Polygonal Approximation of Flows

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Multivalued maps on grids

Definition [Mrozek]: A grid on a compact space X is a finite collection of subsets \mathcal{G} for which

- $\bigcup_{G \in \mathcal{G}} G = X,$
- G = cl(int(G)),
- $G \cap \operatorname{int}(H) = \emptyset$ if $G \neq H$.

A multivalued map \mathcal{F} on G maps a point $G \in \mathcal{G}$ to a set of elements of \mathcal{G} (and hence maps sets to sets), and we will write $\mathcal{F}: \mathcal{G} \rightrightarrows \mathcal{G}$. The geometric realizations of a subset of $\mathcal{S} \subset \mathcal{G}$ and the map \mathcal{F} will be denoted by

$$|\mathcal{S}| = \bigcup_{G \in \mathcal{S}} G$$
 and $F(\mathcal{S}) = |\mathcal{F}(\mathcal{S})|.$

Outer approximations

A multivalued map \mathcal{F} is an **outer approximation** to the dynamical system $\varphi : T \times X \to X$ if for some k > 0

 $\varphi((0,1],G) \subset int(F^{[0,k]}(G)) \text{ and } \varphi([-1,0),G) \subset int(F^{[-k,0]}(G))$

where $\mathcal{F}^{[a,b]}(\mathcal{S}) = \bigcup_{n \in [a,b]} \mathcal{F}^n(\mathcal{S}).$

In the case of a map $f : X \to X$, there is a natural (minimal) outer approximation on any grid \mathcal{G} , namely $H \in \mathcal{F}(G)$ iff $f(G) \cap H \neq \emptyset$.

For flows one can of course use this approximation on the time- τ map to obtain an outer approximation. In practice, this may not be desirable in certain situations.

Dynamics of (combinatorial) multivalued maps

Definition: A set $\mathcal{A} \in \mathcal{G}$ is an **attractor** if $\mathcal{F}(\mathcal{A}) = \mathcal{A}$ and \mathcal{B} is a **repeller** if $\mathcal{F}^{-1}(\mathcal{B}) = \mathcal{B}$. If $\mathcal{S} = \mathcal{A} \cap \mathcal{B}$ for some attractor \mathcal{A} and some repeller \mathcal{B} , then \mathcal{S} is a **Morse set**.

Theorem [K., Mischaikow, VanderVorst]: If \mathcal{F} is an outer approximation of φ and S is a Morse set of \mathcal{F} , then |S| is an isolating block for a Morse set of φ .

Key fact: The combinatorics of finding Morse sets is easy. The complexity of the computation of the finest Morse decomposition of \mathcal{F} is **linear** in the number of elements of \mathcal{G} plus the number of images of \mathcal{F} .

Recurrence

Definition: The recurrent set is

$$\mathcal{R}(\mathcal{G}) = \{ G : G \in \mathcal{F}^n(G) \text{ for some } n > 0 \}.$$

On $\mathcal{R}(\mathcal{G})$, $G \sim H$ if $\exists m, n > 0$ such that $G \in F^m(H)$ and $H \in F^n(G)$.

Equivalence classes of \sim are the **recurrent components** of \mathcal{F} .

The recurrent components are the finest Morse decomposition for \mathcal{F} , i.e. the intersection of all attractor/repeller pairs.

Lyapunov functions

The graph algorithms also yield a discrete 'Lyapunov function' L for the graph, i.e. L(G) = L(H) iff $G \sim H$ and L(G) > L(H) whenever $H \in \mathcal{F}(G)$.

This function obviously does not contain any topological information. However, a function V can be constructed which is piecewise constant on the grid and approximates a true Lyapunov function for the dynamics of φ , under certain conditions, [K., Mischaikow, and VanderVorst; Ban].



Flows and polygons [Eidenschink and Mischaikow]

Given:

- $\Omega \in \mathbb{R}^d$ a polygonal (rectangular) region
- $f:\Omega \to \mathbb{R}^d$ a vector field on Ω

Find:

- ${\mathcal P}$ a polygonal decomposition of a full finite simplicial complex ${\mathcal K}$ with $|{\mathcal K}|=\Omega$
- $F: \mathcal{P} \rightrightarrows \mathcal{P}$ a multivalued map which is an outer approximation to the flow of $\dot{x} = f(x)$ on Ω

Definition: A simplex $L \in \mathcal{K}^{d-1}$ is a **flow transverse face** if $\nu(L) \cdot f(x) \neq 0$ for all $x \in L$ where $\nu(L)$ is normal to L. A polygon $P \in \mathcal{P}$ is a **flow transverse polygon** if every face $L \in \partial P$ is flow transverse.

Definition: Given a simplicial complex \mathcal{K} the **minimal flow transverse polygonal decomposition** \mathcal{P} of Ω consists of the equivalence classes of ~ defined by $K_1 \sim K_2$ in \mathcal{K}^d if $K_1 \cap K_2 =$ $L \in \mathcal{K}^{d-1}$ and $\nu(L) \cdot f(x) = 0$ for some $x \in L$ and extended by transitivity.



Flow induced multivalued map

Definition: The flow induced multivalued map $\mathcal{F} : \mathcal{P} \rightrightarrows \mathcal{P}$ is defined by:

- $P \in \mathcal{F}(P)$ iff P contains an equilibrium point (necessarily in its interior)
- $Q \in \mathcal{F}(P)$ if $P \cap Q = L \in \mathcal{K}^{d-1}$ and $\nu_P(L) \cdot f(x) > 0$ for $x \in L$ where $\nu_P(L)$ is normal to L pointing out of P, i.e. $Q \in F(P)$ if Q is *adjacent* to P along an **exit face** L of P.



Isolating blocks

A compact set $N \subset \Omega$ is an **isolating block** if every point on $\partial |S|$ immediately leaves |S| in either forward or backward time. Then an exit set for N can be taken as a subset of ∂N for computing the Conley index.

Theorem: If S is a Morse set of $\mathcal{F} : \mathcal{P} \rightrightarrows \mathcal{P}$ and either $|S| \cap \Omega = \emptyset$ or Ω is an isolating block, then |S| is an isolating block.

Essentially, \mathcal{F} is an outer approximation of the flow on the interior grid elements (after suitable rescaling of time).

In fact, $\varphi(t,x) \in int(|\mathcal{F}^{\omega}(P)|)$ for any P containing x as long as $\varphi([0,t],x) \subset \Omega$.

Note: to be an isolating block a polygonal set must not only be flow transverse on its boundary (automatic), but it must also have no **re-entrant points** on its boundary.



Recap 1

[1] We have a license to start computing. Any simplicial complex will yield a multivalued map which can be made rigorous.

[2] The results of the computation can be arbitrarily bad. However it will not give a wrong result. If a Morse set is found which contains no invariant set for the flow, it will have trivial Conley index.

[3] The advantage of using the vector field directly is avoiding problems which can arise in approximating the time- τ map on a cubical grid: how long to integrate, large errors/ small cubes, lack of transversality...

[4] The disadvantage is of course the geometry of simplicial complexes can be complicated.

Outstanding issues 1

[1] How closely can we approximate the dynamics of the flow with this type of multivalued map?

[2] What (computable) properties of the base simplicial complex are sufficient to guarantee a close approximation?

[3] Does such an approximating complex always exist?

[4] What are the algorithms to compute such a complex?

Parallel flow

Consider the flow on \mathbb{R}^d of $\dot{x} = \Pi = (1, 0, ..., 0)$.

Let \mathcal{K} be a full finite simplicial complex and $\Omega = |\mathcal{K}|$.

What type of triangulation gives the best approximation?

Intuitively, triangles should be long in the direction of the vector field and thin in directions orthogonal to it.





Delaunay triangulations

Definition: A *d*-simplex satisfies the **Delaunay property** if its open circumball contains no vertices. A simplicial complex in which every *d*-simplex satisfies the Delaunay property is a **Delaunay complex (triangulation).**

Theorem: Given any finite set $S \subset \mathbb{R}^d$, there exists a triangulation of the convex hull of S which is a Delaunay complex. Moreover, if no set of d+2 points in S lie on a d-1 sphere, the Delaunay complex of S is unique.

Remark: In \mathbb{R}^2 , the Delaunay triangulation maximizes the minimum angle over all possible triangulations.



Voronoi diagrams

Definition: Given a finite set $S \subset \mathbb{R}^d$, the locus of the set of all points in \mathbb{R}^d which are closer to $x_0 \in S$ than any other point in S is called the **Voronoi polygon at** x_0 . The collection of all Voronoi polygons is a polygonal decomposition of the plane called the **Voronoi diagram**.



Delaunay-Voronoi duality

Remark: The Delaunay triangulation is dual to the Voronoi diagram of a generic set of points.



Computing Delaunay triangulations

There are many methods for computing a Delaunay triangulation (or Voronoi diagram) especially in 2 dimensions:

- Delaunay (or Bowyer-Watson) (or Incremental) insertion
- Bistellar flip methods
- Polytope methods. . .

The Delaunay insertion method works in any dimension and is readily generalizable to our situation.

Delaunay insertion

Given \mathcal{K} the Delaunay triangulation of a finite set $S \subset \mathbb{R}^d$, a new point p in the convex hull of S can be inserted to compute the Delaunay triangulation of $S \cup \{p\}$ as follows:

[1] Delete all simplices of \mathcal{K} whose open circumball contains p.

[2] The union of all such simplices, called the **Delaunay kernel**, is a polygon which is star-shaped at p.

[3] Retriangulate the Delaunay kernel by simplices composed of a face of the boundary and p.

[4] The new complex is a Delaunay triangulation of $S \cup \{p\}$.



Anisotropic Delaunay triangulations

Implementation of Delaunay insertion requires only the circumscription test:

Does point p lie inside the circumball of simplex K?

Let ${\cal M}$ be a symmetric, positive definite matrix and

$$g_M(u,v) = \langle Mu,v \rangle$$

the corresponding bilinear form.

The circumballs in this metric are ellipsoids, and the circumscription test is just linear algebra. **Theorem:** If $S \subset \mathbb{R}^d$ is a finite set of points. Then generically there is a unique triangulation of the convex hull of S which satisfies the Delaunay property using circumballs in (\mathbb{R}^d, g_M) , which therefore can properly be called an (anisotropic) Delaunay complex in (\mathbb{R}^d, g_M) .

This constant metric situation will work for parallel flow with

$$M = \Lambda(\mu) = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & \mu^{-1} & 0 & \dots & 0 \\ \dots & & & & \\ 0 & 0 & 0 & \dots & \mu^{-1} \end{bmatrix}$$

but what about arbitrary flows?

Is there such a concept as a Delaunay complex on an arbitrary Riemannian manifold (X, g)? How can it be computed?

Riemannian-Delaunay triangulations

Theorem/Conjecture: Let (X,g) be a compact Riemannian manifold. Let $S \subset X$ be a finite set satisfying the density condition that every ball of sufficiently small radius (some fraction of the convexity radius) contains a point of S. Then there exists a triangulation of \mathcal{K} of (M,g) with $\mathcal{K}^0 = S$ and every $K \in \mathcal{K}^d$ has a (unique?) strongly convex (minimal radius?) circumball which contains no points of S. Generically, this Delaunay triangulation is unique.

Proved for 2-dimensional manifolds. [Libon]

Proof for any dimension?

Flow-induced Riemannian metric

Let \mathcal{E} be the equilibrium set in Ω .

Choose a vector field $f^{\perp}(x)$ orthogonal to f(x) on $\Omega \setminus \mathcal{E}$ and let $Dh(x) = [f(x)|f^{\perp}(x)].$

For $R(x) \neq 0$, define

$$M^f_{\mu}(x) = \frac{1}{R(x) \|f(x)\|^2} Dh(x) \cdot \Lambda(\mu) \cdot Dh^T(x).$$

Let $g^f_{\mu}(x)(u,v) = \langle M^f_{\mu}(x)u,v \rangle$ be the corresponding bilinear form.

Then $(\Omega \setminus \mathcal{E}, g^f_{\mu})$ is a Riemannian manifold.

Fix $M^f_{\mu}(x_0)$ at some x_0 and $\mu < 1$.

Then the unit ball in the metric determined by this constant matrix is an ellipsoid with:

- major axis in the direction of the vector field $f(x_0)$,
- the eccentricity determined by μ ,
- and the size determined by $R(x_0)$.

Remark: The function R(x) is related to the curvature of the vector field at x_0 .

Practical Riemannian-Delaunay triangulations

In practice a triangulation can be constructed from approximations by constant metrics.[Borouchaki, George, Hecht, Laug, and Saltel]

Using the constant metrics determined by p and the vertices of K, a star-shaped kernel can be constructed by modifying the circumscription condition to require that p lie inside one or all such circumballs or satisfy some average condition.

Example: Let

 $\alpha_p = d_p(p, O_p)/d_p(v, O_p)$ and $\alpha_v = d_v(p, O_v)/d_v(v, O_v)$ be the ratios of the distances from p to the circumcenter of Kto the circumradii of K measured in the constant metrics at pand at some vertex v of K.

Then K is in the Delaunay kernel of p if $\alpha_p + \alpha_v < 2$.



Recap 2

[1] Given any finite set of vertices S in $\Omega \setminus \mathcal{E}$ we can (in principle) compute a Riemannian-Delaunay triangulation of the convex hull of S which is aligned to the flow.

[2] The insertion algorithm is dimension independent and computational geometers have developed very fast algorithms.

[3] There are a number of computational issues:

- Early in the insertion, the metric approximation is bad.
- In dimensions higher than 2, the Delaunay kernel must be reduced to ensure that it is star-shaped with respect to the insertion point.
- If simplices get too thin, then multiple precision arithmetic may be needed.

Outstanding issues 2

[1] If the location of the vertices is bad, then the triangulation can be bad.

[2] Is there an algorithm to choose the vertices?

[3] How closely can we approximate the dynamics of the flow?

[4] Are there computable properties of the base simplicial complex are sufficient to guarantee a close approximation?

[5] Does such an approximating complex always exist/ can always be computed?

Centroidal Voronoi diagrams

Strategy: Place vertices so that each vertex is located at the centroid of its Voronoi region. [Du, Faber, Gunzberger, and Ju]





- Resulting elements are high quality, i.e. close to equilateral.
- There are naturally parallelizable algorithms involving probablistic averaging.

Overall procedure

[1] Choose a set of initial vertices randomly with respect to the curvature density R(x).

[2] Perform probablistic centroidal averaging using distances in the Riemannian metric to obtain a final set of vertices.

[3] Compute Riemannian-Delaunay triangulation.

[4] (Rigorously) compute the flow transverse polygonal decomposition and multivalued map.

[5] Apply graph algorithms to obtain rigorous isolating blocks.

[6] Extract isolating blocks and repeat with a smaller μ if necessary.

Reverse Van der Pol

$$\begin{aligned} \dot{x} &= -y \\ \dot{y} &= (x^2 - 1)y + x. \end{aligned}$$

The next figure will show the minimal flow transverse decomposition of a triangulation containing 20,000 vertices, 39,9940 triangles, and 27,552 polygons.



From this triangulation, we can extract the recurrent component the multivalued map which contains the periodic orbit. This set contains 8,093 triangles in 5,812 polygons.



This triangulation can be refined to approximate the periodic orbit more closely. The next figure shows a recurrent set for the multivalued map containing 20, 119 triangles.



3-d connecting orbit

$$\dot{x} = -x(x+1) - z$$

$$\dot{y} = y(2+6x-y) + 3\left(x+\frac{z}{3}\right)$$

$$\dot{z} = z(2-x+5y)$$

This 3-dimensional system, related to the ground state problem for a system of coupled semilinear Poisson equations with critical exponents, has a connecting orbit between the equilibria (0,0,0)and (-1,-1,0) as the intersection of a 2-dimensional unstable manifold and a 2-dimensional stable manifold. The connecting orbit is also proven in **[Hulshof and VanderVorst]** to be a parabola over the line x = y, z = 0. An isolating neighborhood of this connecting orbit with 12,326 simplices is shown in the next figure.



Current/ future work

• There are implementation, algorithmic, and parallelization issues to resolve.

• Automate the (human) choices to be made in performing any given computation: R(x), μ , # vertices, # averaging points, metric approximation...

• 4-d fast/slow system [Gameiro and Mischaikow]

• PDE bifurcation diagrams [Day, Hiraoka, Junge, Mischaikow, Ogawa, Wanner]

• Theoretical results

Flow oriented triangulations and chain recurrence

Definition: A simplex K is δ -oriented to the flow if every entrance face L has the following property:

 $\exists y \in L$ such that $v - y \in C(f(y), \delta) = \{w : w \cdot f(y) \ge \delta \|w\| \|f(y)\|\}$ where v is the vertex opposite L in K.

Theorem: For every $\epsilon > 0$, there exists $\delta > 0$ so that if \mathcal{K} is δ -oriented and diam(\mathcal{K}) is sufficiently small, then the recurrent set of the multivalued map $\mathcal{R}(\mathcal{P})$ is contained in $\mathcal{R}_{\epsilon}(|\mathcal{K}|, \varphi)$.