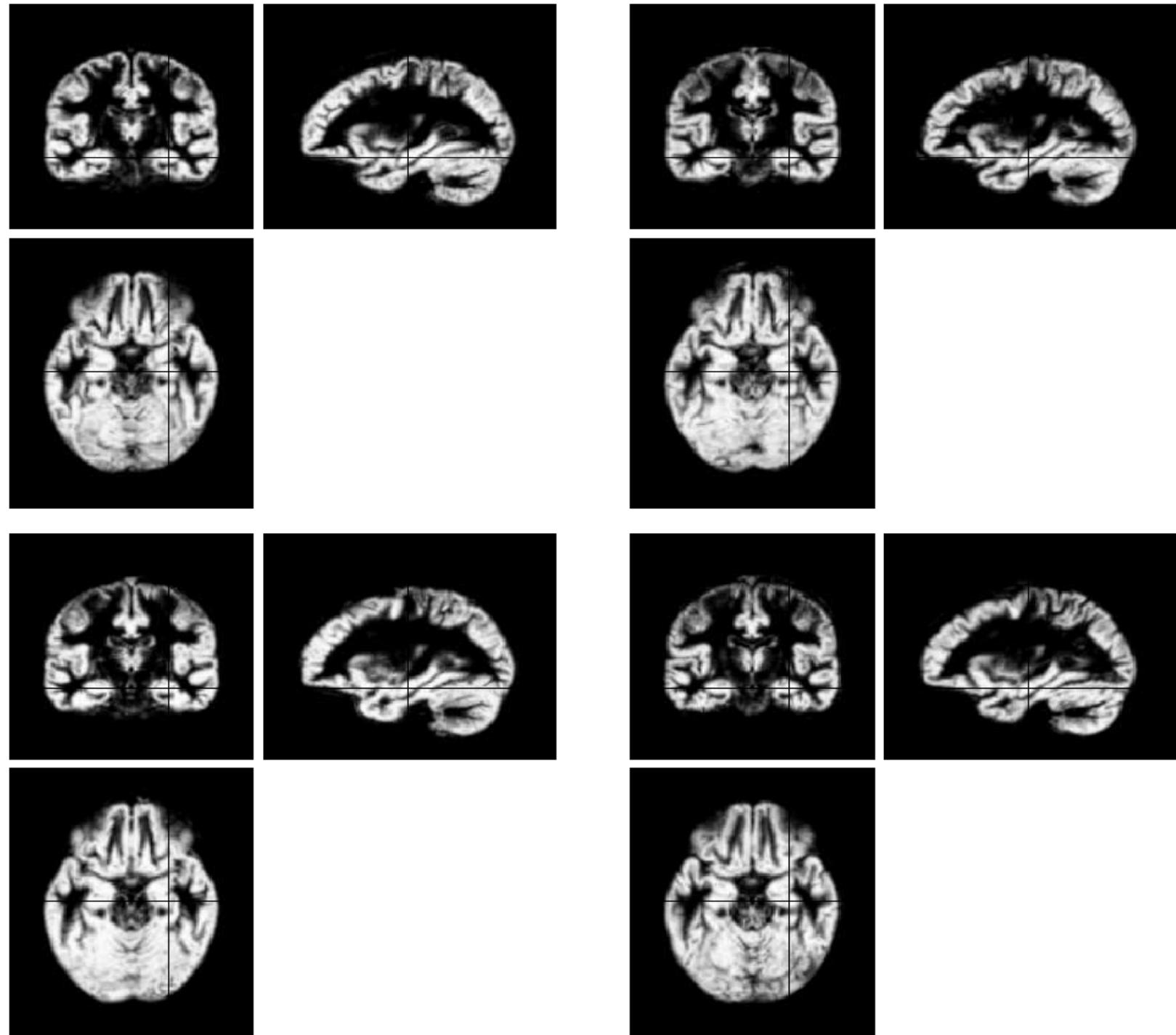


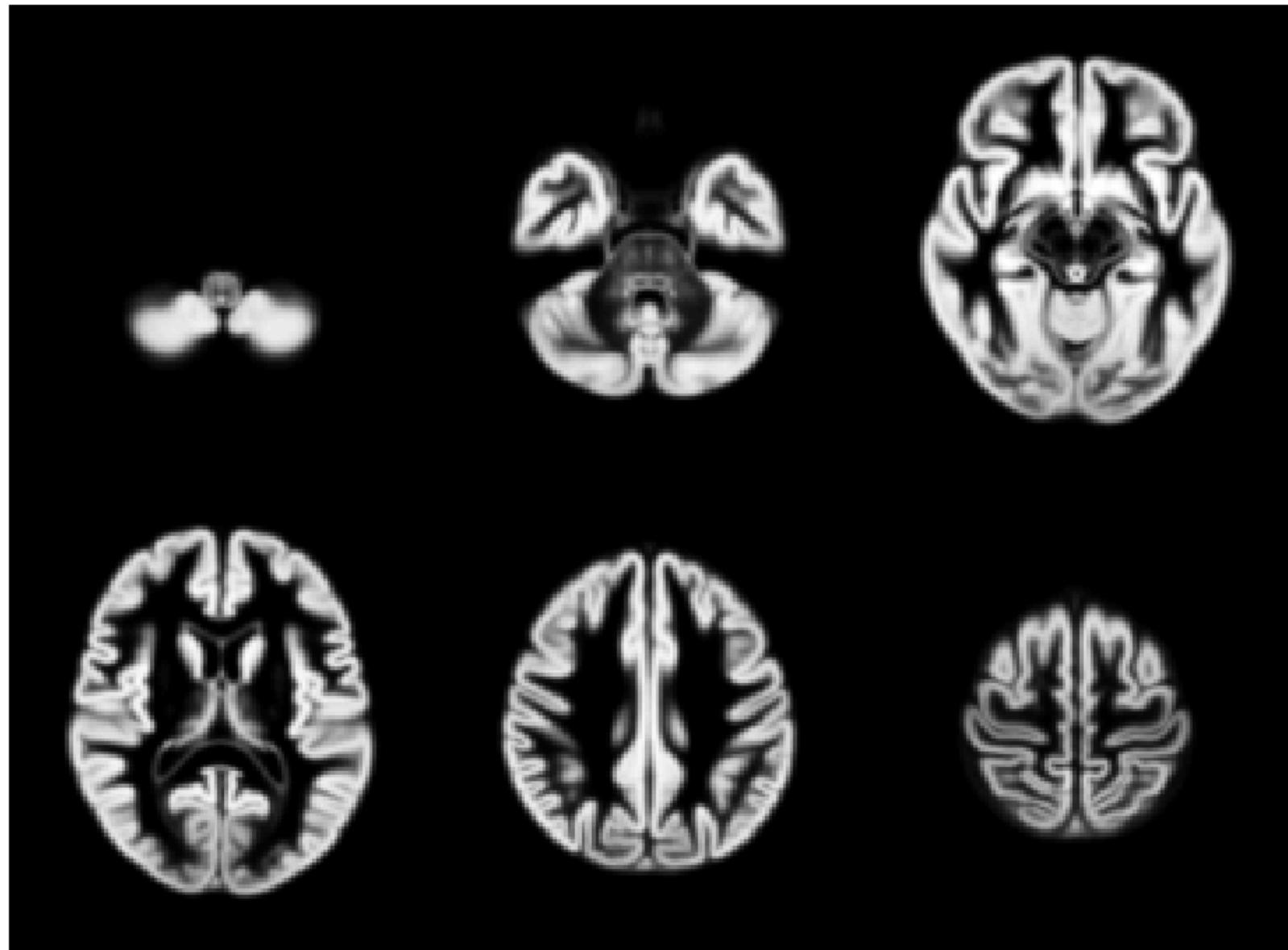
Grey matter  
average of 471  
subjects

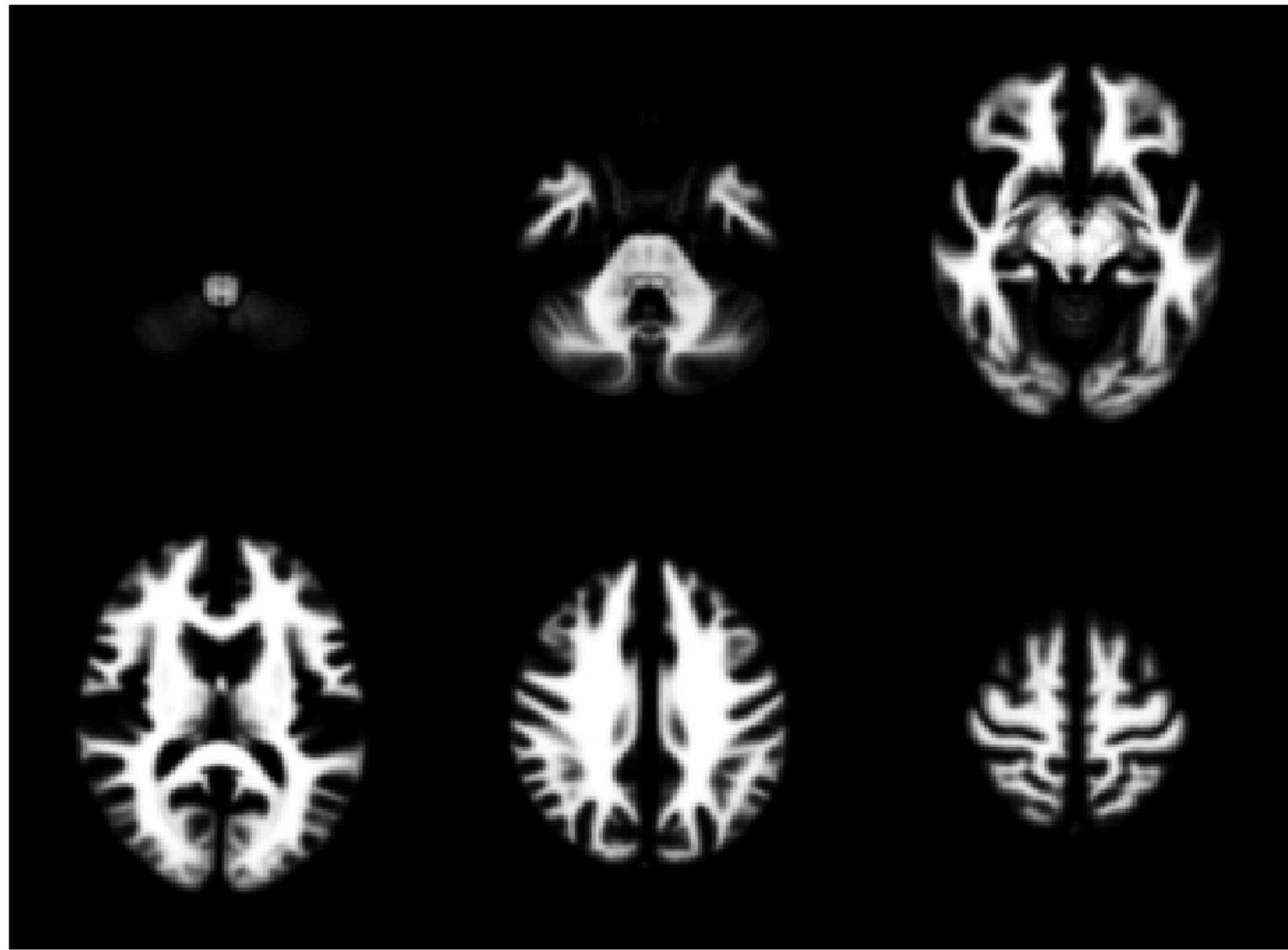
Initial  
GM images

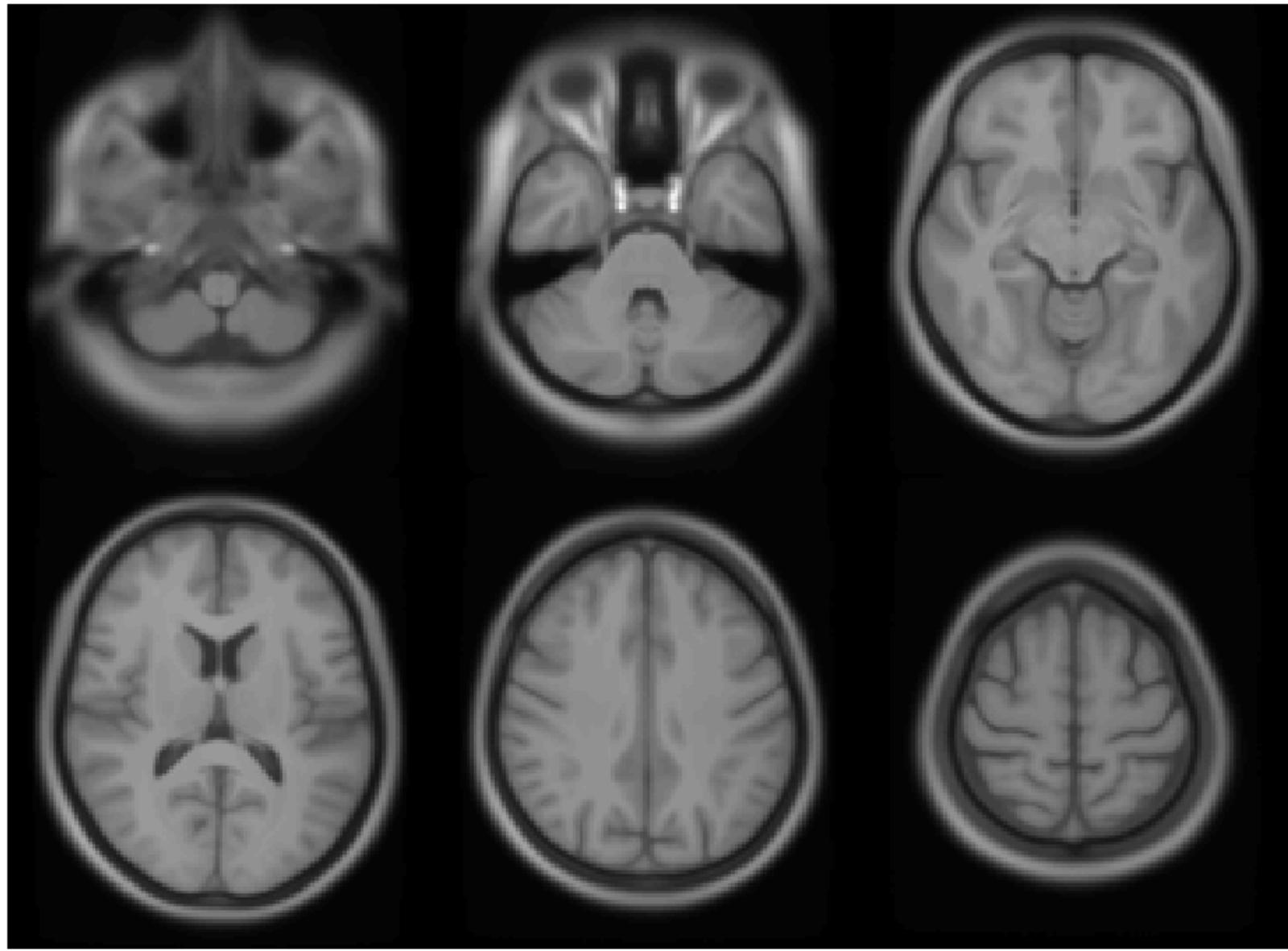


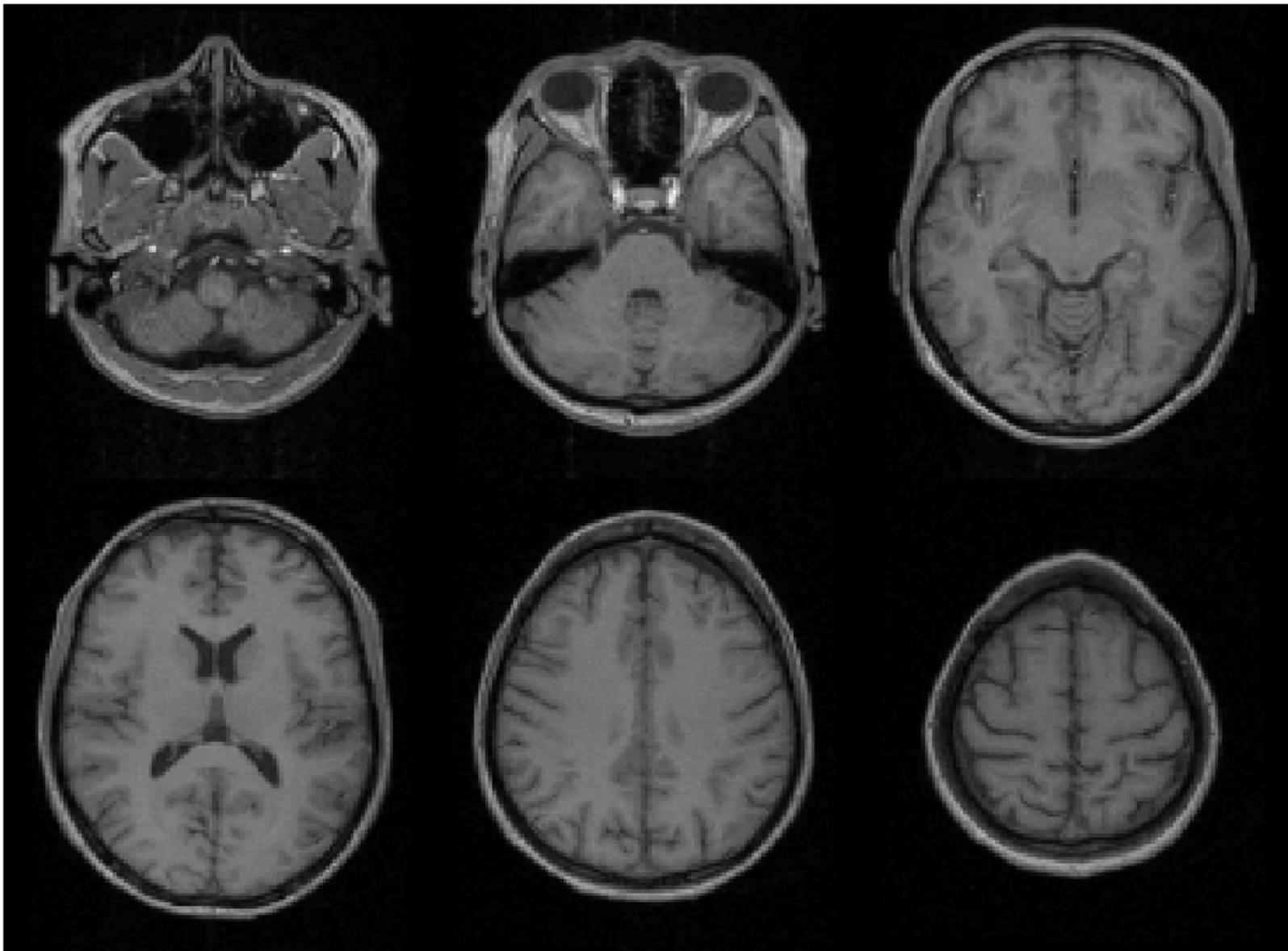
Warped  
GM images

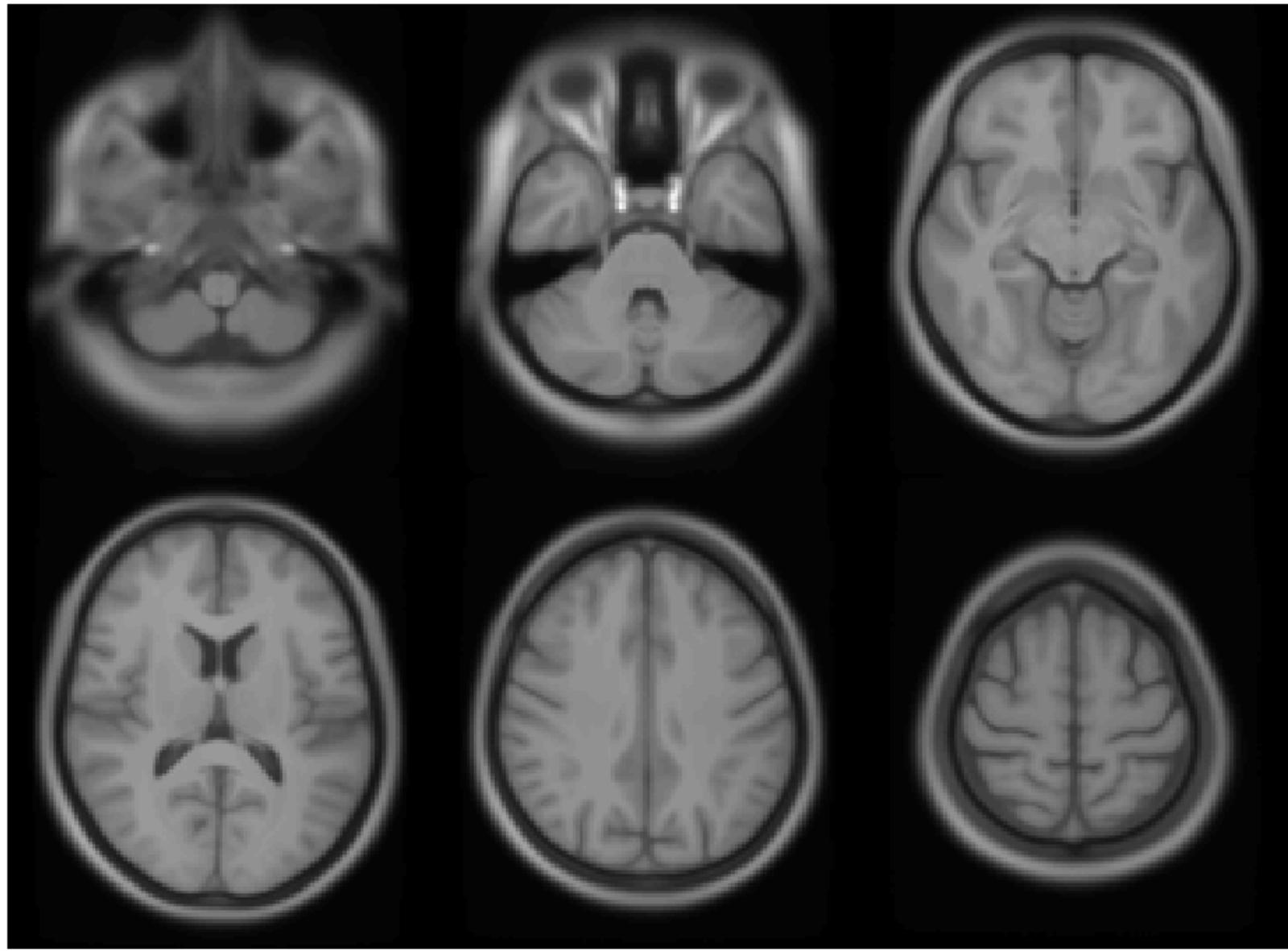


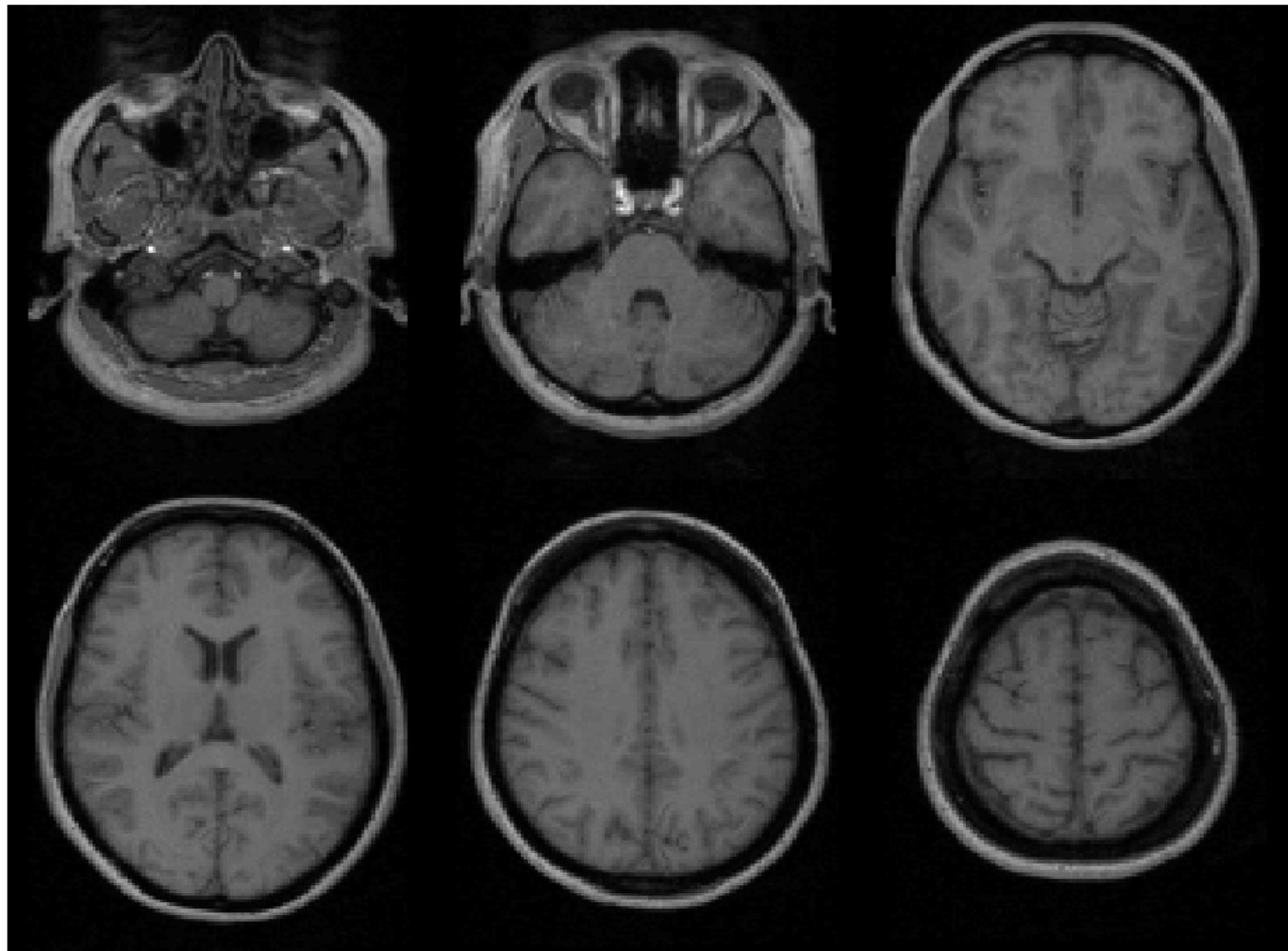


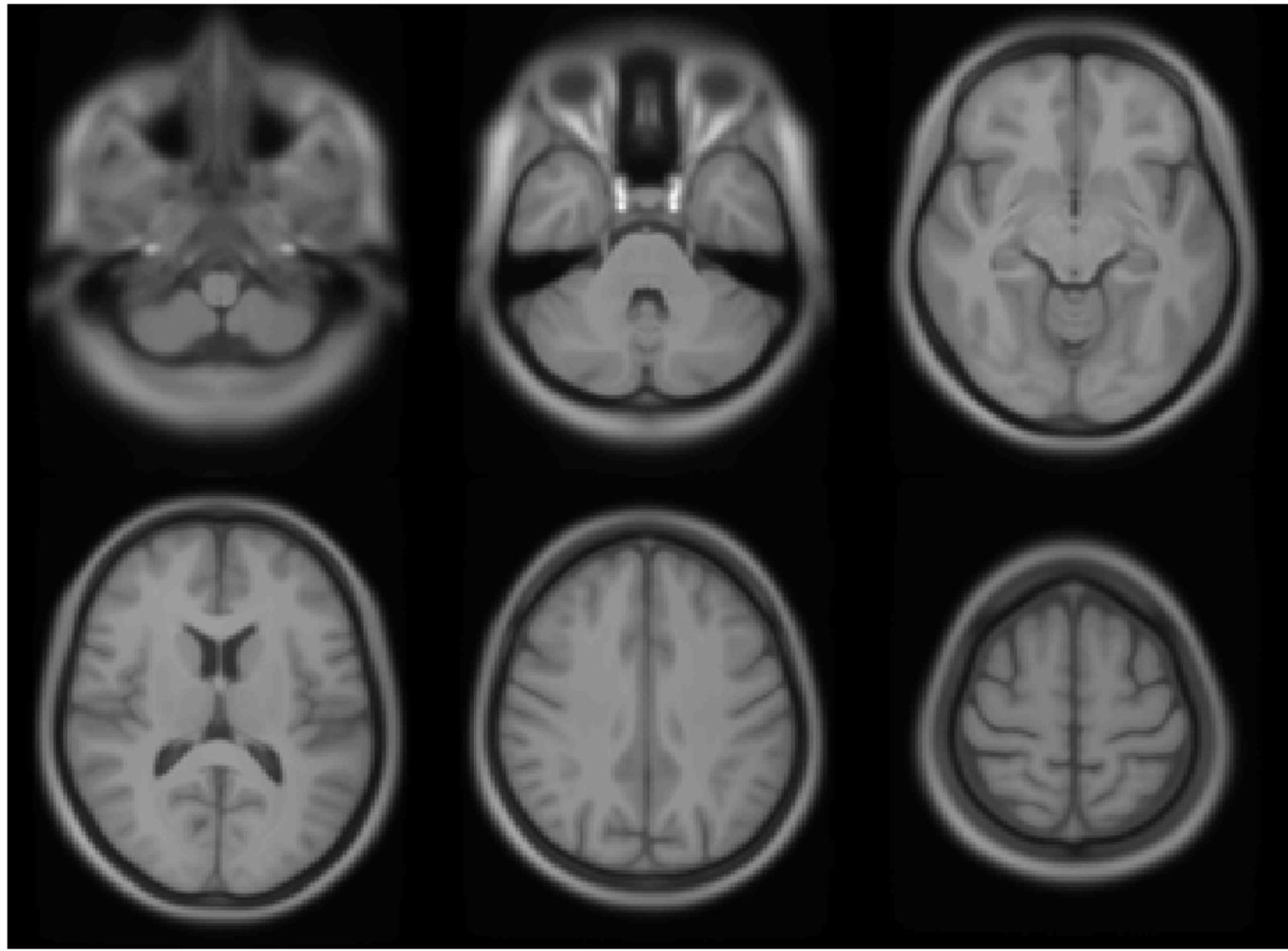


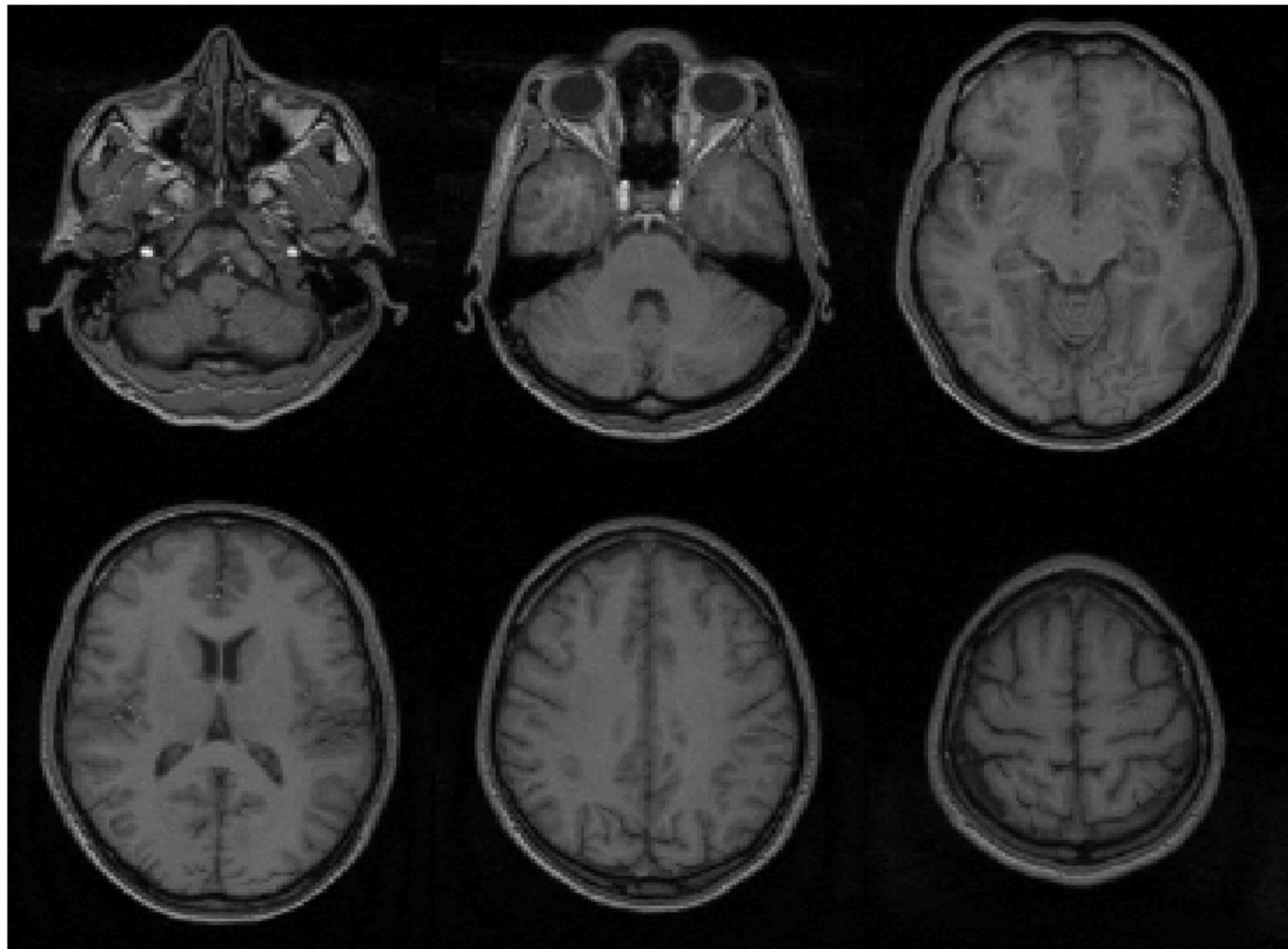


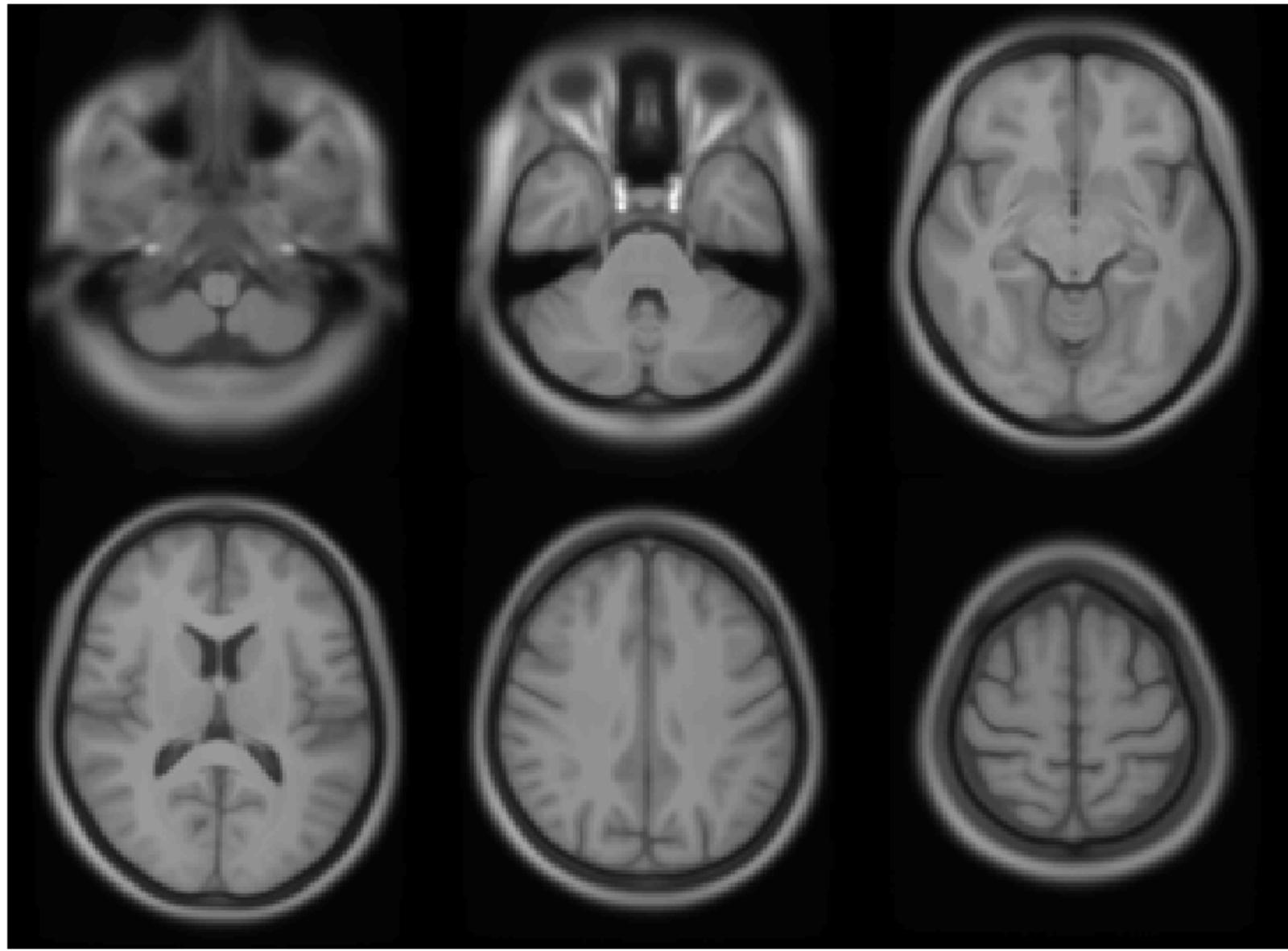












# Contents

- \* Segment
- \* Dartel
- \* **Smoothing**
  - \* **Compensating for inaccuracies in inter-subject alignment**

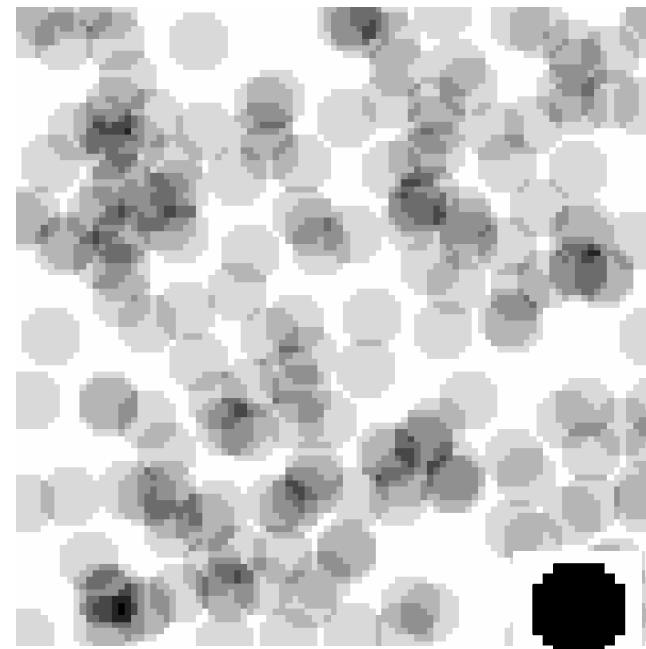
# Smooth

Blurring is done by convolution.

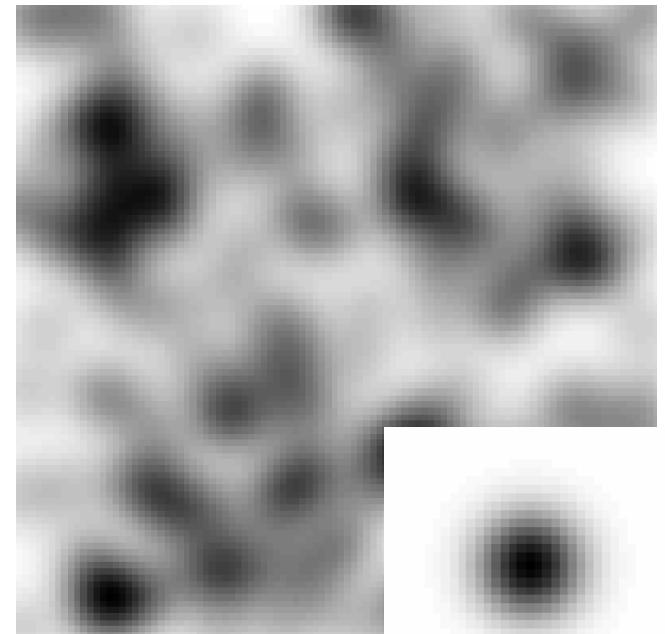
Each voxel after smoothing effectively becomes the result of applying a weighted region of interest (ROI).



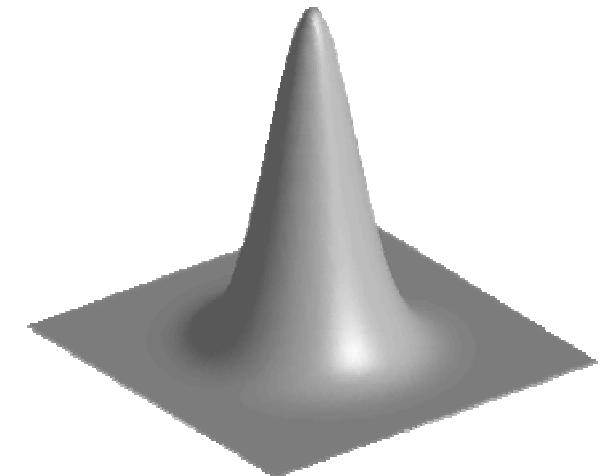
Before convolution



Convolved with a circle

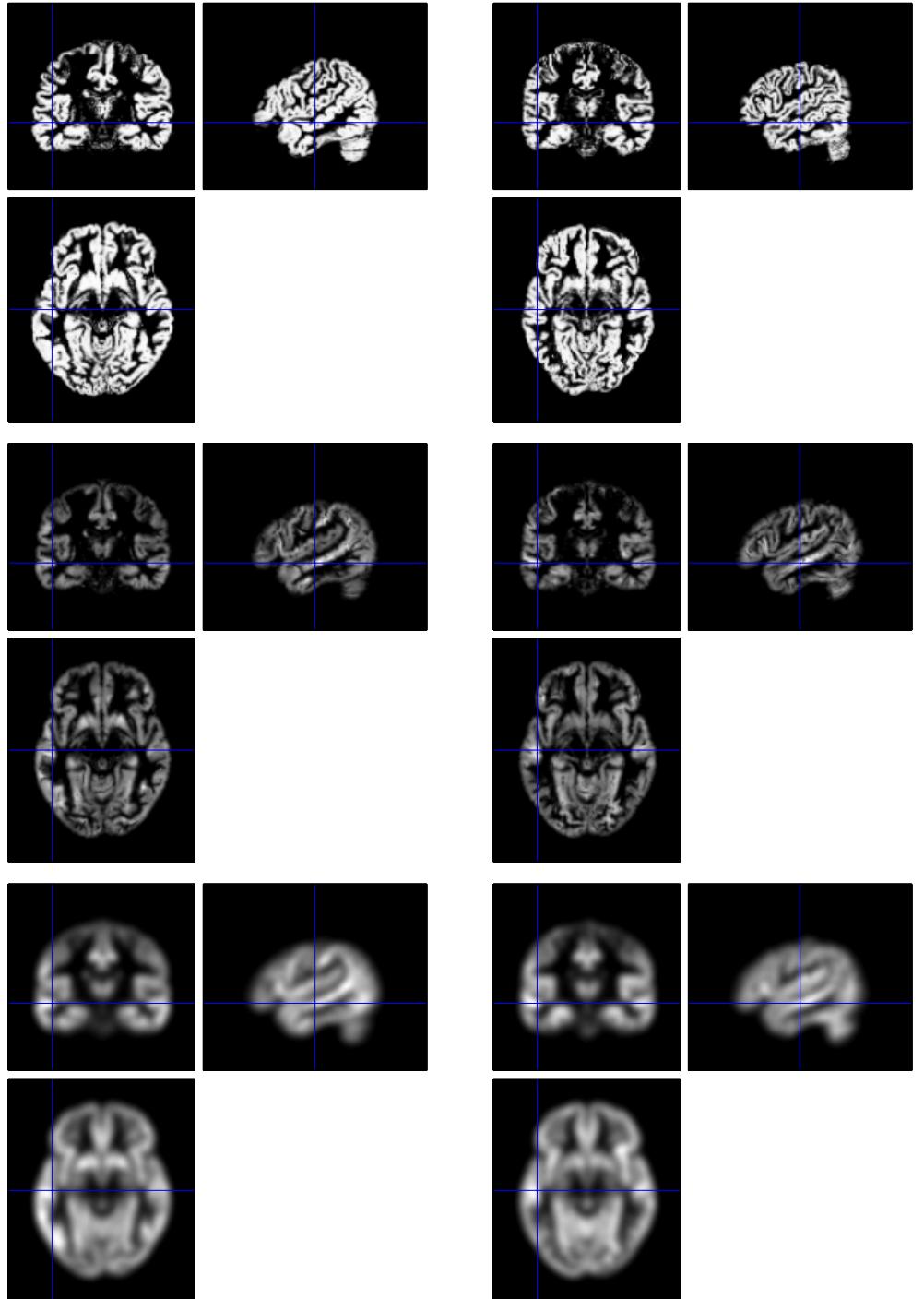


Convolved with a Gaussian



# VBM Pre-processing

- \* Use Segment button for characterising intensity distributions of tissue classes.
- \* DARTEL import, to generate rigidly aligned grey and white matter maps.
- \* DARTEL warping to generate “modulated” warped grey matter.
- \* Smooth.
- \* Statistics.



# References

- \* **Ashburner & Friston.** *Unified Segmentation.* NeuroImage 26:839-851 (2005).
- \* **Ashburner.** *A Fast Diffeomorphic Image Registration Algorithm.* NeuroImage 38:95-113 (2007).
- \* **Ashburner & Friston.** *Computing average shaped tissue probability templates.* NeuroImage 45(2): 333-341 (2009).
- \* **Klein et al.** *Evaluation of 14 nonlinear deformation algorithms applied to human brain MRI registration.* NeuroImage 46(3):786-802 (2009).