

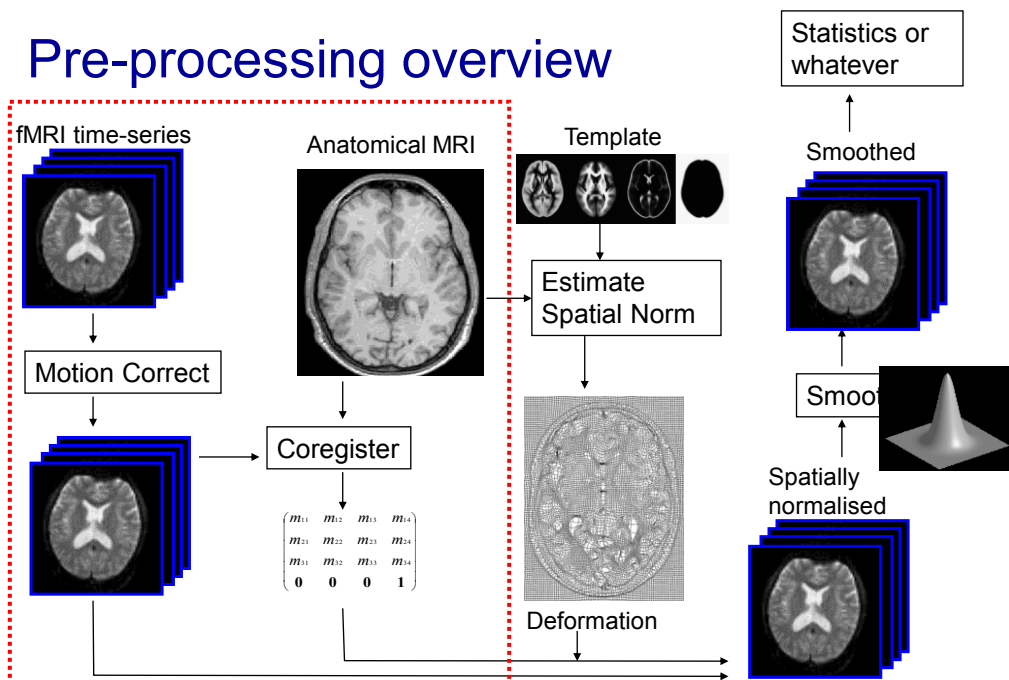


# Preprocessing I: Within Subject

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## Pre-processing overview



# Contents

## \* Preliminaries

- \* Rigid-body and affine transformations
- \* Optimisation and objective functions
- \* Transformations and interpolation
- \* Realignment
- \* Coregistration

## Rigid-body transformations

- \* Assume that brain of the same subject doesn't change shape or size in the scanner.
  - \* Head can move, but remains the same shape and size.
  - \* Some exceptions:
    - \* Image distortions.
    - \* Brain slops about slightly because of gravity.
    - \* Brain growth or atrophy over time.
- \* If the subject's head moves, we need to correct the images.
  - \* Do this by image registration.

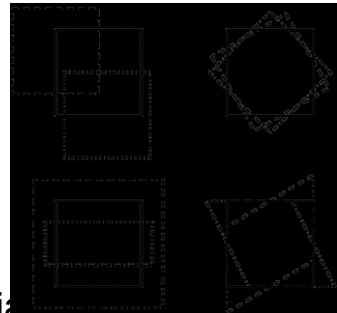
# Image registration

## Two components:

- **Registration** - i.e. Optimise the parameters that describe a spatial transformation between the source and reference images
- **Transformation** - i.e. Re-sample according to the determined transformation parameters

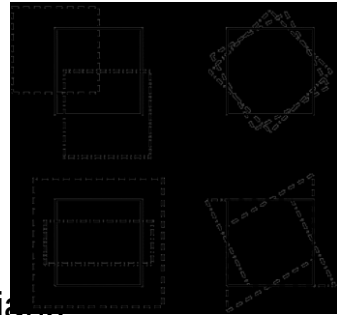
## 2D affine transforms

- \* Translations by  $t_x$  and  $t_y$ 
  - \*  $x_1 = x_0 + t_x$
  - \*  $y_1 = y_0 + t_y$
- \* Rotation around the origin by  $\Theta$  radians
  - \*  $x_1 = \cos(\Theta) x_0 + \sin(\Theta) y_0$
  - \*  $y_1 = -\sin(\Theta) x_0 + \cos(\Theta) y_0$
- \* Zooms by  $s_x$  and  $s_y$ 
  - \*  $x_1 = s_x x_0$
  - \*  $y_1 = s_y y_0$
- \* Shear
  - \*  $x_1 = x_0 + h y_0$
  - \*  $y_1 = y_0$



## 2D affine transforms

- \* Translations by  $t_x$  and  $t_y$ 
  - \*  $x_1 = 1 x_0 + 0 y_0 + t_x$
  - \*  $y_1 = 0 x_0 + 1 y_0 + t_y$
- \* Rotation around the origin by  $\Theta$  radians



- \*  $x_1 = \cos(\Theta) x_0 + \sin(\Theta) y_0 + 0$
- \*  $y_1 = -\sin(\Theta) x_0 + \cos(\Theta) y_0 + 0$
- \* Zooms by  $s_x$  and  $s_y$ :
  - \*  $x_1 = s_x x_0 + 0 y_0 + 0$
  - \*  $y_1 = 0 x_0 + s_y y_0 + 0$

### \* Shear

- \*  $x_1 = 1 x_0 + h y_0 + 0$
- \*  $y_1 = 0 x_0 + 1 y_0 + 0$

## 3D rigid-body transformations

- \* A 3D rigid body transform is defined by:
  - \* 3 translations - in X, Y & Z directions
  - \* 3 rotations - about X, Y & Z axes
- \* The order of the operations matters

$$\begin{pmatrix} 1 & 0 & 0 & X_{trans} \\ 0 & 1 & 0 & Y_{trans} \\ 0 & 0 & 1 & Z_{trans} \\ 0 & 0 & 0 & 1 \end{pmatrix} \times \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & \cos\Phi & \sin\Phi & 0 \\ 0 & -\sin\Phi & \cos\Phi & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \times \begin{pmatrix} \cos\Theta & 0 & \sin\Theta & 0 \\ 0 & 1 & 0 & 0 \\ -\sin\Theta & 0 & \cos\Theta & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \times \begin{pmatrix} \cos\Omega & \sin\Omega & 0 & 0 \\ -\sin\Omega & \cos\Omega & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

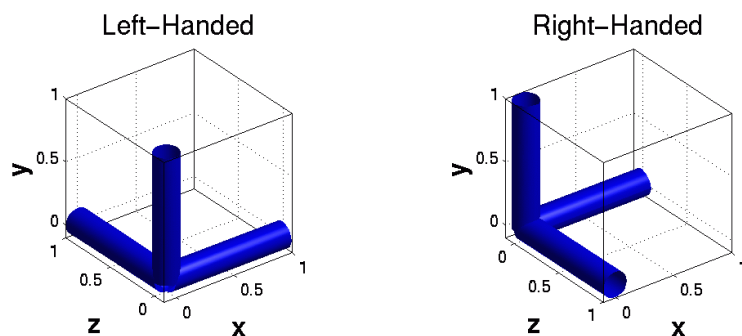
Translations
Pitch about x axis
Roll about y axis
Yaw about z axis

## Voxel-to-world transforms

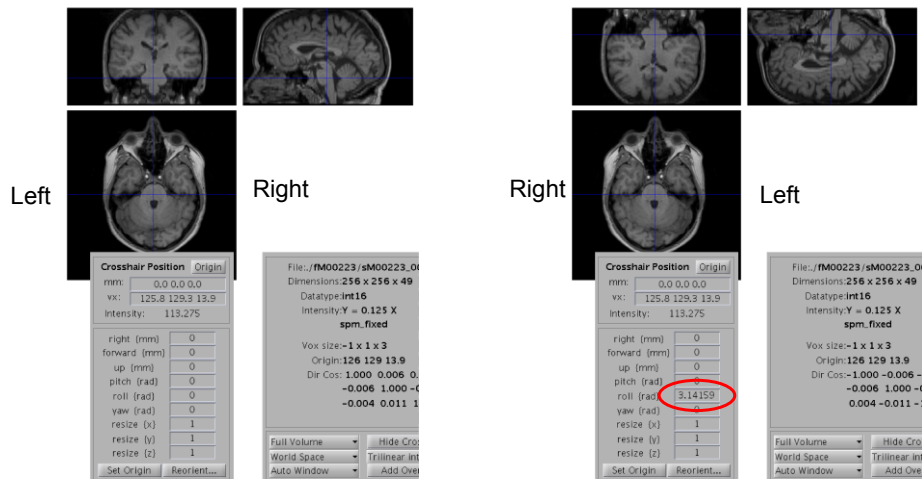
- \* Affine transform associated with each image
  - \* Maps from voxels ( $x=1..n_x$ ,  $y=1..n_y$ ,  $z=1..n_z$ ) to some world co-ordinate system. e.g.,
    - \* Scanner co-ordinates - images from DICOM toolbox
    - \* T&T/MNI coordinates - spatially normalised
- \* Registering image B (source) to image A (target) will update B's voxel-to-world mapping
  - \* Mapping from voxels in A to voxels in B is by
    - \* A-to-world using  $M_A$ , then world-to-B using  $M_B^{-1}$
    - \*  $M_B^{-1} M_A$

## Left- and right-handed coordinate systems

- \* NIfTI format files are stored in either a left- or right-handed system
  - \* Indicated in the header
- \* Talairach & Tournoux uses a right-handed system
- \* Mapping between them sometimes requires a flip
  - \* Affine transform has a negative determinant

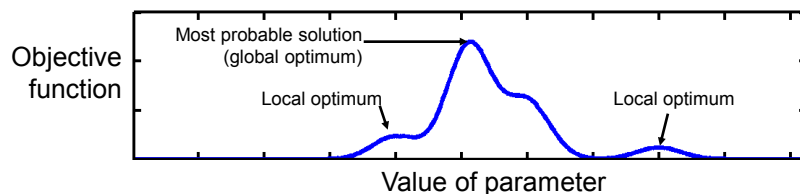


## “Radiological” and “neurological” conventions



## Optimisation

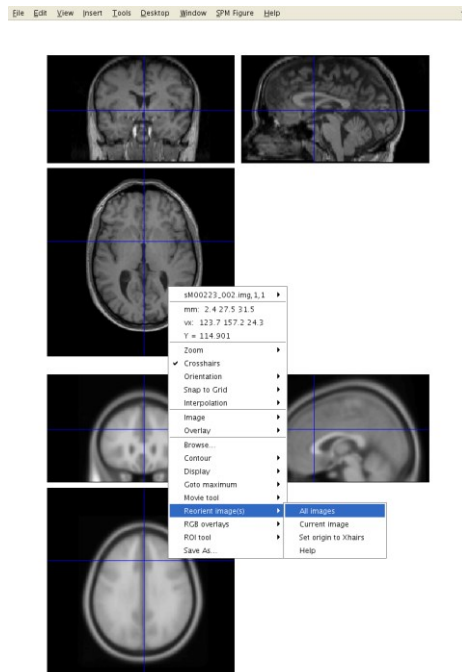
- \* Image registration is done by optimisation.
- \* Optimisation involves finding some “best” parameters according to an “objective function”, which is either minimised or maximised
- \* The “objective function” is often related to a probability based on some model



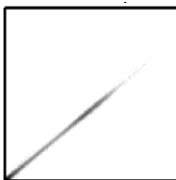
## Optimisation

- \* Because registration only finds a *local optimum*, some manual reorienting of the images may be needed before doing anything else in SPM.

An MNI-space image from spm12/canonical directory.

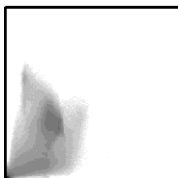


## Objective functions



### \* Intra-modal

- \* Mean squared difference (minimise)
- \* Normalised cross correlation (maximise)

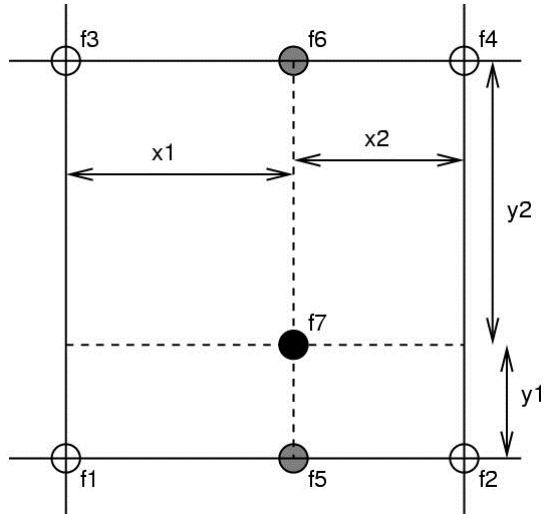


### \* Inter-modal (or intra-modal)

- \* Mutual information (maximise)
- \* Normalised mutual information (maximise)
- \* Entropy correlation coefficient (maximise)

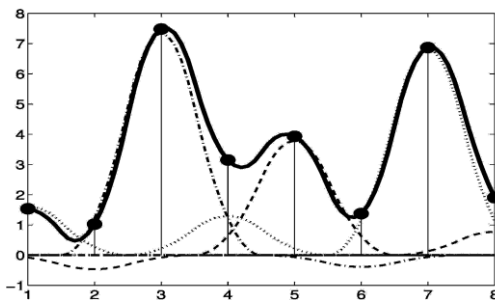
## Simple interpolation

- \* Nearest neighbour
  - \* Take the value of the closest voxel
- \* Tri-linear
  - \* Just a weighted average of the neighbouring voxels
  - \*  $f_5 = f_1 x_2 + f_2 x_1$
  - \*  $f_6 = f_3 x_2 + f_4 x_1$
  - \*  $f_7 = f_5 y_2 + f_6 y_1$

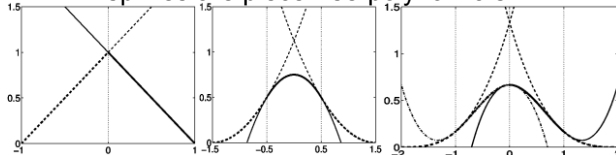


## B-spline interpolation

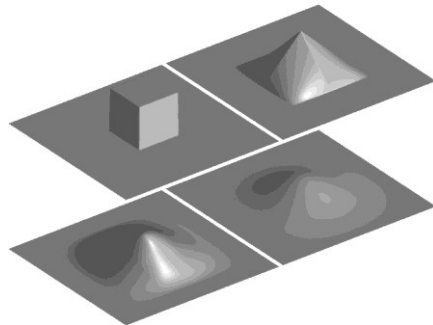
A continuous function is represented by a linear combination of basis functions



B-splines are piecewise polynomials



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.



# Contents

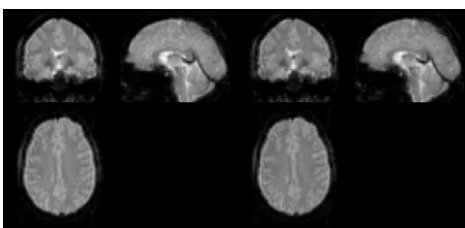
- \* Preliminaries

- \* **Realignment**

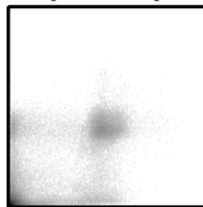
- \* **Realignment by minimising mean-squared difference**
  - \* **Residual artifacts and distortion correction**

- \* Coregistration

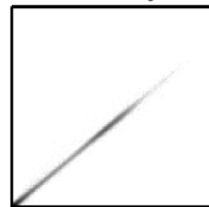
## Mean-squared difference



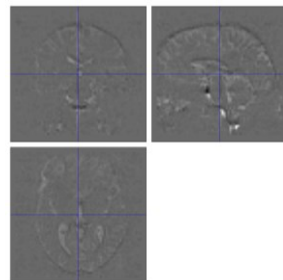
Original Joint Histogram



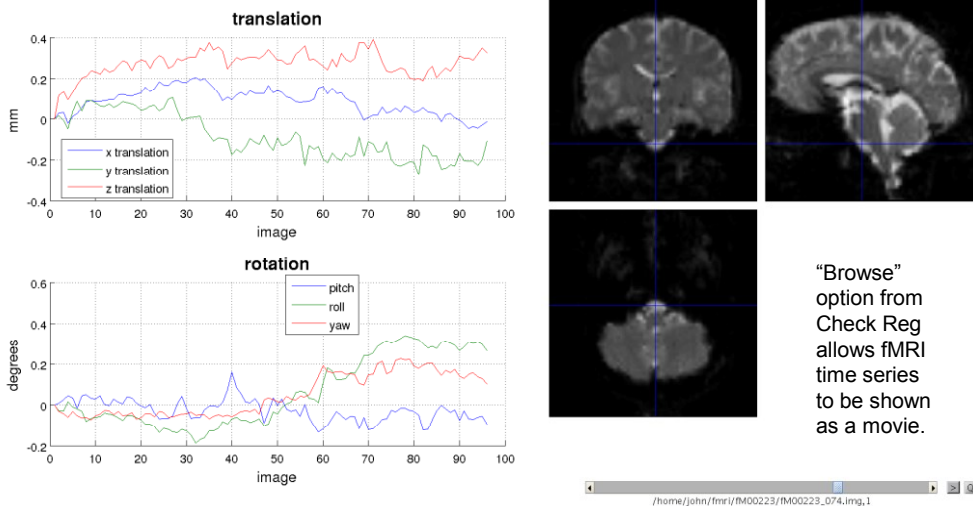
Final Joint Histogram



- \* Minimising mean-squared difference works for intra-modal registration (realignment)
- \* Simple relationship between intensities in one image, versus those in the other
  - \* Assumes normally distributed differences



## Motion estimates

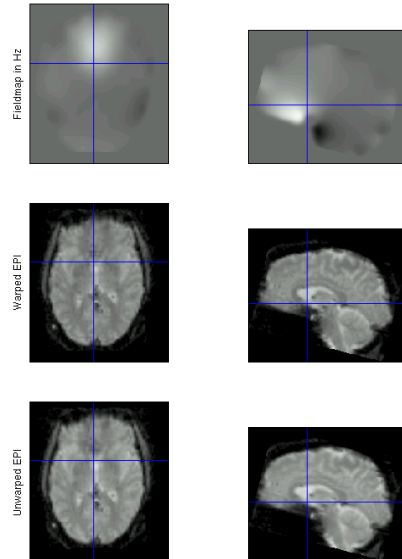


## Residual errors from aligned fMRI

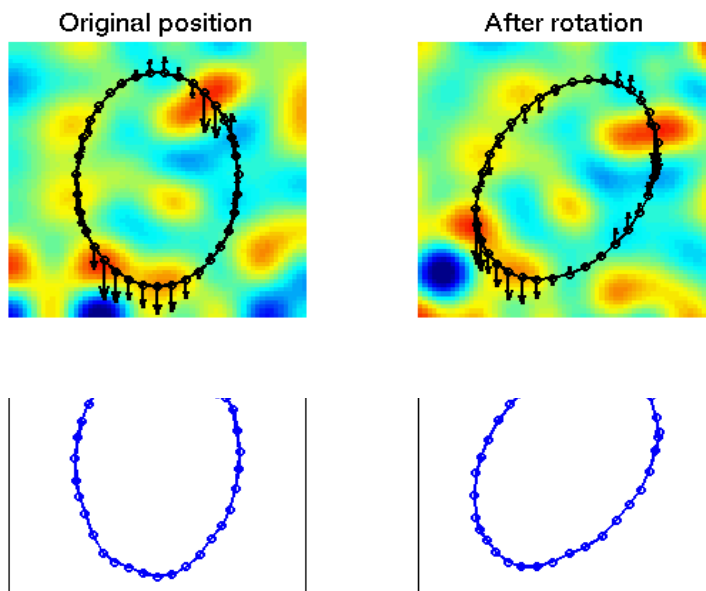
- \* Re-sampling can introduce interpolation errors
  - \* especially tri-linear interpolation
- \* Gaps between slices can cause aliasing artefacts
- \* Slices are not acquired simultaneously
  - \* rapid movements not accounted for by rigid body model
- \* Image artefacts may not move according to a rigid body model
  - \* image distortion
  - \* image dropout
  - \* Nyquist ghost
- \* BOLD signal changes influence the estimated motion.
- \* Functions of the estimated motion parameters can be modelled as confounds in subsequent analyses

# Movement-by-distortion interaction of fMRI

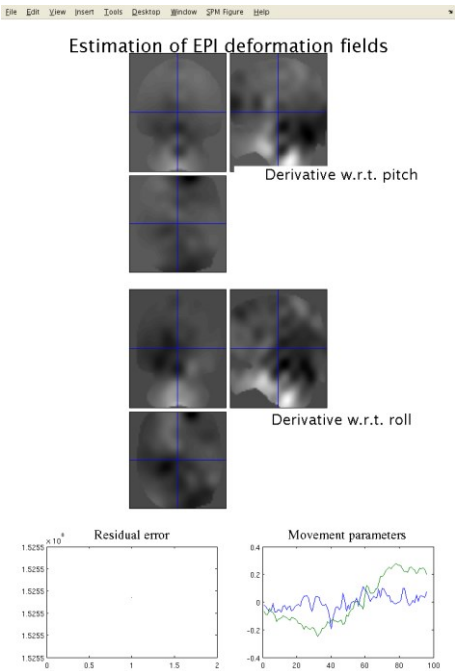
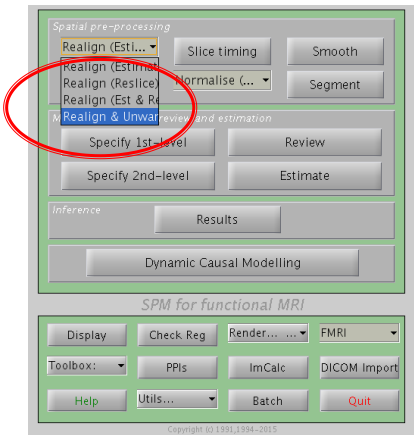
- \* Subject disrupts  $B_0$  field, rendering it inhomogeneous
  - \* distortions in phase-encode direction
- \* Subject moves during EPI time series
- \* Distortions vary with subject orientation
  - \* shape varies



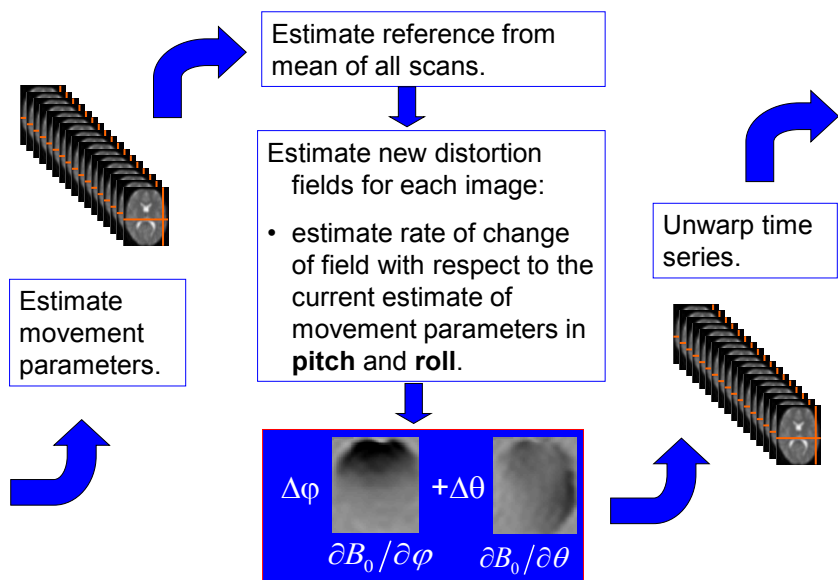
## Movement-by-distortion interaction



# Realign & Unwarp



# Correcting for distortion changes



Andersson et al, 2001

# Contents

- \* Preliminaries

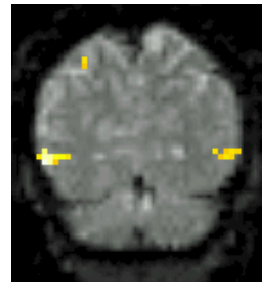
- \* Realignment

- \* **Coregistration**

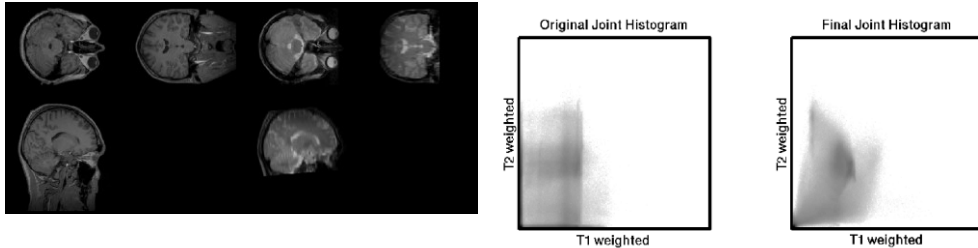
- \* Coregistration by maximising mutual information

## Coregistration

- 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.

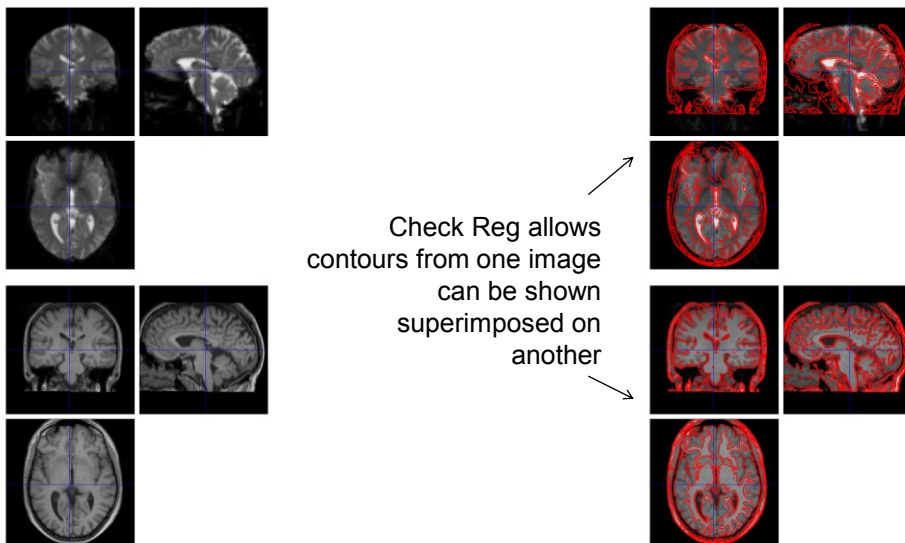


## Coregistration maximises Mutual Information

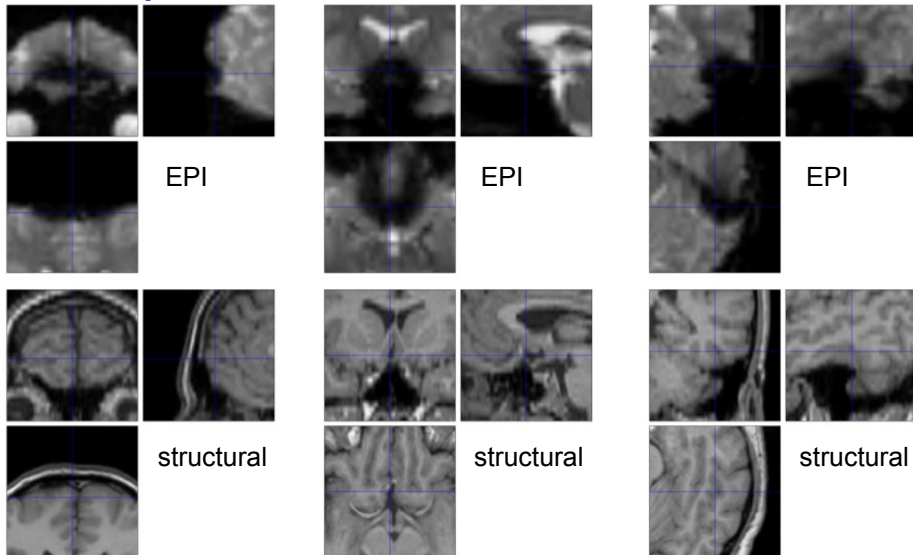


- \* Used for between-modality registration
- \* Derived from joint histograms
- \*  $MI = \int_{ab} P(a,b) \log_2 [P(a,b) / (P(a) P(b))]$ 
  - \* Related to entropy:  $MI = -H(a,b) + H(a) + H(b)$ 
    - \* Where  $H(a) = -\int_a P(a) \log_2 P(a)$  and  $H(a,b) = -\int_a P(a,b) \log_2 P(a,b)$

## “Check Reg” to assess alignment



## EPI dropout and distortion



## References

- \* [Friston et al.](#) *Spatial registration and normalisation of images.* Human Brain Mapping 3:165-189 (1995).
- \* [Collignon et al.](#) *Automated multi-modality image registration based on information theory.* IPMI'95 pp 263-274 (1995).
- \* [Thévenaz et al.](#) *Interpolation revisited.* IEEE Trans. Med. Imaging 19:739-758 (2000).
- \* [Andersson et al.](#) *Modeling geometric deformations in EPI time series.* Neuroimage 13:903-919 (2001).