

fMRI Preprocessing

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Contents

- Registration basics
- Motion and realignment
- Inter-modal coregistration
- Spatial normalisation
- Gaussian smoothing

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Contents

Registration basics

- Voxel-to-world mapping
- Transformation
- Similarity measure
- Optimisation
- Interpolation
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- Inter-modal coregistration
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Representation of imaging data

- Three dimensional images are made up of voxels.
- Voxel intensities are stored on disk as lists of numbers.
- Meta-information about the data:
 - The image dimensions
 - Allowing conversion from list to 3D array
 - The "voxel-to-world mapping"

- Spatial transformation that maps from data coordinates (voxel column *i*, row *j*, slice *k*) into a real-world position (*x*,*y*,*z* mm) in a coordinate system e.g.:
 - Scanner coordinates
 - T&T/MNI coordinates

Image registration

Process of transforming different set of images into one coordinate system.

Two key ingredients:

> Transformation type:

- Rigid Within-subject
- Affine

Between-subject

Non-linear

> Similarity measure:

- Mean-squared difference
- Correlation coefficient
- Mutual information

Within-modality

Between-modality







Optimisation

- Automatic image registration is done by using an optimisation algorithm.
- Optimisation involves finding some "best" parameters according to an "objective function", which is either minimised or maximised.





Reslicing / Interpolation



 Applying the transformation parameters, and re-sampling the data onto the same grid of voxels as the target image
reslicing, interpolation, regridding, transformation, and writing



Simple Interpolation

Nearest neighbour

 Take the value of the closest voxel

🖵 Tri-linear

- Just a weighted average of the neighbouring voxels
- $f_5 = f_1 x_2 + f_2 x_1$
- $f_6 = f_3 x_2 + f_4 x_1$
- $f_7 = f_5 y_2 + f_6 y_1$





B-spline Interpolation

A continuous function is represented by a linear combination of basis functions



2D B-spline basis functions of degrees 0, 1, 2 and 3





Nearest neighbour and trilinear interpolation are the same as B-spline interpolation with degrees 0 and 1.



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Motion correction

- Head movement is a very large source of variance in fMRI data.
- Motion correction: realign a time-series of images acquired from the same subject.
 - Within-subject transformation: rigid-body (6 parameters)
 - Within-modality: least squares objective function













Residual Errors from aligned fMRI

- Slices are not acquired simultaneously
 - rapid movements not accounted for by rigid body model
- Resampling can introduce interpolation errors
 - especially tri-linear interpolation
- Image artefacts may not move according to a rigid body model
 - image distortion
 - image dropout
- Functions of the estimated motion parameters can be modelled as confounds in subsequent analyses

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Movement by Distortion Interaction of fMRI

- Subject disrupts B₀ field, rendering it inhomogeneous
 - Distortions in phase-encode direction
- Subject moves during EPI time series
 - Distortions vary with subject orientation
 - Shape varies (non-rigidly)
- "Realign & Unwarp": generative model that combines a model of geometric distortions and a model of subject motion to correct images.









After rotation



After rotation



Correction by Unwarp

Movement correction strategies

No correction









Correction by covariation



















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Coregistration (intra-subject, inter-modal)

- Inter-modal registration.
- Match images from same subject but different modalities:
 - anatomical localisation of single subject activations
 - achieve more precise spatial normalisation of functional image using anatomical image.





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Coregistration maximises Mutual Information

Between-modality registration:

Seek to measure shared information in some sense.



Joint histogram sharpness correlates with image alignment. **Mutual information** and related measures attempt to quantify how well one image predicts the other.

Coregistration



/coreg_demo/T2w.nii ./coreg_demo/T1w.nii 1.12 Normalised mutual information 1.1 1.08 1.06 1.04 1.02 1 -50 -40 -30 -20 -10 0 10 20 30 40 50 L/R translation (mm)

Final Joint Histogram

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 - Unified segmentation
- Gaussian smoothing



Spatial Normalisation





Spatial Normalisation - Reasons

Inter-subject averaging

- Increase sensitivity with more subjects
 - Fixed-effects analysis
- Extrapolate findings to the population as a whole
 - Random-effects analysis

Make results from different studies comparable by aligning them to standard space.



Standard spaces

Talairach Atlas

MNI/ICBM AVG152 Template



The MNI template follows the *convention* of T&T, but doesn't match the *particular brain* (<u>http://imaging.mrc-cbu.cam.ac.uk/imaging/MniTalairach</u>)

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Unified Segmentation

- Normalising segmented tissue maps should be more robust and precise than using the original images.
- Tissue segmentation benefits from spatially-aligned prior tissue probability maps.
- Combining normalisation and segmentation in a unified model:
 - Gaussian mixture model segmentation
 - Intensity inhomogeneity (bias field) correction
 - Warping (non-linear registration)



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Tissue intensity distributions (T1-weighted MRI)





Mixture of Gaussians

Classification is based on a Mixture of Gaussians model, which represents the intensity probability density by a number of Gaussian distributions.



Image Intensity ——



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 inhomogeneity

A multiplicative bias field is modelled as a spatially smooth image (a linear combination of basis functions).



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Corrupted image

Bias Field

Corrected image

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Tissue Probability Maps

- Each TPM indicates the prior probability for a particular tissue at each point in MNI space.
- SPM12's TPMs are derived from the IXI dataset initialised with the ICBM 452 atlas and other data.



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Deforming the Tissue Probability Maps

- Tissue probability images are warped to match the subject.
- The inverse transform warps to the TPMs.
- Warps are constrained to be reasonable by penalising various distortions (regularisation).





Spatial Normalisation – Overfitting





Spatial Normalisation – Results



Affine registration

Non-linear registration

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Spatial Normalisation – Limitations

Seek to match functionally homologous regions, but...

- No exact match between structure and function
- Different cortices can have different folding patterns
- Challenging high-dimensional optimisation, many local optima

Compromise

- Correct relatively large-scale variability (sizes of structures)
- Smooth over finer-scale residual differences



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Smoothing

- Why would we deliberately blur the data?
 - Improves spatial overlap by blurring over minor anatomical differences and registration errors
 - Averaging neighbouring voxels suppresses noise
 - Increases sensitivity to effects of similar scale to kernel (matched filter theorem)
 - Makes data more normally distributed (central limit theorem)
 - Reduces the effective number of multiple comparisons

□ How is it implemented?

 Convolution with a 3D Gaussian kernel, of specified full-width at halfmaximum (FWHM) in mm



Effect of smoothing



3D Gaussian smoothing with FWHM: 0, 2, 4, 6, 8, 10, 12, 14, 16 mm isotropic

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