SHAPES AND DYNAMICS OF BIOLOGICAL SYSTEMS

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Biology = Quantitative Science





Acetylaminofluorene







Comparing (biological) shapes

Part I: Optimal diffeomorphism



Part II: Geodesics in shape space



Optimal diffeomorphims

We want to compare two surfaces by finding an *optimal diffeomorphism* between them.



If the surfaces have identical geometry then the optimal diffeomorphism is given by an isometry.

But what if they have different geometries?

What map is *closest* to being an isometry?

Optimal diffeomorphims

Optimal diffeomorphisms do more than give a distance. They also give a *correspondence*.



Diffeomorphic mapping

General maps between two surfaces deform lengths and angles

Isometries conserve lengths and angles.... but they are rarely appropriate

➤Conformal maps are the next best options, as they distort lengths but preserve angle

The Uniformization Theorem

Theorem [Poincaré, Koebe]

Any two genus-zero surfaces are conformally equivalent

Given any two shapes (with no holes), there is a map from one to the other that preserves angles.

The UC Davis Version...



The UC Davis Version...







 \tilde{g}

g

 $\tilde{g} = e^{2u}g$





Q

g': E -> R+ (i,j) -> l'_{ij}

g: E -> R+ (i,j) -> I_{ij} 8

Continuous:

$$\tilde{g} = e^{2u}g$$

Discrete:

$$l'_{ij} = e^{u(i)+u(j)}l_{ij}$$

Many algorithms exist:

- 1. Discrete Ricci Flow
- 2. Discrete Yamabe Flow
- 3. Conformal Mean Curvature Flow
- 4. Harmonic Maps
- 5. Finite Elements
- 6. Optimize a cost function
- 7. Discrete Differential Equation
- 8. Wilmore Flow
- 9. Circle Packings

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(Springborn et al, 2008)

- 7. Discrete Differential Equation
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Parametrizing a conformal map between two surfaces



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Discrete stretching energy F_1 F_2 u U

$$E_{sd}(f) = \sqrt{\sum_{(v,v')\in F_1} \frac{A_{vv'}}{3} \left(\frac{l(f(v), f(v'))}{l(v,v')} - 1 \right)^2} + \sqrt{\sum_{(u,u')\in F_1} \frac{A_{uu'}}{3} \left(\frac{l(f^{-1}(u), f^{-1}(u'))}{l(u,u')} - 1 \right)^2}$$

How round are proteins?



How round are proteins?



Dataset of proximal first metatarsals from 38 prosimian primates, and 23 New and Old World monkeys





Prosimian: lemur







Simian: Cape baboon (old world)



Simian: White eared titi (new world)





Comparing lower molars from primates



Comparing lower molars from primates:

A10-A13: same order, same family Q06 : same order, different family



A10 0.0



0.30

Comparing lower molars from primates:

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Comparing lower molars from primates



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Minimum Action Paths and Shape Similarity

1. Defining a (geodesic) distance between shapes

2. Applications to simple 2D potentials

3. Applications to proteins: a simplified potential

4. Applications to large shapes: more simplifications

Distance between Shapes...



$$D = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$



 M_2

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$$D = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$

$$L = \sqrt{\left(\frac{dX}{dt}\right)^2}$$
$$\frac{\partial L}{\partial X} = \frac{d}{dt} \left(\frac{\partial L}{\partial \dot{X}}\right)$$
$$M_1$$
$$S = \int_0^T L dt = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$



Diffusive form of Langevin Equation:

$$\eta M \frac{dX}{dt} = -\nabla U(X) + B$$

η: friction M: diagonal mass matrix U: potential energy B: random force



For a trajectory

$$X(0) \to X(t) \to X_f(T_f)$$

$$P(X(0) \to X_f(T_f)) \propto \exp\left(-\frac{S}{k_BT}\right)$$

where the action, S, is given by (Onsager and Machlup, 1953):

$$S = \frac{1}{2\eta} \int_0^{T_f} \left(\eta M \frac{dX}{dt} + \nabla U(X) \right)^2 dt$$

Corresponding Lagrangian:

$$L = \left(\eta M \frac{dX}{dt} + \nabla U(X)\right)^2$$

 $T_X M$

Let M be a (smooth) manifold and E a function from (M,TM) to $[0,\infty)$;



Let X_0 and X_F be two points on M. Then

$$d(X,Y) = \inf\left\{\int_{0}^{T} E(\gamma(t), \dot{\gamma}(t)) dt \mid \gamma \in C^{1}([0,T], M), \gamma(0) = X_{0}, \gamma(T) = X_{F}\right\}$$

defines an intrinsic quasi-metric on M.

There always exist length minimizing curves on (M,E). Such curves can always be reparametrized to be geodesics, and any geodesic must satisfy the Euler-Lagrange equation for $F(\gamma)$:

$$F(\gamma) = \int_{X_0}^{X_F} L(\gamma(t), \dot{\gamma}(t)) dt = \int_{X_0}^{X_F} \left[E(\gamma(t), \dot{\gamma}(t)) \right]^2 dt$$

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Onsager and Machlup (1953) action:

$$S = \frac{1}{2\eta} \int_0^{Tf} \left(\eta M \frac{dX}{dt} + \nabla U(X) \right)^2 dt$$

Corresponding Lagrangian:

$$L = \left(\eta M \frac{dX}{dt} + \nabla U(X)\right)^2$$

Lagrangian:
$$L = \left(\eta M \frac{dX}{dt} + \nabla U(X)\right)^2$$

Euler Lagrange equations:

$$\frac{\partial L}{\partial X} = \frac{d}{dt} \left(\frac{\partial L}{\frac{\partial L}{\partial X}} \right)$$

$$\frac{d^2 X}{dt^2} = \nabla \nabla U(X) \nabla U(X)$$

Boundary conditions:

$$X(0) = X_0 \qquad X(T_f) = X_f$$

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Elastic network for biomolecules:



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Elastic potential:

$$V_{ENM}(X) = \frac{1}{2} \sum_{i=1}^{N} \sum_{j>i} k_{ij} (r_{ij} - r_{ij}^{0})^{2}$$

. .

2nd order Taylor expansion:

 $V_{ENM}(X) \approx V_{ENM}(X^0) + \nabla V_{ENM}(X^0)^T (X - X^0) + \frac{1}{2}(X - X^0)^T H(X - X^0)$

$$V_{ENM}(X) \approx \frac{1}{2} (X - X^0)^T H (X - X^0)$$

Mixing potential for transition path:



Mixing potential:

$$U(X) = -\log(e^{-V_0(X)} + e^{-V_F(X)})$$

Solve:

 $\frac{d^2 X}{dt^2} = \nabla \nabla U(X) \nabla U(X)$

with boundary conditions:

$$X(0) = X_0 \qquad X(T_f) = X_f$$

using a relaxation method.

Transition Path for a Ribonuclease III





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An elastic model for shapes

$$U_{1}(X) = \sum_{edges(i,j)} (d_{ij}(X) - d_{ij}(X_{0}))^{2}$$

$$U(X) \approx E(X_{0}) + \nabla E(X_{0})^{T} + \frac{1}{2} (X(t) - X_{0})^{T} H(X_{0}) (X(t) - X_{0})$$

$$= \frac{1}{2} (X(t) - X_{0})^{T} H(X_{0}) (X(t) - X_{0})$$

Mixing potential for transition path:



Mixing potential:

$U(X) = \min(V_0(X), V_1(X))$

Euler-Lagrange equations for stationary action:

$$\begin{aligned} X(t \rightarrow t_0) &= X_t = X(t_0 \leftarrow t) \\ \dot{X}(t \rightarrow t_0) &= \dot{X}(t_0 \leftarrow t) \\ U(X(t \rightarrow t_0)) &= U(X(t_0 \leftarrow t)) \end{aligned}$$

Transitions between two states

$$X(t_0) = X_t$$

$$X(t) = X_0$$

$$X(t) = \sinh(Ht) \operatorname{csch}(Ht_0)(X_t - X_0) + X_0$$

$$X(t) = \sinh(P(t-T))\operatorname{csch}(P(T-t_0))(X_f - X_t) + X_f$$





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Trained morphometrist



Variational distance



Thank You



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