

Image Registration 2:

Spring 2021

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Handouts & Lecture Notes

- Report in Scientific American (June 2014):
“In each study, however, those who wrote out their notes by hand had a stronger conceptual understanding and were more successful in applying and integrating the material than those who used [sic] took notes with their laptops.”

The Pen Is Mightier Than the Keyboard

P. A. Mueller, D. M. Oppenheimer, *Psychological Science*, Vol 25, Issue 6, pp. 1159 – 1168, April-23-2014.

- Handouts are to aid note taking, not a total replacement for note taking
- Podcasts, slides, pdfs etc on BlackBoard

Image Warping & Regularisation

Warp Regularisation

- **Arbitrary** image warp $\phi : \underline{x} \mapsto \phi(\underline{x})$ is **ill-posed problem**
- Need to **constrain/control** image warps:

By construction:

- **Parametric Image Warps:**

Warp defined by some small set of parameters
Spline with some set of control/knot points
Interpolation/extrapolation of displacement

e.g. AAM: Triangles & barycentric coordinates

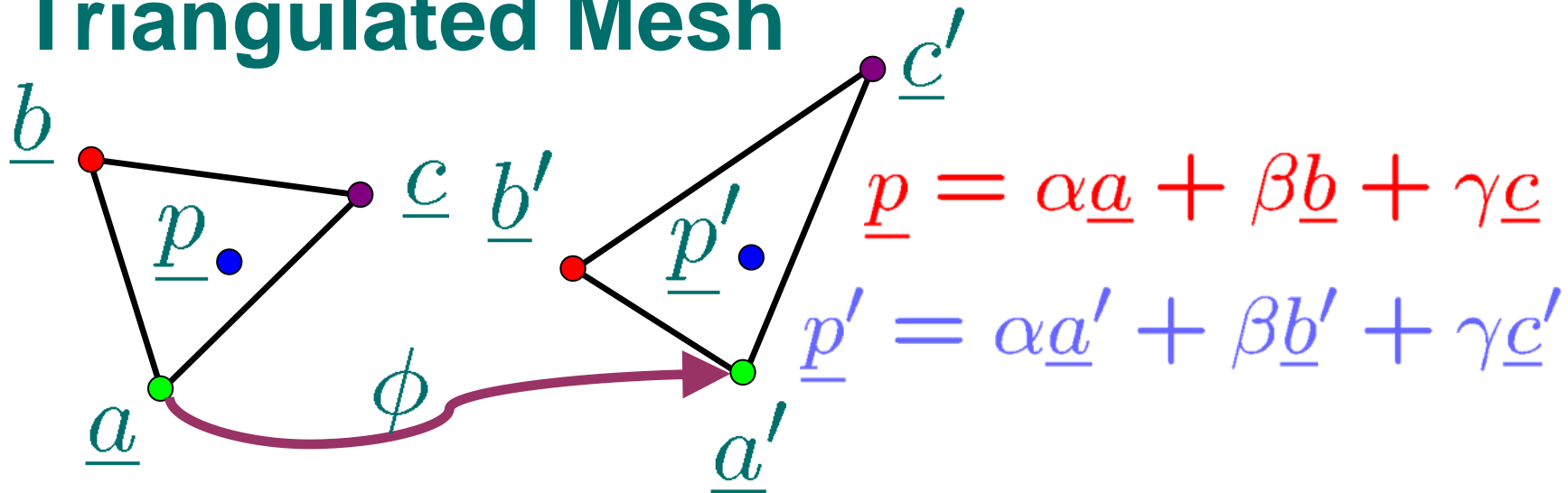
By warping penalty:

- **Non-Parametric Image Warps:**

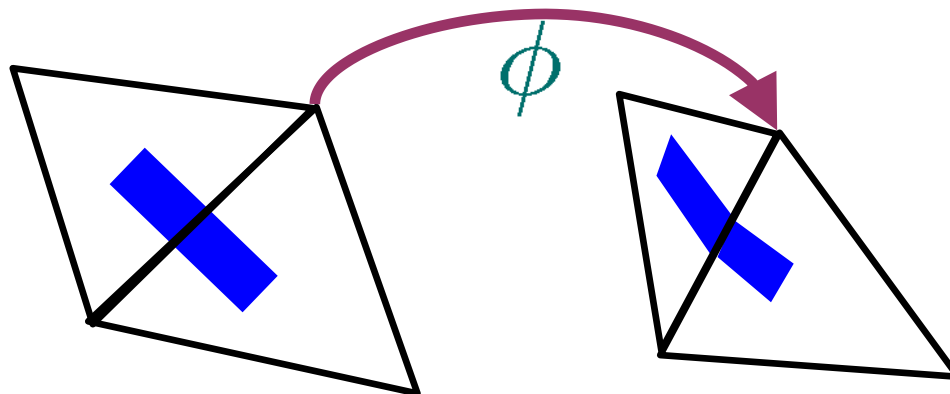
Trade-off between image match versus smoothness
e.g. **elastic** and **fluid** registration

Parametric Image Warps

Triangulated Mesh

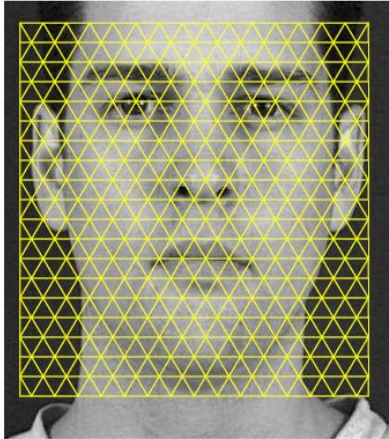


- **Triangulated** mesh and **barycentric** coordinates
- Tends to bend objects at edges



**Continuous but
not differentiable
at edges**

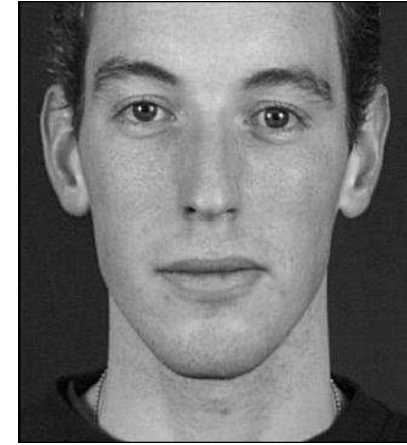
Image Registration: Triangulated Meshes



example mesh

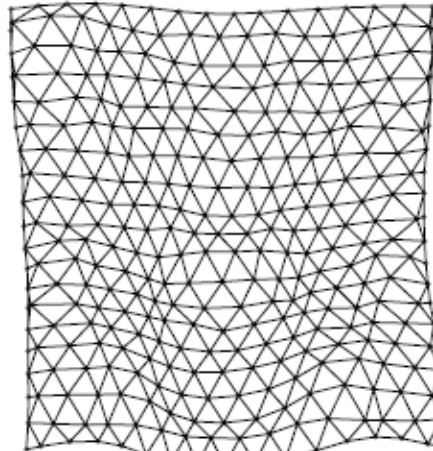


warped reference



Target

Note:
Uses group of
images,
reference is a
mean face

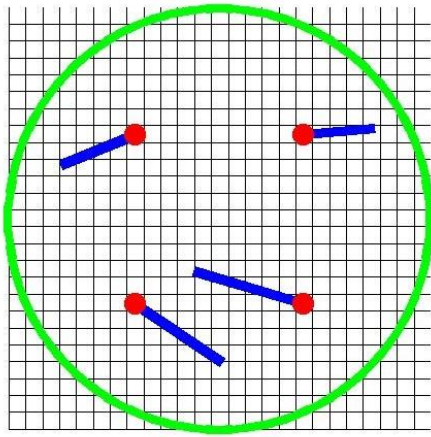


warped mesh

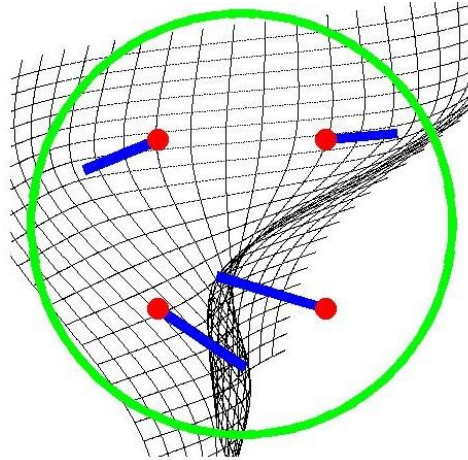


comparison

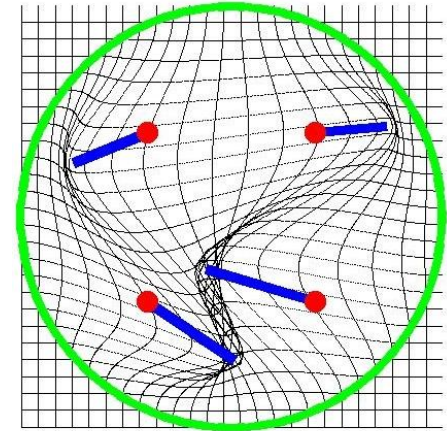
Thin-Plate & Clamped-Plate Splines



Initial knotpoint positions and **displacements** and **unit circle**.



Biharmonic Thin-Plate Spline (TPS)
asymptotically flat.
(Duchon 1976)

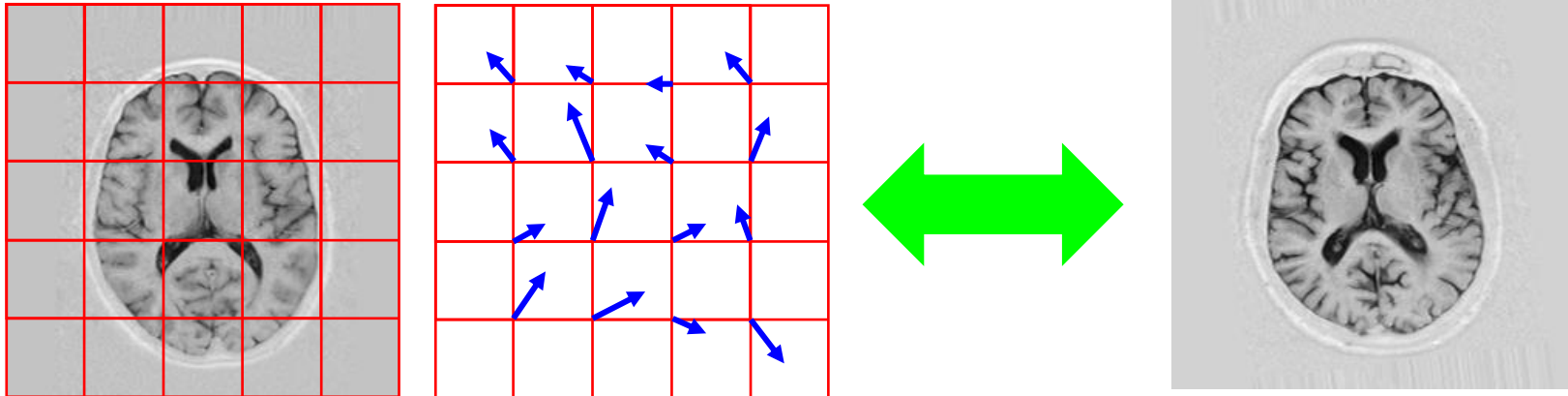


Biharmonic Clamped-Plate (CPS)
zero outside **unit circle**.

(Marsland & Twining 2002)

- **Displacement field: analytic function of knotpoints**
- **Can fold if displacements too large**
- **Other splines: B-splines (Rueckert), cubic splines etc**

Parametric Registration:



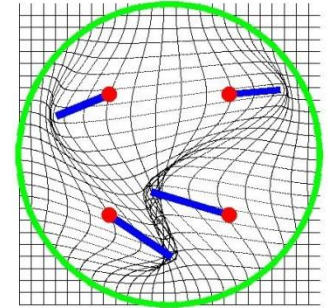
- **Initialise:** Rigid (translation & rotation), affine (+scale)
- Define initial knotpoint positions
- Define knotpoint displacements
- Extrapolate/interpolate to find warp $\phi : \underline{x} \mapsto \phi(\underline{x})$
- Create warped image and do comparison
- Optimise displacements (& initial positions)
- Repeat at finer scale

Non-Parametric Image Warps

Warp Regularisation:

Parametric Warps:

- Limited number of degrees of freedom
- Controls warps & easier to optimise
- Still have to check for folding
- Limited in terms of flexibility



Dense, non-parametric Warps:

- Pixel-by-pixel deformations

Varying notation:
where it goes to
OR
where it came from

$$\phi : \underline{x} \mapsto \phi(\underline{x}) \quad \underline{u} : \underline{x} \mapsto \underline{x} \boxed{\pm} \underline{u}(\underline{x})$$

deformation field displacement field

- Need to add **regularisation term** to control warps
- Physics-inspired algorithms:

Elastic solid or **visco-elastic fluid**

Framework: Non-Parametric Registration

- **Two images:** $\mathcal{I}_s(\underline{x})$ & $\mathcal{I}_t(\underline{x})$ & **warp:** $\phi(\underline{x})$
- **Image-matching/difference term:** $\mathcal{D}(\mathcal{I}_s, \mathcal{I}_t; \phi)$

Warp & resample, push-forward or pull-back mapping
SAD, SSD, mutual information etc

- **Warp regularisation term:** $\mathcal{R}(\phi)$
- **Total objective function/energy:**

$$\mathcal{E} = \mathcal{D}(\mathcal{I}_s, \mathcal{I}_t; \phi) + \lambda \mathcal{R}(\phi)$$

image
matching term

trade-off
parameter

warp roughness
term

Energies & Forces

Energy minimisation:

$$\mathcal{E} = \mathcal{D}(\mathcal{I}_s, \mathcal{I}_t; \phi) + \lambda \mathcal{R}(\phi)$$

- **Less-smooth warps vs better match/smaller difference**

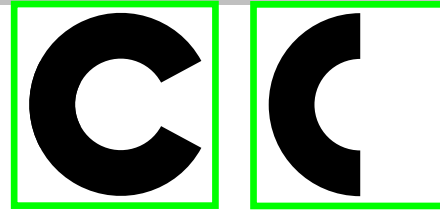
Forces:

- **Image forces: tries to align structures**
- **Warp forces: resists deformation**
- **Forces = ‘gradient’ of relevant energy term**

Solution:

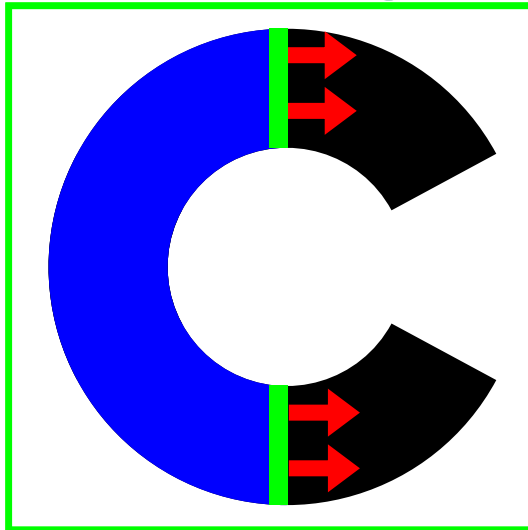
- **Minimize energy OR zero net force**

Image Forces:

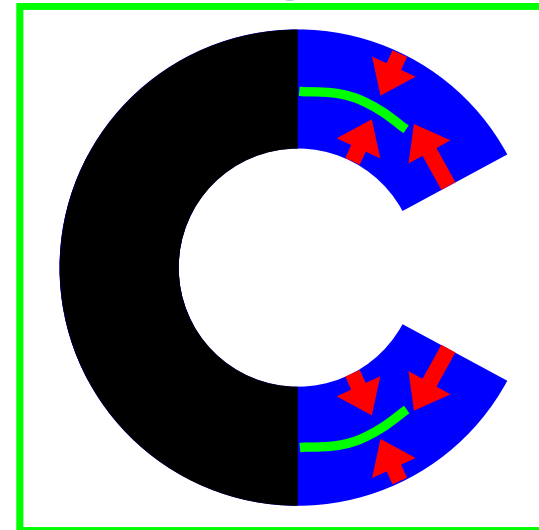


Two choices for **moving** image

Arch warping to C

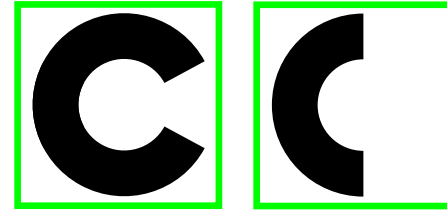


C warping to arch



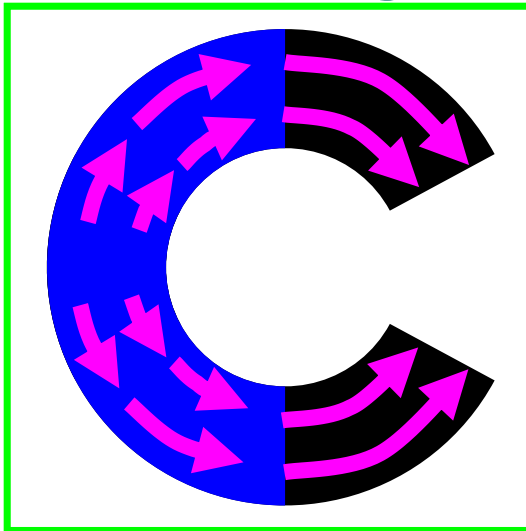
- Moving leading edge of arch improves image match
- Without warp regulariser, only these pixels move
- Image forces try to shrink non-overlap region
- Only shrink whole shape if warp regulariser acts

Preferred Solution:

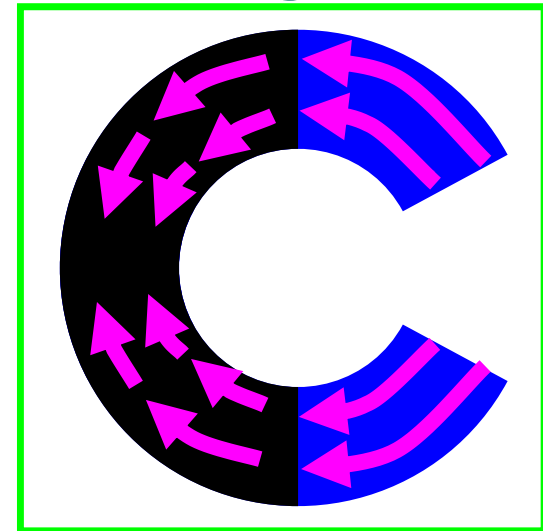


Two choices for **moving** image

Arch warping to C



C warping to arch



- Edge of arch moves to edge of C
- Rest of arch stretches as well
- Edge of C contracts to edge of arch
- Rest of C contracts as well

Only movement of object so far. What about background?

Suggestion: Elasticity

Make object elastic:

- Agrees with our intuition
- Allows more natural correspondence between warped and target object

Problems:

- Can't warp just object - if we knew how to separate object & background, wouldn't need registration!
- Real images not this simple

Solution:

Treat whole moving image as if printed on elastic sheet

Can both stretch and compress

- Need deformation energy for warped elastic sheet

Mathematical Aside:

Mathematics in Image Registration

- **Maths as a language:**

 - Concise and precise statements

 - Own special symbols and own syntax

 - Need a logical approach

- **If you can program, already have required skills**

- **Most cases:**

 - Translation** rather than manipulation

Vector Calculus in 3D: Notation

$$\vec{\nabla} = \left(\frac{\partial}{\partial x}, \frac{\partial}{\partial y}, \frac{\partial}{\partial z} \right)$$

first derivative, vector-like

Note: Will find \mathbf{X} as cartesian coordinate and \underline{x} as position vector

$$\text{div(ergence)}: \vec{\nabla} \cdot \underline{n}(\underline{x}) = \frac{\partial n_x}{\partial x} + \frac{\partial n_y}{\partial y} + \frac{\partial n_z}{\partial z}$$

first derivative acting on vector gives scalar

del squared:

$$\vec{\nabla}^2 \underline{n} = (\vec{\nabla}^2 n_x, \vec{\nabla}^2 n_y, \vec{\nabla}^2 n_z)$$

$$\vec{\nabla}^2 n_x = \frac{\partial^2 n_x}{\partial x^2} + \frac{\partial^2 n_x}{\partial y^2} + \frac{\partial^2 n_x}{\partial z^2}$$

second derivative acting on vector gives another vector

- See Maths Primer for more details

Mathematics in Image Registration

The diagram illustrates the advection equation with color-coded annotations:

- vector field** (blue text): Points to the velocity vector \underline{v} in the first term and the divergence term $(\underline{v} \cdot \underline{\nabla})$.
- vector field** (red text): Points to the scalar field u in the second term.
- position** (magenta text): Points to the position vector \underline{x} .
- time** (magenta text): Points to the time variable t .
- first derivatives** (green text): Points to the time derivative $\frac{\partial}{\partial t}$ and the spatial gradient operator $\underline{\nabla}$.

$$\underline{v}(\underline{x}, t) = \frac{\partial u(\underline{x}, t)}{\partial t} + (\underline{v}(\underline{x}, t) \cdot \underline{\nabla}) u(\underline{x}, t)$$

- Relation between a pair of vector fields that depend on time and on position
- Related via first-derivatives acting on \underline{u}
- If you know \underline{u} , can compute Eulerian velocity \underline{v}

Mathematics in Image Registration

The diagram shows the Navier-Stokes equation for image registration with color-coded annotations:

$$\underline{F}^{\text{visc}}(\underline{x}, t) = \text{constants}$$

Annotations:

- result** (pink arrow) points to $\underline{F}^{\text{visc}}(\underline{x}, t)$.
- Eulerian velocity** (blue arrow) points to $\underline{v}(\underline{x}, t)$.
- second-order spatial derivatives** (green arrow) points to ∇^2 .

The equation is:

$$\mu \nabla^2 \underline{v}(\underline{x}, t) + (\lambda + \mu) \nabla (\nabla \cdot \underline{v}(\underline{x}, t))$$

- Output is a vector field, function of space & time
- Depends on second-order spatial derivatives of Eulerian velocity (previous slide)
- Two constants that need to be assigned values
- Viscous forces in a fluid, vary according to flow

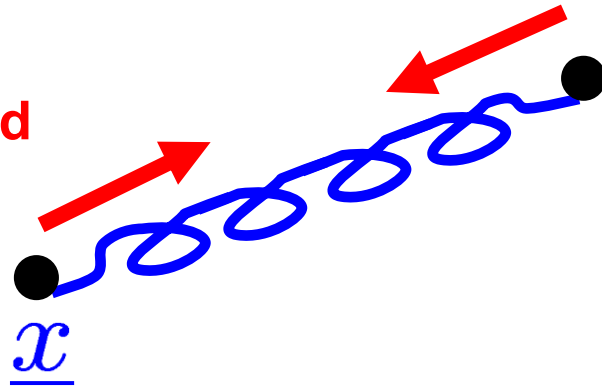
Elastic Registration

Elastic Registration: Thought Expt

- Single moving point, spring connected to its point of origin

tension in stretched
spring

in image 1



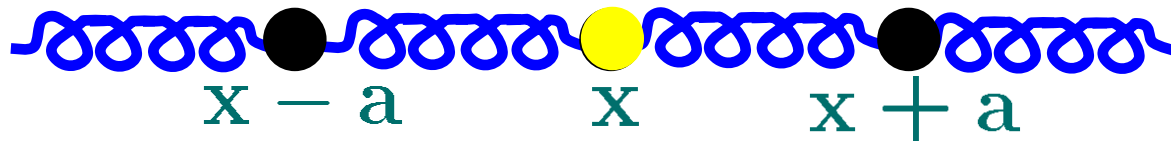
$$\underline{x} + \underline{u}(\underline{x})$$

in image 2

- Where does the force act?
- Push-forward & take image value with me:
- Spring force acts at $\underline{x} + \underline{u}(\underline{x})$ but depends on $\underline{u}(\underline{x})$
- Pull-back: bring value at $\underline{x} + \underline{u}(\underline{x})$ back with me
- Image difference computed at \underline{x} and spring force at \underline{x}
- Different coordinate systems can be confusing!

Elastic Registration: simple 1D Model

$$u(x - a) \longrightarrow \longrightarrow u(x) \longrightarrow u(x + a)$$



- Displace $x \mapsto \phi(x)$, $x \mapsto x + u(x)$
- Spring: force proportional to extension
- Uniform translation: no net force
- Scaling: no net force
- Net force depends on second-derivatives of $u(x)$, $\phi(x)$

$$\left. \begin{array}{l} \phi(x) = sx + t, \\ F(x) = 0 \end{array} \right\}$$

Navier-Lame/Navier-Cauchy Equation:

Second spatial derivatives, rotationally invariant, vector-valued:

$$\underline{F}_{\text{elas}}(\underline{x}) = \mu \vec{\nabla}^2 \underline{u}(\underline{x}) + (\lambda + \mu) \vec{\nabla} (\vec{\nabla} \cdot \underline{u}(\underline{x}))$$

Two free parameters

Elastic Registration:

- **Forces:**

$$\underline{F}_{\text{elas}}(\underline{x}) = \mu \vec{\nabla}^2 \underline{u}(\underline{x}) + (\lambda + \mu) \vec{\nabla} (\vec{\nabla} \cdot \underline{u}(\underline{x}))$$

- **Energies:**

$$\mathcal{E}_{\text{elas}} = \int d\underline{x} \left[\frac{\mu}{4} \sum_{i,j} (\partial_i u_j + \partial_j u_i)^2 + \frac{\lambda}{2} (\vec{\nabla} \cdot \underline{u})^2 \right]$$

shorthand notation $\partial_{\underline{x}} \equiv \frac{\partial}{\partial \underline{x}}$
 functional derivative $\underline{F}_{\text{elas}}(\underline{x}) = -\frac{\delta \mathcal{E}_{\text{elas}}}{\delta \underline{u}(\underline{x})}$

- **Elastic registration:**

general expression: $\mathcal{E} = \mathcal{D}(\mathcal{I}_s, \mathcal{I}_t; \phi) + \lambda \mathcal{R}(\phi)$

$$\phi(\underline{x}) = \underline{x} + \underline{u}(\underline{x})$$

warp & displacement

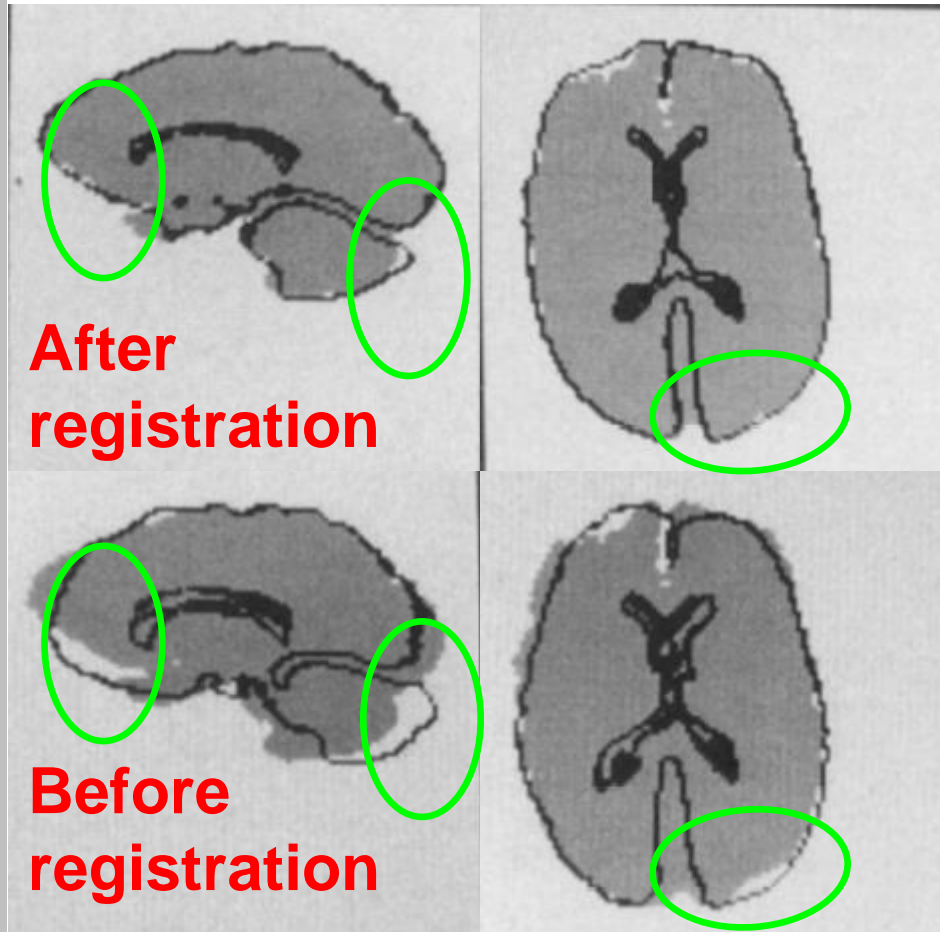
$$\mathcal{R}(\phi) = \mathcal{E}_{\text{elas}}(\underline{u})$$

elastic regulariser

Elastic Registration:

Bajcsy & Kovacic, *Multiresolution elastic matching*,
Computer Vision, Graphics, and Image Processing, Volume 46, (1989)

Atlas to Brain in 3D

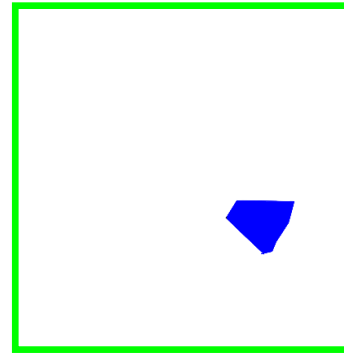
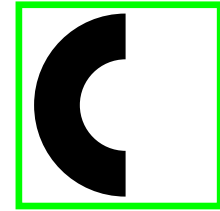
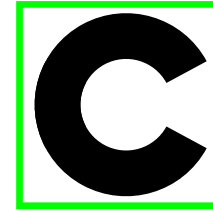
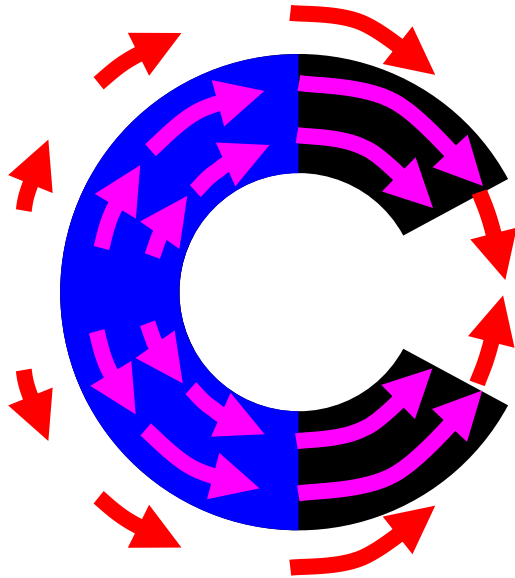


Rather old images, but
pure elastic:

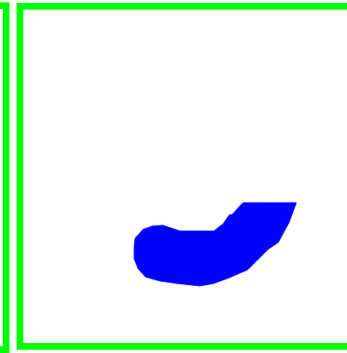
More recent papers use
more complicated
techniques

Elastic Registration

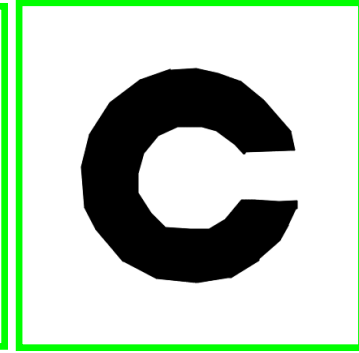
Direction of motion



moving
image



result



target

- Background pulled along as well & compressed
- Elastic force increases as deformation increases
- Will it go all the way?
- Christensen et al: Elastic fails for large deformations

Summary: Elastic Registration

- Potentially greater flexibility than parametric case
- Parametric case, choice of spline basis rather artificial
- Whereas elastic deformation accords with physical properties of actual objects
- Grossly simplified, elasticity same across whole image
- Elastic doesn't fold or tear provided small deformations
- Works well for some types of images
- But limited to small deformations

Can't handle extreme cases as arch to C shows

- Has problems when deformations only localized

Elasticity pulls rest of image along with it

Summary:

This Lecture:

- Warp **regularisation**
- **Parametric** warps
 - triangulated meshes, splines
- **Non-Parametric** Warps:
 - general framework,
 - trade-off between match and warp smoothness
 - Elastic** registration: formulation & problems

Next Lecture:

- **Fluids** and flows of warps, **groupwise** registration

Additional Information:

Bernd Fischer and Jan Modersitzki,
Ill-posed medicine—an introduction to image registration,
Inverse Problems 24 (2008)

Jan Modersitzki,
Numerical Methods for Image Registration,
Oxford Science Publications