Images, Geometric Hydrodynamics, and Shape Evolution

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Outline

1 Motivation

2 Building Geodesics Between Shapes

3 Numerical Model

4 Learning Heart Motion

5 Electrophysiology

A problem inspired by Image Registration

How similar/different are 2 images/shapes?

FACES









A problem inspired by Image Registration

How similar/different are 2 images/shapes?

Brain





A problem inspired by Image Registration

How similar/different are 2 images/shapes?

HEART





A problem inspired by Image Registration

How similar/different are 2 images/shapes?

Hypothalamus



Healthy



Schizophrenia

Assuming a correlation between healthy state or diseased, and the shape of the anatomical structure, Can we define a <u>distance</u> between these shapes?

A problem inspired by Image Registration:

Image Registration (Alignment): Estimation of **optimal transformation** between: Images, Points (Landmarks), Curves, Edges, Scalar functions, etc.

• Intensity	Cross-correlation, Mutual Information, Sum of Squared Intensity Differences, Ratio image uniformity.
• Models $\left\{ \right.$	Linear: Rigid, Affine, Piecewise Affine. Non-Rigid: Radial basis (thin plate, surface splines), Viscous fluids, Elastic, Large deformations.

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Building Geodesics: Background

D'Arcy Thompson: "On Growth and Form" (1917)

"To what extent is it true that individuals of closely related species can be transformed, one into the other, by a conformal transformation which carries every significant feature of one into the corresponding feature of the other?"



Building Geodesics: Background

- D'Arcy Thompson: "On Growth and Form" (1917).
- Ulf Grenander Mathematical rigor: Use of Lie group action.
- A. Trouve, An infinite dimensional group approach for physics based models in patterns recognition, International Journal of Computer Vision, 1995. (D. Ebin and J. Marsden, Groups of diffeomorphisms and the motion of an incompressible fluid, Ann. of Math, 1970.)
- Diffeomorphic matching framework, M. Miller, A. Trouve, L. Younes, On the metrics and Euler Lagrange equations of computational anatomy. Annual Review of Biomedical Engineering, 2002.
- M. Bruveris, F. Gay-Balmaz, D.D. Holm, T.S. Ratiu, *The momentum map representation of images*, Journal of Nonlinear Science, 2011. (Geometric Mechanics)

Building Geodesics: Generalized Euler Equations (V. Arnold)

- Lie group G with Lie algebra \mathfrak{g} , acting on a general manifold \mathcal{M} .
- (Right-Left) Invariant Lagrangian on $TG: \mathcal{L}: TG \to \mathbb{R}$.
- Euler-Poincaré reduction theorem, construct a reduced Lagrangian $l: u \in \mathfrak{g} \to \mathbb{R}$



• The kinetic \mathcal{L} defines a metric on G, so E - P represents a geodesic motion on G.

Generalized Euler Equations: SO(3)

Rotation of rigid bodies in $\mathbb{R}^3 \longrightarrow$ Geodesic motion in SO(3) Lie group:

L - P equation:

$$SO(3): \ \frac{d\Pi}{dt} - ad_{\Omega}^*\Pi = 0$$

where $ad_{\Omega}^{*}\Pi = \Pi \times \Omega$



Generalized Euler Equations: $Diff_{\mu}(\Omega)$

- Ideal fluid confined on $\Omega \subset \mathbb{R}^3$, configuration space $Diff_{\mu}(\Omega)$.
- Position of every particle $X \in \Omega$: $\varphi(t, X), \varphi \in Diff_{\mu}(\Omega)$.
- Right-Invariant Lagrangian on $TDiff_{\mu}(\Omega)$: $\mathcal{L}: TDiff_{\mu}(\Omega) \to \mathbb{R}$.
- With this setup, Ebin and Marsden proved well-possessedness for incompressible Euler's equations, starting the field of geometric hydrodynamics.

Generalized Euler Equations: $Diff(\Omega) : EPDiff$ equation

$$\partial_t \boldsymbol{m} + \boldsymbol{u} \cdot \nabla \boldsymbol{m} + (\nabla \boldsymbol{u})^T \cdot \boldsymbol{m} + \boldsymbol{m} (div \, \boldsymbol{u}) = 0$$

or
$$\partial_t \boldsymbol{m} + a d_u^* \boldsymbol{m} = 0$$

m: Momentum corresponding to the vector field u, ad^* : Coadjoint action of the Lie algebra \mathfrak{g} of $Diff(\Omega)$ on \mathfrak{g}^* .

- When $\Omega = S^1$, KdV equation: Geodesic flow with respect to the right invariant L^2 metric on the Bott-Virasoro group.
- When $\Omega = S^1$, Camassa-Holm (C-H) equation: Geodesic flow with respect to the right invariant H^1 metric on the Bott-Virasoro group. (Existence of geodesics has been proven for this PDE).
- In 1-D, EPDiff is:

$$m_t + u m_x + 2m u_x = 0 \qquad \text{with} \qquad m = (1 - \alpha^2 \partial_x^2) u, \tag{1}$$

i.e., the dispersionless limit of the C-H equation for shallow water motion, with **train of peakons** solution.

Generalized Euler Equations: $Diff(\Omega) : EPDiff$ equation



Generalized Euler Equations: $Diff(\Omega) : EPDiff$ equation



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Diffeomorphic Registration: Numerical Model Target



EPDiff (Geodesics on Diff) \Leftrightarrow Hamiltonian System

Diffeomorphic Registration: Numerical Model

Energy on the initial momentum: $E(\rho_0) = (\rho_0 | K\rho_0) + \lambda U(\varphi(1))$ Variation on the initial momentum: $\rho_0 \to \rho_0 + \delta \rho_0$

$$\delta E(\rho_0) = 2\left(\left.\delta\rho_0\right| K\rho_0\right) + \lambda \left(\left.\frac{\delta U}{\delta\varphi}(\varphi(1))\right| \delta\varphi(1)\right) \tag{1}$$

1) From *EPDiff* ... Linearized model:

$$\partial_t \left\{ \begin{array}{l} \delta\varphi\\ \delta\mu \end{array} \right\} = \mathcal{J}_{\varphi,\mu} \left\{ \begin{array}{l} \delta\varphi\\ \delta\mu \end{array} \right\} \qquad \delta\varphi(0) = 0, \delta\mu(0) = \delta\rho_0$$

2) Adjoint system: $\xi_{\varphi} := \delta \varphi^*, \, \xi_{\mu} := \delta \mu^*$

3) So (1) becomes

$$\delta E(\rho_0) = (\delta \rho_0 | 2K\rho_0 + \lambda \xi_\mu(0))$$
$$\Rightarrow \nabla E(\rho_0) = 2\rho_0 + \lambda K^{-1} \xi_\mu(0) \qquad \Rightarrow \boldsymbol{\rho_0}^{n+1} = \boldsymbol{\rho_0}^n - \varepsilon \, \nabla E(\rho_0^n)$$

Diffeomorphic Registration: Numerical Model



Target



Felipe Arrate. "Evolution Equations On The Group Of Diffeomorphisms, With Applications In Computational Anatomy".

Examples: Medical Images Registration





L. Younes, Felipe Arrate, M.I. Miller: "Evolution equations in computational anatomy". NeuroImage 45 (2009).

Examples: Miocardial Fibers



Vadakkumpadan, Felipe Arrate, T. Ratnanather, R. Winslow, May 2011.

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Problem: Cardiovascular diseases



GOAL

Classification of pathologic hearts by individualized model of cardiac motion and electrophysiology.

- Heart diseases are today responsible for the 28% of deaths in western countries (Cancer is 30%), around 400,000 deaths per year in Europe
 - fatigue
 - insufficient ventricular contraction
- Sudden Death affects 1 in 10,000 per year in developed countries
 - electrical disorder
 - ventricular fibrillation

- Diagnosis.
- Efficacy of defibrillation in infarcted hearts.
- Ablation location for correcting arrhythmias.

Problem: How can we classify pathological heart states?

1) Understand the physics and background on the problem

- Heart Mechanics involves a detailed knowledge of elastic constants at different levels inside the myocardium.
- Importance of trabecular fibers.
- Electrophysiology.
- Detailed description of boundary conditions.

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2) "Brainstorming"

- Is Aortic pressure enough? Can we accurately (to a good level) get those measurements?
- Attempts using electrocardiogram measurements have their own line of research, but rough approximations.
- Mechanics?. Difficult measurements. Proposed models have more or less detail on the tensor structure of the fibers, stress tensor, ...
- Motion Learning? → Images (cine CT, MRI). http://www.youtube.com/watch?v=lZZvPgquif4 http://www.youtube.com/watch?v=rhallml-juk

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3) Motion Learning

- What can we infer from a study mainly based on medical images?
- Is it possible to infer good/bad behaviour just from deformation?.
- Do we need to develop new Mathematics?.

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3) Motion Learning

• What do we follow, Tip of the heart? "Patterns of spiral tip motion in cardiac tissues", Kim, D.T. and Kwan, et al., Chaos, 8, 1, 1998



Problem: How can we classify pathological heart states?

3) Motion Learning: Start SIMPLE...

- 2D.
- Left ventricle.



Tracking Shape Deformation



Tracking Shape Deformation



Meier, D., D.D. Holm, F. Gay-Balmaz, F.X. Vialard, T. Ratiu, *Invariant higher-order variational problems*, Communications in Mathematical Physics, 2011.

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Motivation

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Maxwell's equations, but...

- Electrical field strengths are not too high \Rightarrow biological tissue is assumed to behave linear with regard to its electrical properties.
- Orardiac electrical activity is reflected by low frequency components only (≤ several kHz) ⇒ derivatives with respect to time can be neglected ('quasi-static' approximation of Maxwell's equations)

Motivation

Geodesics

Num. Model

Heart Motion

1 Electrophysiology



Maxwell:

$$\nabla \times E = -\frac{\partial B}{\partial t}$$

$$J = -DE$$
and...
$$\frac{\partial B}{\partial t} = 0 \Rightarrow E = -\nabla u$$



Monodomain Model: 1) $\frac{du}{dt} = c_1 f(u, w) + \nabla \cdot (D\nabla u)$ 2) $\frac{dw}{dt} = \epsilon(u - \gamma w)$ 3) $f(u, w) = c_1 u(u - \alpha)(u - 1) + c_2 uw$

FitzHugh-Nagumo Model

$$I_{ion} = f(u, w)$$
$$\frac{dw}{dt} = \epsilon(u - \gamma w)$$

Mesh or ... Meshless Method?

- Finite element methods have been extensively used for the spatial discretization of the myocardium.
- Complicated meshing procedures and element-based interpolation functions often result in algorithms which are either easy to implement, but numerically inaccurate, or accurate but labor-intensive
- The meshfree platform is more adaptive to different cardiac geometries and thus beneficial to individualized analysis.

AND...







Particle Methods

- Complicated volume meshing procedures are excluded.
- No re-meshing is needed for improving spatial accuracy when deformation occurs.

Moving Least Squares (MLS) Approximation

- **1.** $\{x_1(t), \ldots, x_N(t)\}$ nodes (particles) in $\Omega \subset \mathbb{R}^3$
- **2.** $\mathbf{p}^T(x) = [p_1(x), \dots, p_m(x)]$ polynomial basis.

Governing equation

Given:

- locations $x_i(t)$, for i = 1, ..., N,
- values $u(x_i, t)$, for i = 1, ..., N

Solve the associated ODE system

$$\frac{d}{dt} \begin{cases} u \\ w \end{cases} = \mathbf{\Phi}(u, w)$$

 $and \ obtain$

- new locations $x_i(t + \Delta t)$, for i = 1, ..., N,
- new values $u(x_i, t + \Delta t)$, for i = 1, ..., N

Moving Least Squares

• Approximate the solution by:

$$u(x) = \sum_{k=1}^{m} p_k(x) a_k(x)$$

• Minimizing the functional

$$\mathcal{J} = \sum_{i=1}^{N} w(x - x_i) \left[\mathbf{p}^T(x_i) \mathbf{a}(x) - u_i \right]^2$$

 $(w(x - x_i)$ weighing function with compact support)

• Solve the MLS problem

 $A(x)\mathbf{a}(x)=B(x)\mathbf{u}$

MLS Approximation: Monodomain Model

1. Monodomain model	3. Meshfree approximation
u - membrane potential	$\Phi = [\phi_1(x), \dots, \phi_N(x)]$ - shape function
w - recovery variable.	$u \sim \Phi \mathbf{u} \qquad w \sim \Phi \mathbf{w}$
$\frac{\partial u}{\partial t} = f(u, w) + \nabla \cdot (D\nabla u)$	$\left[\int_{\Omega} \Phi^{T} \Phi\right] \frac{\partial \mathbf{u}}{\partial t} = \left[\int_{\Omega} \Phi^{T} \Phi\right] f(\mathbf{u}, \mathbf{w})$
$\frac{\partial w}{\partial t} = \epsilon(u - \gamma w)$ $f(u, w) = c_1 u(u - \alpha)(u - 1) - c_2 u w$	$-\left[\int_{\Omega} abla \Phi^T D abla \Phi ight] \mathbf{u}$
	$\left[\int_{\Omega} \Phi^{T} \Phi\right] \frac{\partial \mathbf{w}}{\partial t} = \left[\int_{\Omega} \Phi^{T} \Phi\right] \epsilon(\mathbf{u} - \gamma \mathbf{w})$

2. Weak formulation

 ϕ - regular test function

$$\begin{split} &\int_{\Omega} \phi \frac{\partial u}{\partial t} = \int_{\Omega} \phi \, f(u,w) - \int_{\Omega} \nabla \phi^{T}(D\nabla u) \\ &\int_{\Omega} \phi \frac{\partial v}{\partial t} = \int_{\Omega} \phi \, \epsilon(u - \gamma w) \end{split}$$

4. ODE system

$$\frac{\partial \mathbf{u}}{\partial t} = f(\mathbf{u}, \mathbf{w}) + M^{-1} K \mathbf{u}$$
$$\frac{\partial \mathbf{w}}{\partial t} = \epsilon (\mathbf{u} - \gamma \mathbf{w})$$
$$f(\mathbf{u}, \mathbf{w}) = c_1 \mathbf{u} \circ (\mathbf{u} - \alpha) \circ (\mathbf{u} - 1) - c_2 \mathbf{u} \circ \mathbf{w}$$

where

$$M = \int_{\Omega} \Phi^T \Phi \qquad K = \int_{\Omega} \nabla \Phi^T D \nabla \Phi$$

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MLS Approximation: Monodomain Model

1. Monodomain model

u - membrane potential w - recovery variable.

$$\frac{\partial u}{\partial t} = f(u, w) + \nabla \cdot (D\nabla u)$$
$$\frac{\partial w}{\partial t} = \epsilon(u - \gamma w)$$
$$(u, w) = c_1 u(u - \alpha)(u - 1) - c_2 uw$$

2. Weak formulation

$$\begin{split} \phi &- \textit{regular test function} \\ & \int_{\Omega} \phi \frac{\partial u}{\partial t} = \int_{\Omega} \phi \, f(u,w) - \int_{\Omega} \nabla \phi^{T}(D\nabla u) \\ & \int_{\Omega} \phi \frac{\partial v}{\partial t} = \int_{\Omega} \phi \, \epsilon(u - \gamma w) \end{split}$$



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$$\frac{\partial u}{\partial t} = f(u, w) + \nabla \cdot (D\nabla u)$$
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$$f(u, w) = c_1 u(u - \alpha)(u - 1) - c_2 u w$$

$$\begin{split} & M \frac{\partial \mathbf{u}}{\partial t} + \left[\int_{\Omega} \Phi^T \left[(J\Phi) \dot{\mathbf{x}} \right]^T \right] \mathbf{u} = M f(\mathbf{u}, \mathbf{w}) - K \mathbf{u} \\ & M \frac{\partial \mathbf{w}}{\partial t} + \left[\int_{\Omega} \Phi^T \left[(J\Phi) \dot{\mathbf{x}} \right]^T \right] \mathbf{w} = M \epsilon (\mathbf{u} - \gamma \mathbf{w}) \end{split}$$

2. Weak formulation

$$\begin{split} \phi &- \textit{regular test function} \\ & \int_{\Omega} \phi \frac{\partial u}{\partial t} = \int_{\Omega} \phi \, f(u,w) - \int_{\Omega} \nabla \phi^{T}(D\nabla u) \\ & \int_{\Omega} \phi \frac{\partial v}{\partial t} = \int_{\Omega} \phi \, \epsilon(u - \gamma w) \end{split}$$

...Some preliminary results... fixed heart

Thanks for listening...