Geometric Flows in Python

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- Review of Plane Curve Geometry
- Ourve Shortening Flow
- In Python
- Review of Surface Geometry
- Gauss Curvature Flows

Curve shortening flow is a plane curve theory so we review the basics

Definitions

- A Immersed Planar Curve is a smooth map X : I → ℝ² satisfying the condition |X'(u)| ≠ 0 for u ∈ I
- The Unit Tangent Vector T is the derivative of an immersed planar curve with respect to an arclength parameter given by $T = \frac{dX}{ds} = \frac{X'}{|X'|}$
- Rotating the unit tangent vector counter clockwise by $\frac{\pi}{2}$ gives us the Unit Normal Vector ${\bf N}$

Review of Plane Curve Geometry



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Differentiating N and T gives us our last invariant, Curvature.

Definition

The *curvature*, κ of a plane curve is defined by the *Frenet-Serret* equations:

$$\frac{d\mathbf{N}}{ds} = \kappa \mathbf{T}$$
$$\frac{d\mathbf{T}}{ds} = -\kappa \mathbf{N}$$

Intuitively, this tells us how much the unit normal changes

Note that since ${\bf T}$ is a unit vector, we can describe it via it's turning angle, ϕ

$$\mathbf{T} = (cos\phi, sin\phi)$$

so, differentiating yields the following

$$rac{d\mathbf{T}}{ds} = rac{d\phi}{ds}(-sin\phi, cos\phi)$$

via Frenet-Serret, we get the nice relation

$$\kappa = \frac{d\phi}{ds}$$

Now, let $X : M^1 \times [0, T) \to \mathbb{R}^2$ be a smooth map which is an immersion for each $t \in [0, T)$

We say X(t) is a solution to the *Curve-Shortening Flow* if

$$\frac{\partial \mathbf{X}}{\partial t}(u,t) = -\kappa(u,t)\mathbf{N}(u,t)$$

or, via the Frenet-Serret equations:

$$\frac{\partial \mathbf{X}}{\partial t}(u,t) = \frac{\partial^2 \mathbf{X}}{\partial s^2}(u,t)$$

Now for some results

Theorem

suppose $X_i : M_i^1 \times [0, T) \to \mathbb{R}^2$ be solutions to the curve shortening flow satisfying $X_1(M_1^1, 0) \cap X_2(M_2^1, 0) = \emptyset$ then $X_1(M_1^1, t) \cap X_2(M_2^1, t) = \emptyset$ for each $t \in [0, T)$

Theorem

let $X_0 : M^1 \to \mathbb{R}^2$ be a smooth embedding of M^1 then the solution to curve shortening flow with initial data X_0 exists on a maximal time interval [0, T) and converges to a point in \mathbb{R}^2

Famously, curve shortening flow was used in a modified version of Hamilton's program by Perelman to prove the Poincare Conjecture in 2003

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Goal: Implement curve shortening flow in Python Steps:

- Approximate a curve with finite number of points
- Ompute invariants at each point
- Use Euler's method to integrate the curve shortening flow equation



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We want to generalize the idea of curvature flow to surfaces, so we review some surface geometry

Definitions

- let M^n be a smooth n-dimensional manifold and let $X : M^n \to \mathbb{R}^{n+1}$ be a smooth immersion. Then we regard the image $M := X(M^n)$ to be a hypersurface in \mathbb{R}^{n+1}
- we define the *Gauss map* to be the map $G : M \to S^n$ taking points $p \in M$ to it's corresponding unit normal vector in $N \in S^n$

Review of Surface Geometry



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Differentiating the Gauss map at a point, gives us the *Weingarten map* given by the linear map

$$-W_{p,M} := D_pG : T_pM \to T_{G(p)}S^2$$

Intuitively, this tells us how the unit normal changes at a point and gives rise to a curvature for hyersurfaces

Definition

The Gaussian curvature of a hypersurface M at the point $p \in M$ is given by $K := det(W_{p,M})$

With this we introduce the Gauss curvature flow

A smooth 1-parameter family $\mathbf{X} : M^n \times I \to \mathbb{R}^{n+1}$ satisfies the α -Gauss curvature flow if it satisfies

 $\partial_t \mathbf{X} = -sign(\alpha) K^{\alpha} \mathbf{N}$

Where K is the Gauss curvature of the family.

When $\alpha = 1$, we call this flow the *Gauss curvature flow*, which is a natural generalization of curve shortening flow

For uniformly convex solutions, in particular, we get a nice theorem regarding convergence

Theorem

Given $\alpha > 0$ and a smooth uniformly convex hypersurface M of \mathbb{R}^{n+1} , then there exists a unique smooth solution $\{M_t\}_{t \in [0,T)}$ with initial data Mwhich remains uniformly convex on [0,T) and shrinks to a point as $t \to T$

However, this does not ensure convergence to a round point!

An improtant integral quantity associated with a hypersurface is the *Gaussian Entropy* given by

$$E_{\alpha}(M) = \left(\frac{Vol(M)}{|B^{n+1}|}\right)^{\frac{n}{n+1}} \left(\frac{1}{|S^n|} \int_M K^{\alpha} d\mu\right)^{\frac{1}{\alpha-1}}$$

Which characterizes regularity of M

an interesting case arises when $\alpha = \frac{1}{n+2}$

Theorem

Let $X : M \times [0, T) \to \mathbb{R}^n + 1$ be a smooth uniformly convex solution to the $\frac{1}{n+2}$ -Gauss curvature flow. Given $L \in SL(n+1, \mathbb{R})$, there is a family of diffeomorphisms ϕ such that $\tilde{X}(x, t) = L(X(\phi(x, t), t))$ is also a solution to the flow

In other words, the $\frac{1}{n+2}$ flow is affine invariant

We're interested in the *self-similar shrinking solutions* to the affine normal flow.

Proposition

let M be a smooth uniformly convex hypersurface. Then M is a critical point of $E_{\frac{1}{n+2}}$ iff M is a self-similar solution to the affine normal flow

Proposition

Any smooth uniformly convex hypersurface which is a critical point of $E_{\frac{1}{n+2}}$ is an ellipsoid

Piecing together the last two propositions we get a nice theorem

Theorem

The self-similar solutions to the affine normal flow are the ellipsoids which are critical points of $E_{\frac{1}{n+2}}$

With this condition we can discern whether the affine normal flow convergence to a round point!

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