

# Medical Image Analysis

CS 778 / 578

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# Outline

- 1 Coordinate Transformations
- 2 Principal Axes Transformation

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## 1 Coordinate Transformations

- Global Transformations
- Local Transformations

## 2 Principal Axes Transformation

## Problem Definition

Image registration is the process of determining a coordinate transformation between two images that are misaligned.

$$\min_T \text{dist}(I_1(\mathbf{x}), I_2(T(\mathbf{x})))$$

- $T$  is a coordinate transformation
- $I_1(\mathbf{x})$  and  $I_2(\mathbf{x})$  are 2 images to be aligned
- $\text{dist}(I_1, I_2)$  is a metric which determines how well the images match.
- $\text{dist}(I_1, I_2)$  can be based on image intensities or extracted features.

# Linear transformations

- Translation
- Rotation
- Scaling
- Shear

## 2D Linear transformations

Translation : 2 parameters

$$T(\mathbf{x}) = \mathbf{x} + \mathbf{t}$$

Rotation about the origin : 1 parameter

$$T(\mathbf{x}) = R_z \mathbf{x} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \mathbf{x}$$

Nonuniform scaling : 2 parameters

$$T(\mathbf{x}) = S \mathbf{x} = \begin{bmatrix} s_x & 0 \\ 0 & s_y \end{bmatrix} \mathbf{x}$$

Shear : 1 parameter

$$T(\mathbf{x}) = C \mathbf{x} = \begin{bmatrix} 1 & \cot \theta \\ 0 & 1 \end{bmatrix} \mathbf{x}$$

## 2D Linear transformations

A more general 2D transformation can be obtained by composing several transformations, such as:

$$T(\mathbf{x}) = R_z S(\mathbf{x} - \mathbf{c}) + \mathbf{t}$$

- Translate so that center of the rotation is  $[0,0]$ .
- Scale the coordinate systems.
- Rotate about the origin.
- Translate.
- Total of 7 parameters.

## 3D Linear transformations

- Translation : 3 parameters
- Scale : 3 parameters
- Shear : 2 parameters
- Rotation : 3 Euler angles

Using Euler angles, the 3D rotation is represented as 3 consecutive rotations about the coordinate axes.

$$T(\mathbf{x}) = R_x R_y R_z \mathbf{x}$$

where  $R_x R_y R_z =$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi & \cos \phi \end{bmatrix} \begin{bmatrix} \cos \psi & 0 & \sin \psi \\ 0 & 1 & 0 \\ -\sin \psi & 0 & \cos \psi \end{bmatrix} \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



## Rotation matrices

Rotation by  $\theta$  about the origin is represented by the matrix

$$R_z(\theta) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}.$$

*Rotation matrices are orthogonal :  $R^{-1} = R^T$ .*

$$R_z(\theta)^{-1} = R_z(-\theta) = \begin{bmatrix} \cos(-\theta) & -\sin(-\theta) \\ \sin(-\theta) & \cos(-\theta) \end{bmatrix}$$

Since  $\cos$  is an even function  $\cos(-\theta) = \cos(\theta)$   
and  $\sin$  is an odd function  $\sin(-\theta) = -\sin(\theta)$ ,

$$R_z(-\theta) = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} = R_z(\theta)^T$$

## 2D Rotation : Geometric derivation

Rewrite in polar coordinates:

$$x = \rho \cos \phi$$

$$y = \rho \sin \phi$$

$$x' = \rho \cos(\theta + \phi)$$

$$y' = \rho \sin(\theta + \phi)$$

Using the trig identities

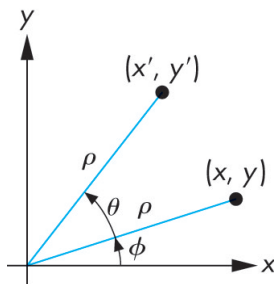
$$\cos(\theta + \phi) = \cos \phi \cos \theta - \sin \phi \sin \theta$$

$$\sin(\theta + \phi) = \cos \phi \sin \theta + \sin \phi \cos \theta$$

Rewrite  $x', y'$  in terms of  $x, y$

$$x' = \rho \cos \phi \cos \theta - \rho \sin \phi \sin \theta = x \cos \theta - y \sin \theta$$

$$y' = \rho \cos \phi \sin \theta + \rho \sin \phi \cos \theta = x \sin \theta + y \cos \theta$$



## 2D Rotation : Geometric derivation

We can rewrite this system of equations

$$\begin{aligned}x' &= x \cos \theta - y \sin \theta \\y' &= x \sin \theta + y \cos \theta\end{aligned}$$

in matrix form as

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

This is equivalent to 3D rotation about the z-axis

$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

## Transformation by a displacement Field

Compute a displacement vector for each voxel.

$$T(\mathbf{x}) = \mathbf{x} + \mathbf{t}(\mathbf{x})$$

To constrain the displacement field to represent physically plausible deformations, we may impose smoothness constraints.

If  $\mathbf{t}(\mathbf{x}) = [u(x, y), v(x, y)]$ ,

$$\min_{u,v} \int \int ((u_x^2 + u_y^2) + (v_x^2 + v_y^2)) dx dy$$

will constrain the displacement field to be smooth

# Displacement Field

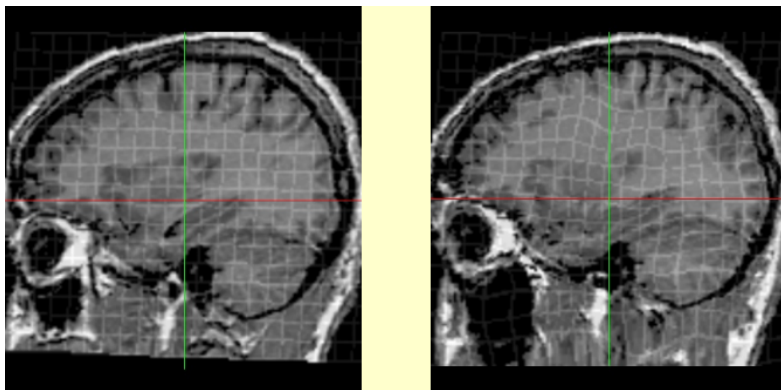
## Viscoelastic regularization methods:

- Consider the deformation field to be the velocity field of some viscous fluid.
- More suitable for large deformations
- Constrain the field to obey the Navier-Stokes equation.
- Smoothness of the field is controlled by the viscosity of the simulated fluid.
- Computationally expensive

# Spline based transformations

- Fewer control points than image pixels.
- The spline may interpolate or approximate the control points.
- Sum of shifted basis functions.
- Basis functions may have local or global support.
- Basis functions are generally low degree (3) polynomials.

# Spline based transformations



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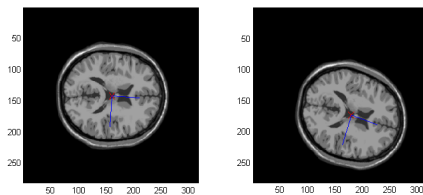
# Principal Axes Transformation

Characterize images by

- Centroid (a 2D or 3D point)
- Principal directions (2 or 3 perpendicular vectors)

Images  $I_1$  and  $I_2$  can be aligned by

- Translating centroid 2 to be coincident with centroid 1
- Rotate about centroid 1 so that the principal directions are aligned.



## Image centroid

The expected value of a function,  $f$ , with respect to another function  $B(x)$  is defined as

$$E_B[f] = \frac{\int_{\Omega} f(x)B(x)dx}{\int_{\Omega} B(x)dx}.$$

The centroid,  $\mathbf{c}_I$ , of the image,  $I(\mathbf{x})$  is  $E_I[\mathbf{x}]$ .

- $\mathbf{c}_I$  is the center of the distribution of image intensities.
- $\mathbf{c}_{I_1}$  and  $\mathbf{c}_{I_2}$  should be corresponding points in the 2 images.

## Image principal axes

The covariance,  $cov_I$  of the image,  $I(\mathbf{x})$  is  $E_I[(\mathbf{x} - \mathbf{c}_I)(\mathbf{x} - \mathbf{c}_I)^T]$ .

$$cov_I = \Sigma \Lambda \Sigma^T$$

where

- $\Lambda$  is a diagonal matrix (scaling) describing the variation of  $I$  in the principal directions.
- $\Sigma$  is an orthogonal matrix (rotation) which rotates the coordinate axes onto the principal directions.
- The columns of  $\Sigma$  are the principal directions.

So if

- $cov_{I_1} = \Sigma_1 \Lambda_1 \Sigma_1^T$
- $cov_{I_2} = \Sigma_2 \Lambda_2 \Sigma_2^T$

then the rotation which aligns the axes of  $I_2$  with the axes of  $I_1$  is  $\Sigma_1 \Sigma_2^T$

# Principal axes image registration

- Compute  $c_{I_1}$ , and  $c_{I_2}$
- Compute  $cov_{I_1}$  and  $cov_{I_2}$
- Compute  $\Sigma_1$  and  $\Sigma_2$
- The coordinate transformation which aligns image 2 with image 1 is  

$$x' = \Sigma_1 \Sigma_2^T (x - c_{I_2}) + c_{I_1}$$

