

Image registration

Quantitative and Functional Imaging

BME 4420/7450

Fall 2022

Topics

- What is image registration?
- Why register images?
- Classes of registration techniques
- Measuring the quality of image registration
- Examples

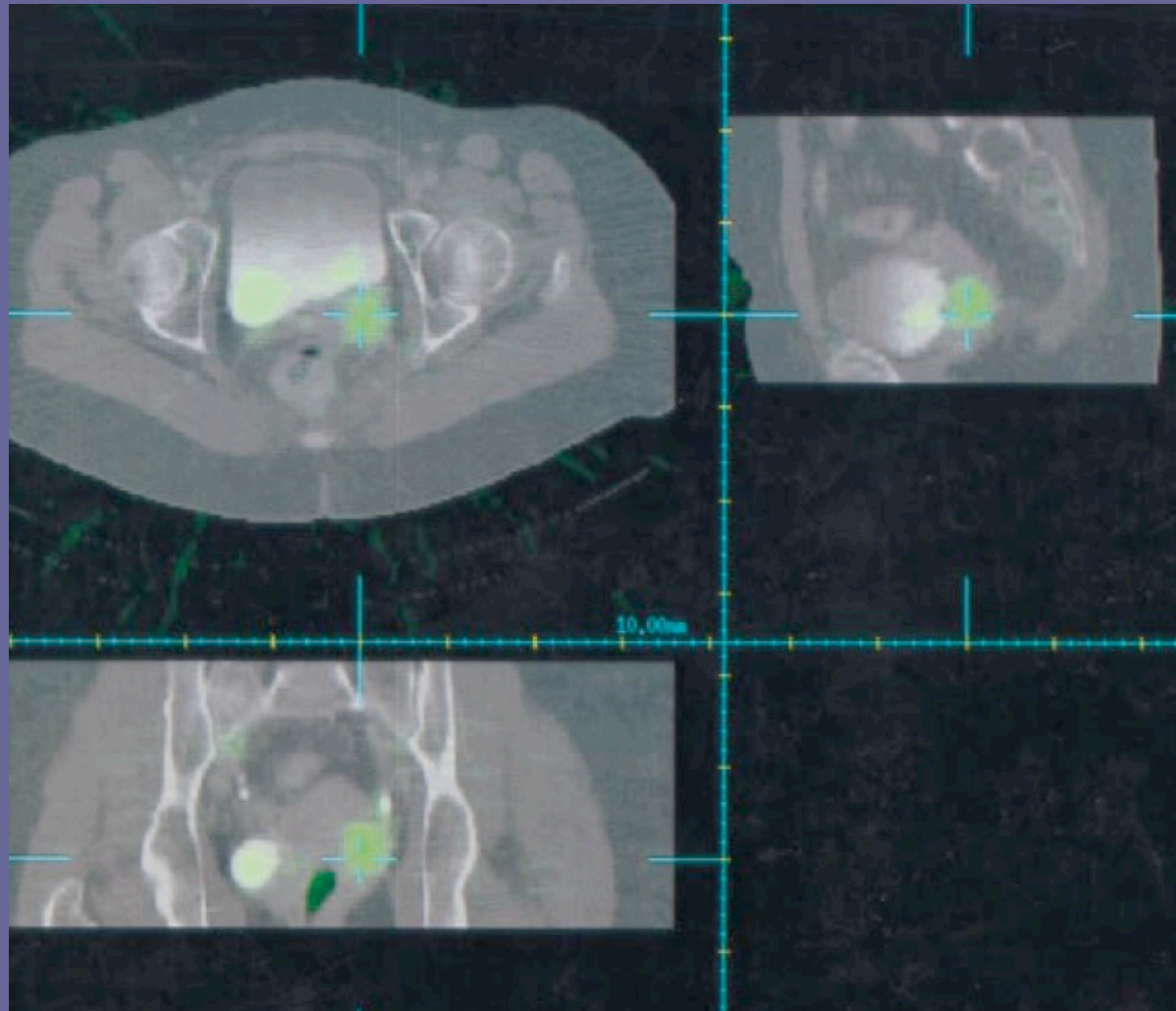
Image registration

- Registration = aligning image data
 - Corresponding physical points lie at the same image coordinates
- In the simplest case, two steps:
 - *Determination of corresponding points in different images*
 - Different times
 - Different imaging modalities
 - Different people
 - *Aligning corresponding points*

Image registration

- Provides a way to combine data across images
 - Without losing spatial specificity
 - An alternative to comparing segmented regions across images
- Simplifies analysis across images
 - Analyze change over time
 - Quantify tumor response to therapy
 - Combine different kinds of information
 - Add function from PET or SPECT to structure from MR or CT
 - Identify effects of disease
 - Do the average image data of patients differ from controls?
 - Group comparisons of brain atrophy
- Makes many quantitative methods of image analysis possible

CT + PET 18-FDG imaging of cervical cancer



Hawkes (1998)

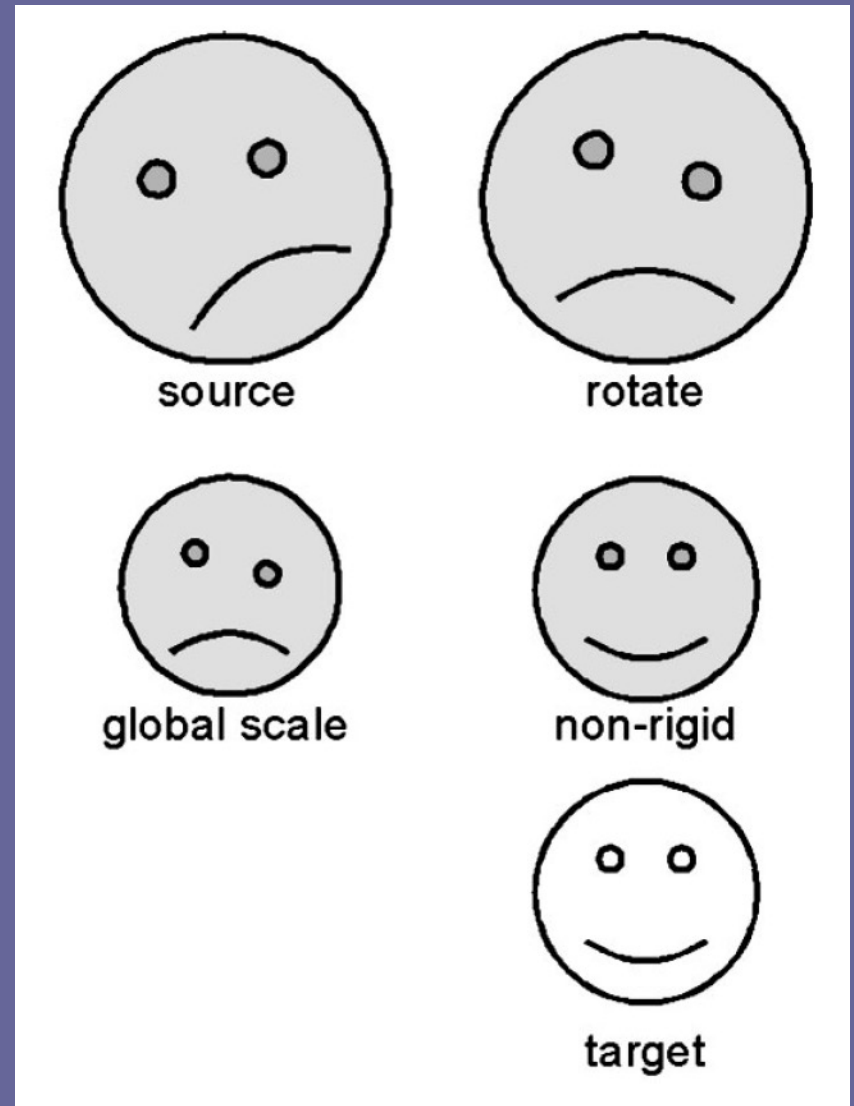
Structure in gray, 18-FDG activity in green

Classes of registration methods

- *Rigid transformations* preserve distances between points
 - Translation
 - Rotation
- *Affine transformations* take parallel lines into parallel lines
 - Includes rigid transformations
 - Scaling (stretching or shrinking)
 - Shearing
- Nonlinear warping

Goal of registration

- To determine the spatial transformation that maps positions in the *source* image to corresponding positions in a *target* image
- Generally consists of both rigid and non-rigid steps



In-class exercise: what transformations are probably required?

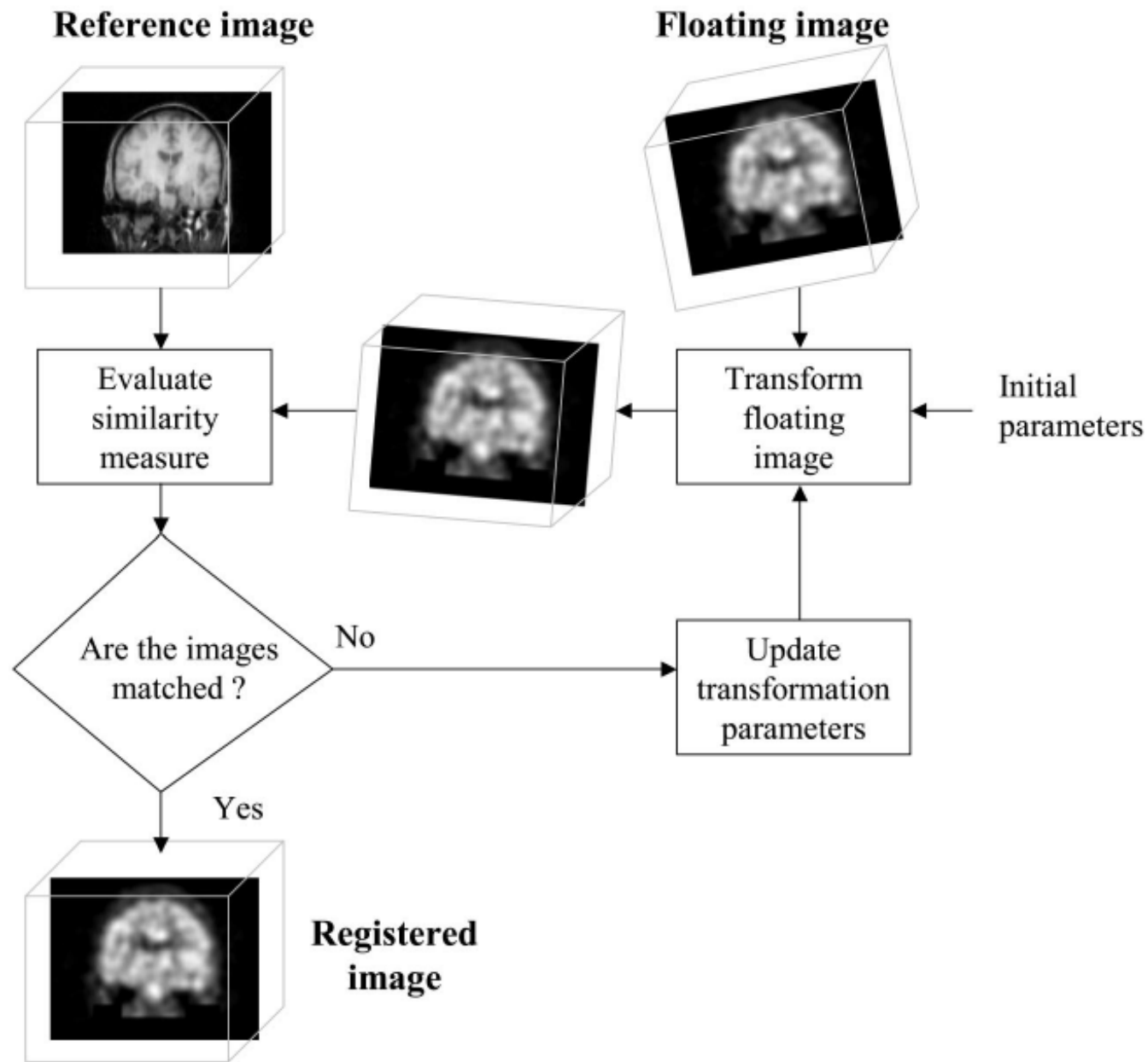
- 1) Attenuation correction of PET images using a CT scan
- 2) Comparing images acquired before and after cancer treatment
- 3) Comparing the heart during contraction (systole) to the relaxed heart (diastole)

Components of registration methods

- A similarity measure
 - How well do source and target images match?
- A transformation model
 - Specifies the type of transformation that makes the source image match the target
 - Rigid
 - Non-rigid
 - Has free parameters that define particular solutions
- An optimization algorithm
 - Finds the parameters of the transformation that maximize similarity between the transformed source and the target

Target

Source



Hutton and Braun (2003)

Similarity measures

- Geometric approaches
 - Best suited to high resolution anatomical modalities
 - Explicit models of anatomic elements
 - Surfaces, curves, point landmarks
 - Must be identified in both source and target images
 - Measure distances between geometric elements
- Intensity approaches
 - Measure similarity of intensity patterns in source and target
- Assume that similarity measure is maximized (or minimized) at the correct registration

Landmark identification

Source



Target



Edges of deformed
image superimposed
on target



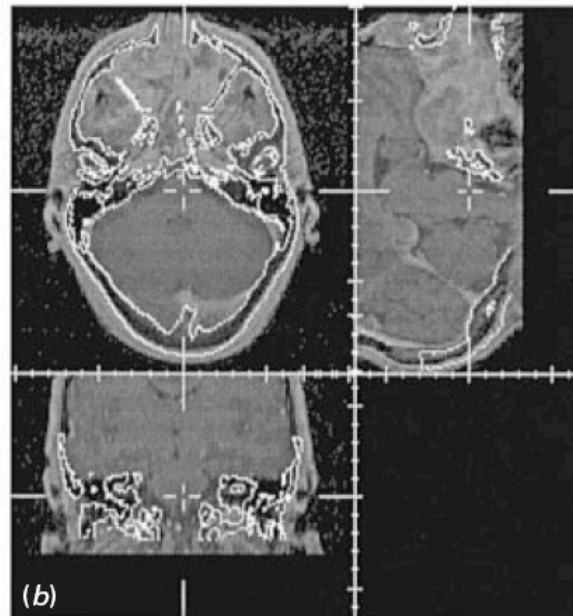
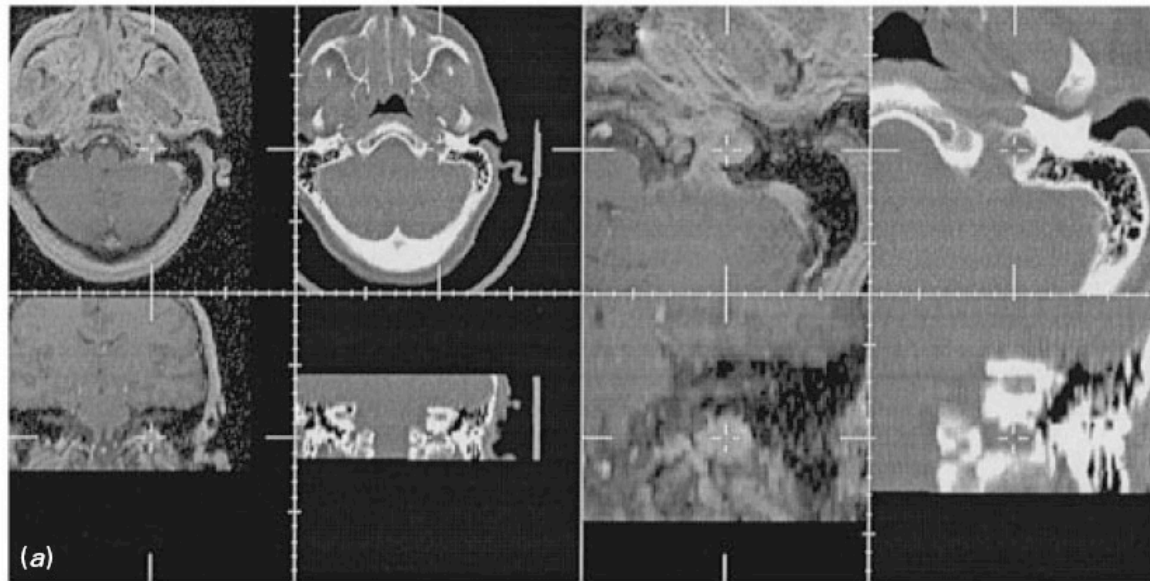
Deformed image



Choosing the similarity measure

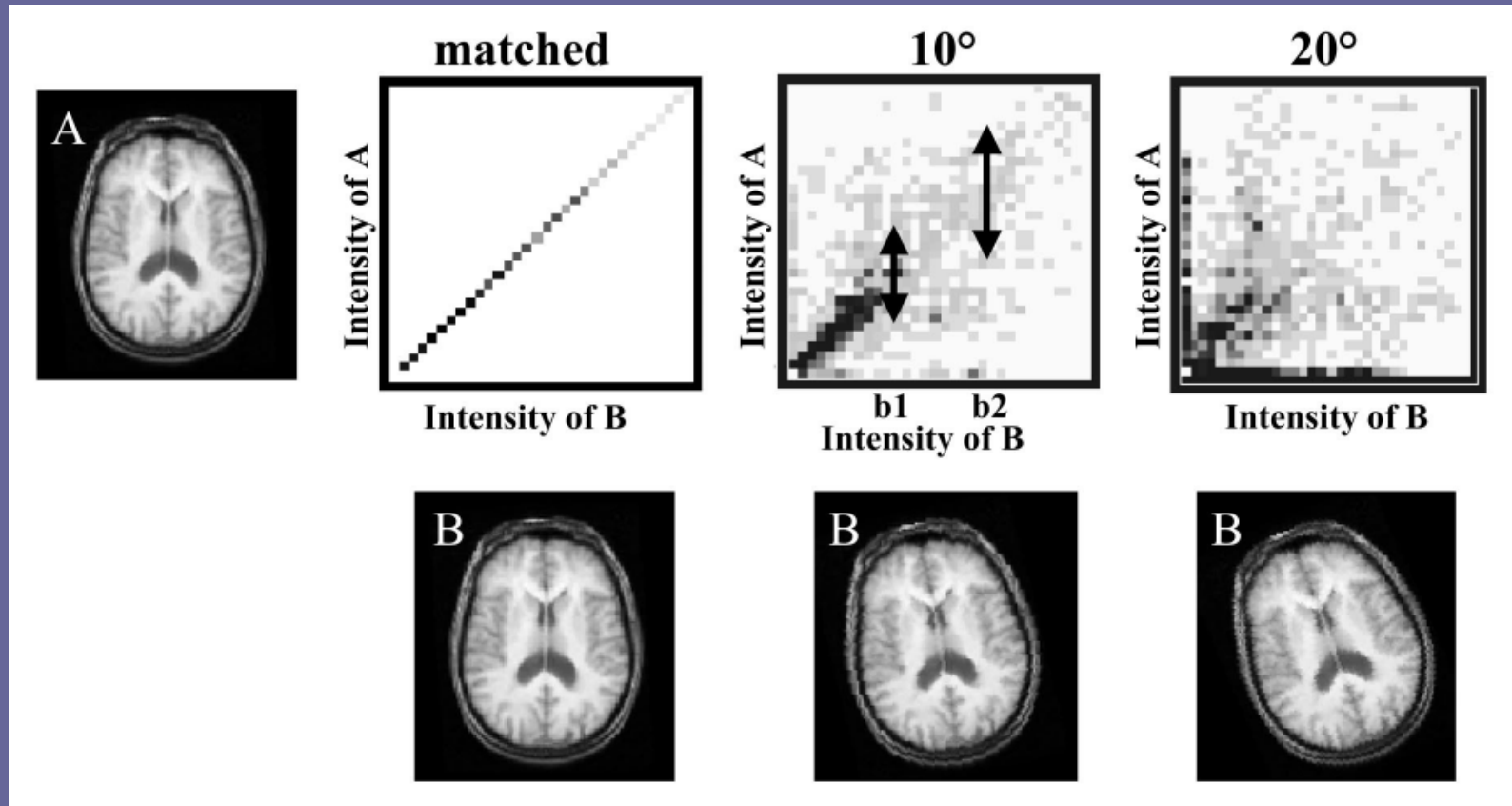
- Images differ only due to noise
 - Sum of squared intensity differences
- Image intensities are linearly correlated
 - Correlation coefficient
- Positions with similar intensities in the source also have similar intensities in the target
 - *Mutual information*
 - Allows for non-linear relationships between image intensities
 - Based on intensity histograms

CT superimposed on MRI



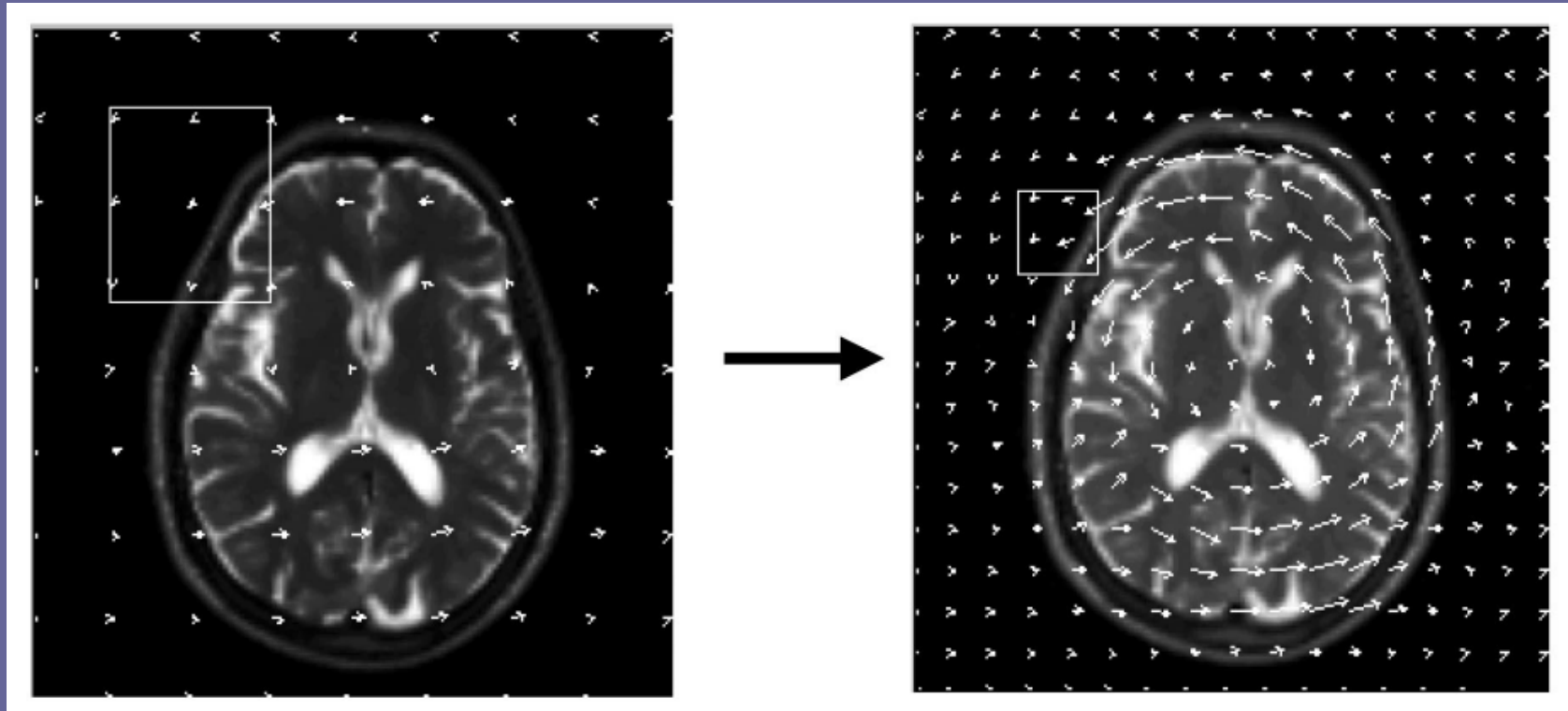
Hawkes (1998)

Joint intensity histogram



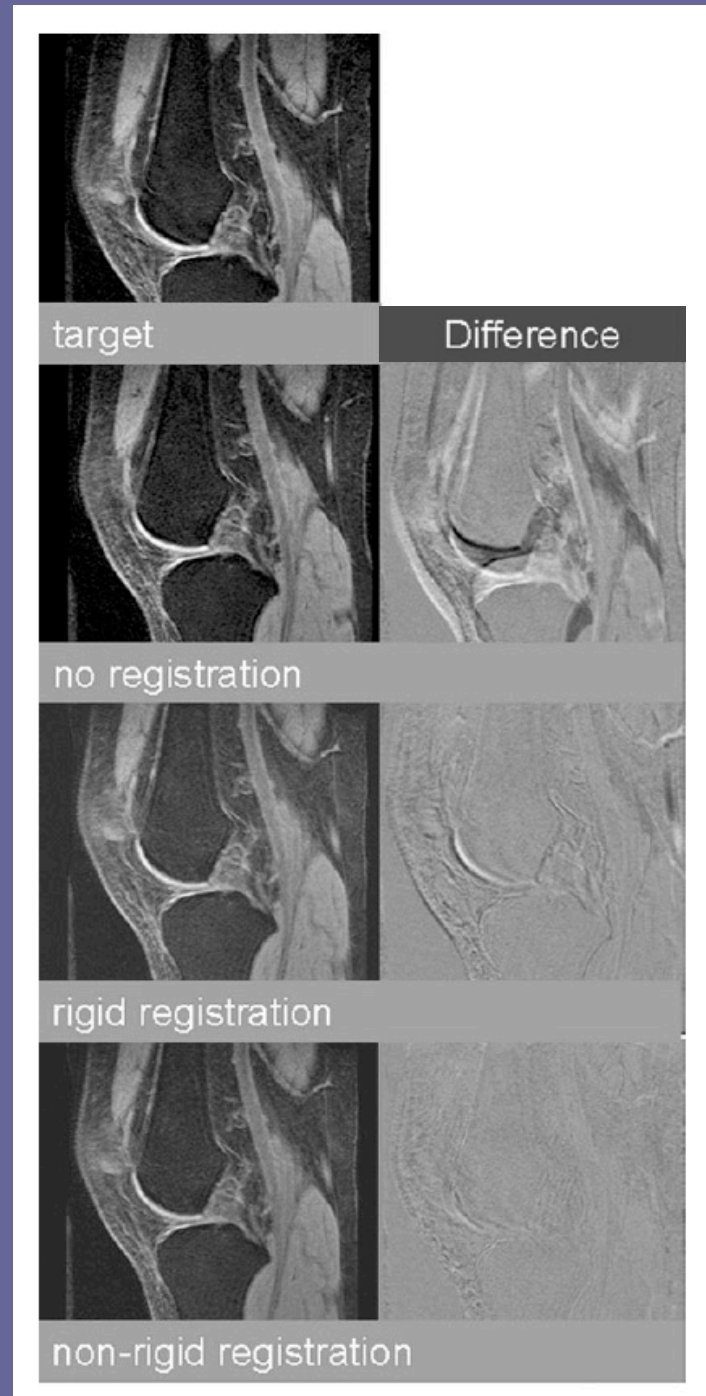
Hutton and Braun (2003)

Non-rigid registration based on local, rigid transformations



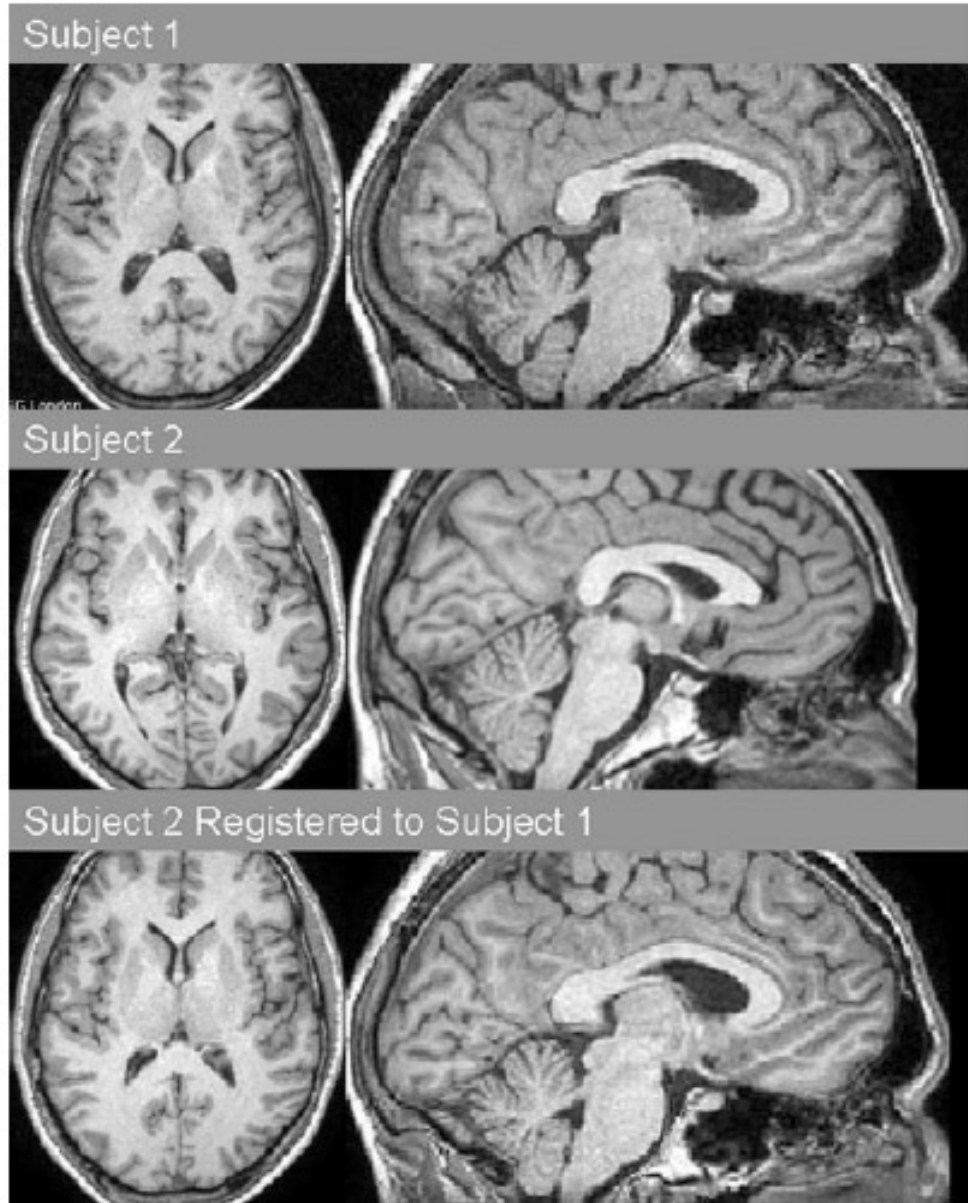
Hutton and Braun (2003)

Example 1: Tracking changes in cartilage thickness

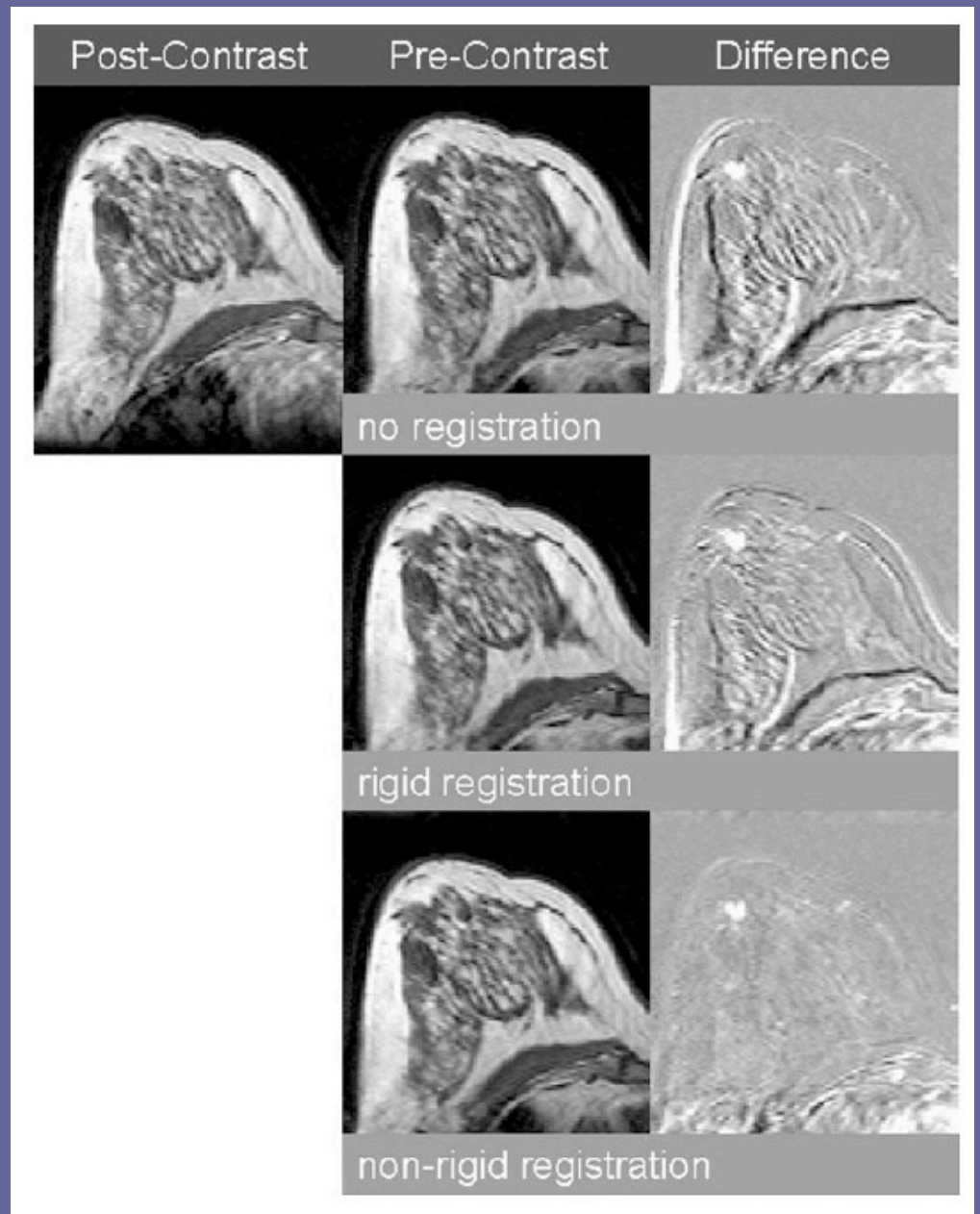


Crum et al (2004)

Example 2: Inter-subject brain matching

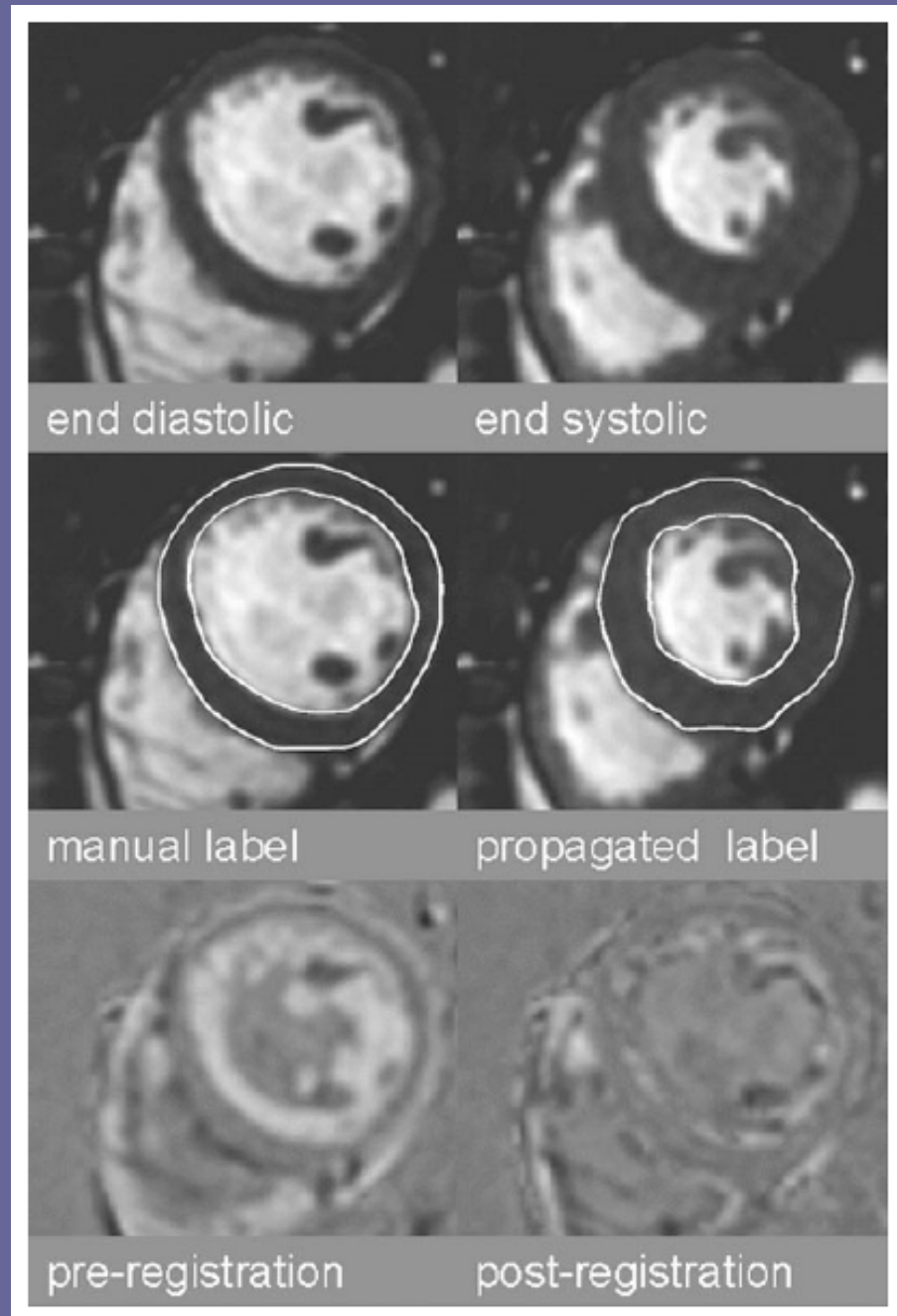


Example 3: contrast-enhanced mammography



Crum et al (2004)

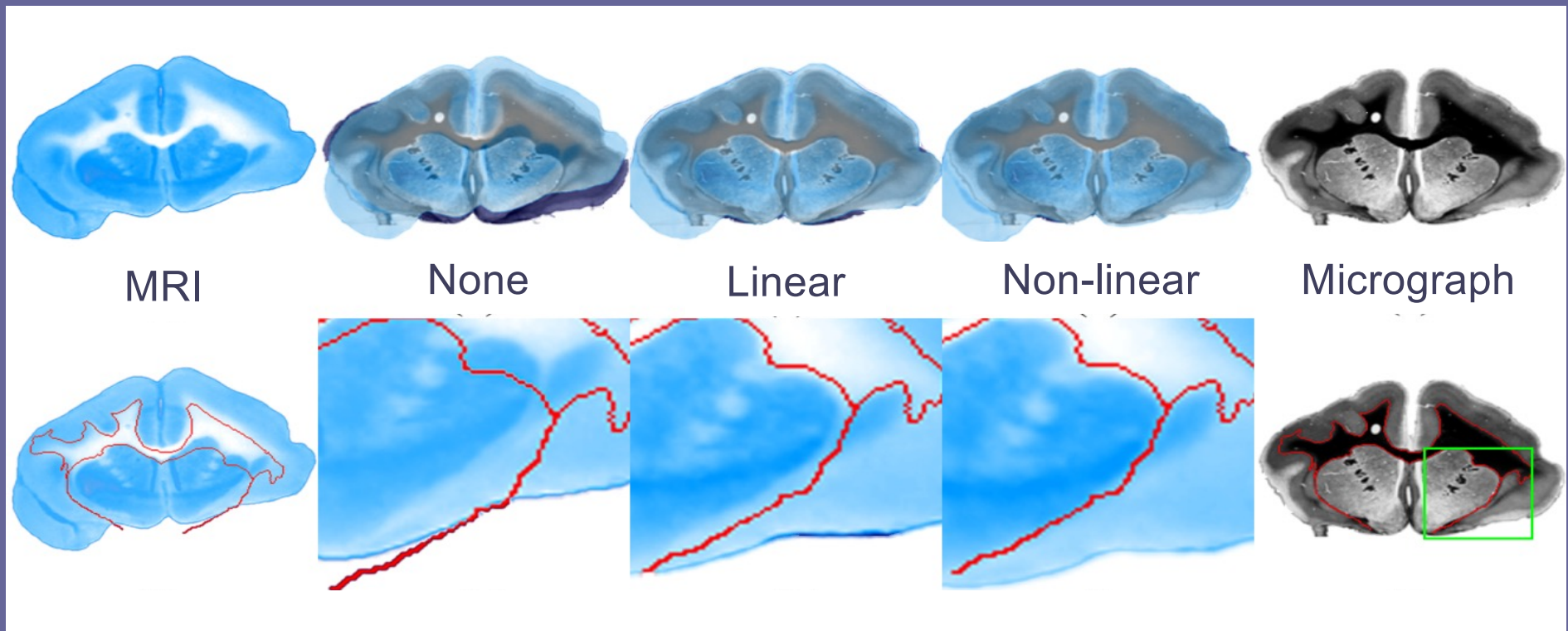
Example 4: myocardial segmentation



Crum et al (2004)

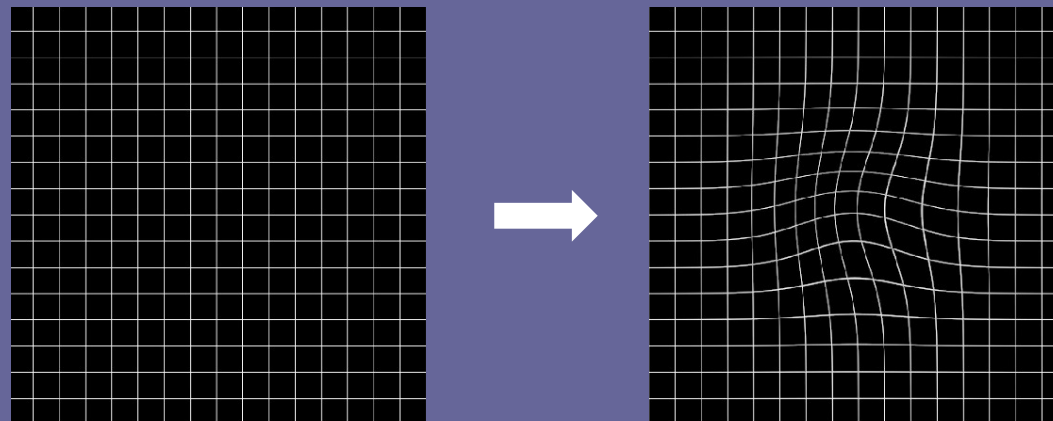
Example 5: MRI to light microscopy

- How is MRI data related to information in light microscopy?
- Section brain tissue and stain for specific molecules (e.g., myelin)



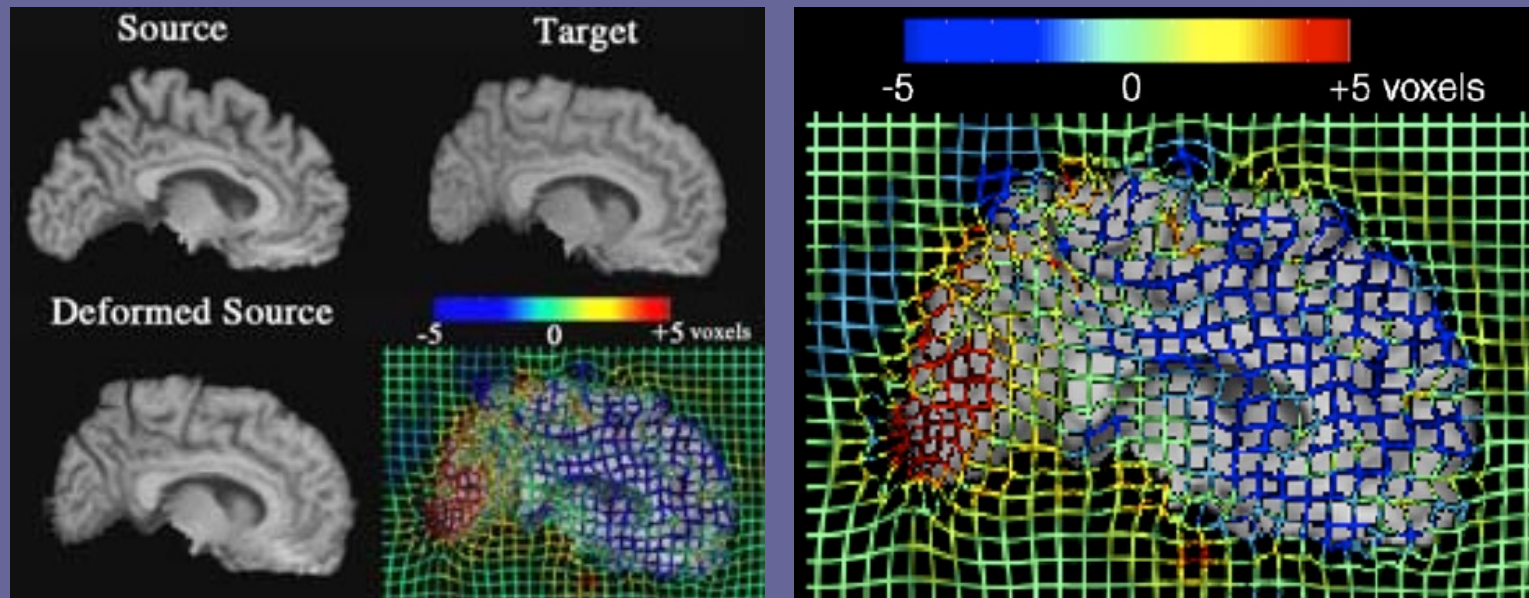
Use information from registration

- Image transformations contain useful information
 - Local increase or decrease in volume
 - Use this to compare local volume differences



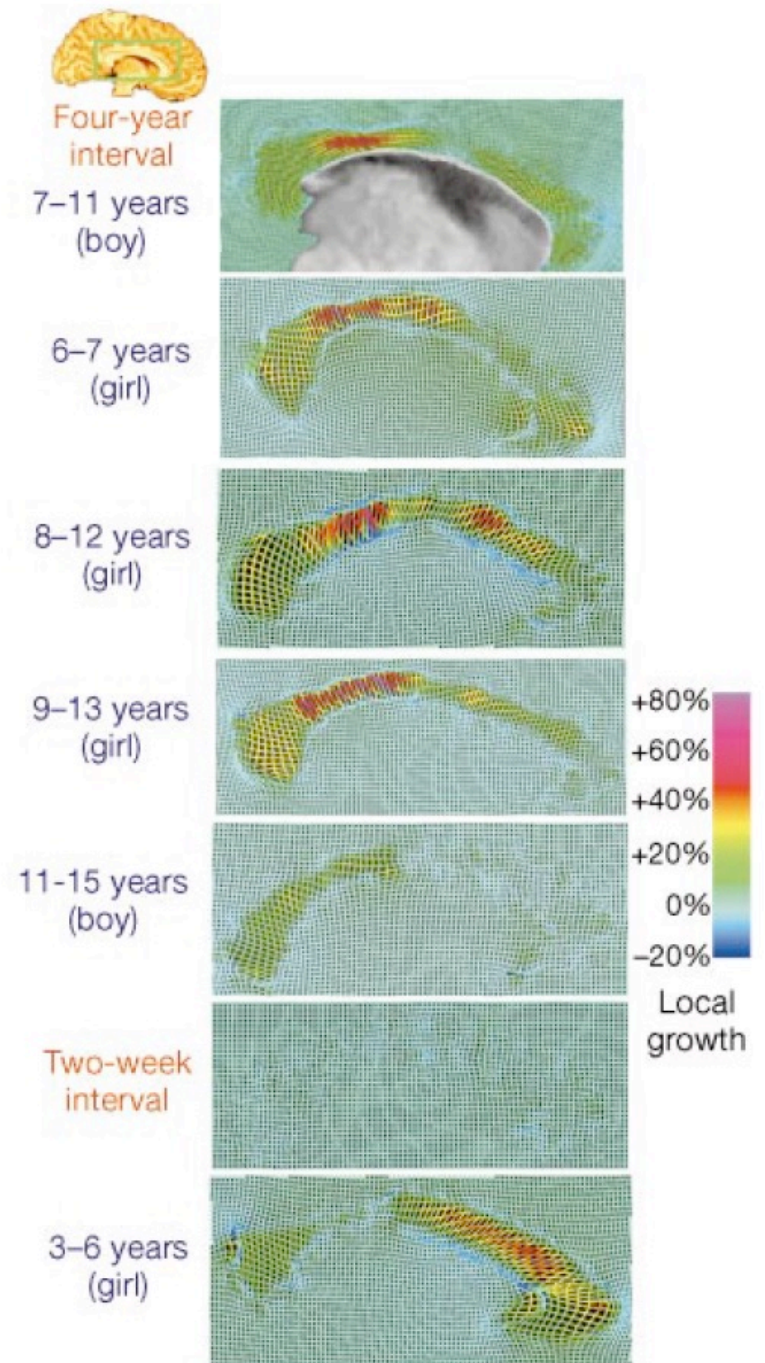
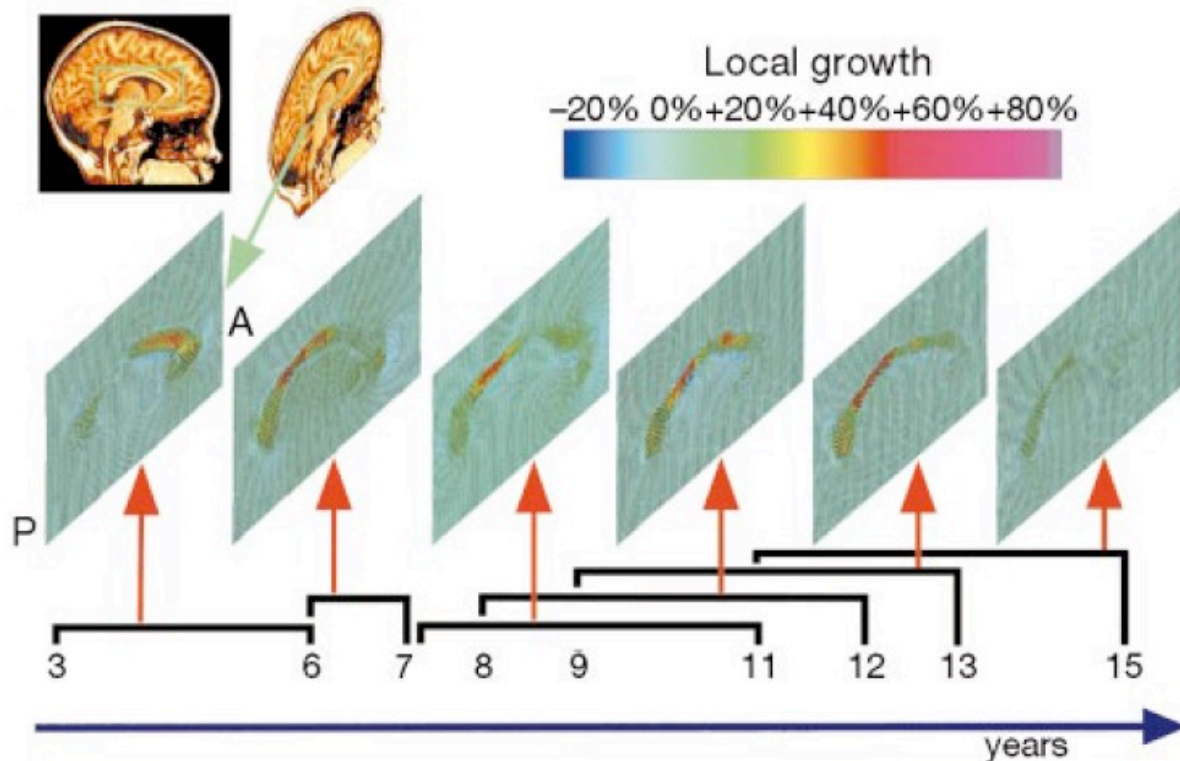
Applications to morphology

- Image registration may involve non-rigid transformations
 - Local increase or decrease in volume
 - Use this to compare local volume differences



Thompson (2007)

Measuring volume changes over time



Example 6: motion correction

- Suppose the subject moves slightly between 2 images—how should the image intensity of the 2nd image (the source) be changed to match the 1st image (the target)?
- Image intensity is

$$I'(x, y) = I(x + \Delta x, y + \Delta y) = I(x, y) + \Delta x \frac{dI}{dx} + \Delta y \frac{dI}{dy} + \dots$$

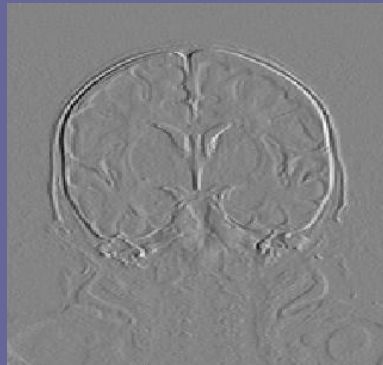
$$I'(x, y) - I(x, y) \approx \Delta x \frac{dI}{dx} + \Delta y \frac{dI}{dy}$$

How can we solve for Δx and Δy ?



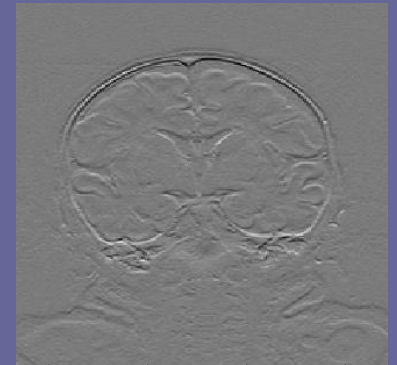
$I' - I$

$= \Delta x$



dl/dx

$+ \Delta y$



dl/dy

- Writing

$$I'(x, y) - I(x, y) = \Delta x \frac{dI}{dx} + \Delta y \frac{dI}{dy}$$

for many image points (x_i, y_i) in a matrix equation:

$$\begin{pmatrix} I'(x_1, y_1) - I(x_1, y_1) \\ I'(x_2, y_2) - I(x_2, y_2) \\ \vdots \\ I'(x_N, y_N) - I(x_N, y_N) \end{pmatrix} = \begin{pmatrix} \left. \frac{dI}{dx} \right|_{x_1, y_1} & \left. \frac{dI}{dy} \right|_{x_1, y_1} \\ \left. \frac{dI}{dx} \right|_{x_2, y_2} & \left. \frac{dI}{dy} \right|_{x_2, y_2} \\ \vdots & \vdots \\ \left. \frac{dI}{dx} \right|_{x_N, y_N} & \left. \frac{dI}{dy} \right|_{x_N, y_N} \end{pmatrix} \cdot \begin{pmatrix} \Delta x \\ \Delta y \end{pmatrix}$$

Fitting for Δx and Δy provides an estimate of $I'(x, y)$ with the effects of motion removed.

Summary

- Registration aims to find corresponding points in two images.
- Makes possible more sophisticated image analysis
 - Combine data from different images
 - Analyze transformation for shape/volume changes
- A registration algorithm has 3 parts:
 - A similarity measure
 - A model for the transformation
 - An optimization algorithm
- Typically large scale features are registered better than small ones—active area of research

References

- Crum WR, Hartkens T, Hill DLG. Non-rigid image registration: theory and practice. The British Journal of Radiology 77:S140-S153 (2004).
- Hawkes DJ. Algorithms for radiological image registration and their clinical application. J. Anat. 193: 347-361 (1998).
- Hutton BF, Braun M. Software for Image Registration: Algorithms, Accuracy, Efficacy. Seminars in Nuclear Medicine, 33(3): 180-192 (2003).