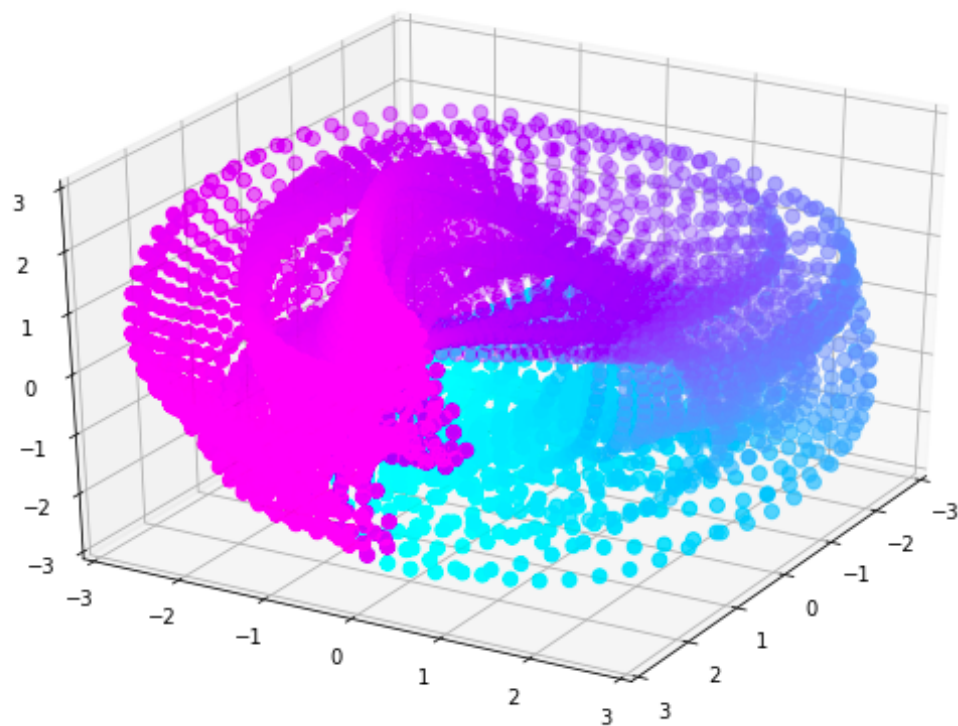


Tangent bundle estimation of immersed manifolds

Raphaël TINARRAGE

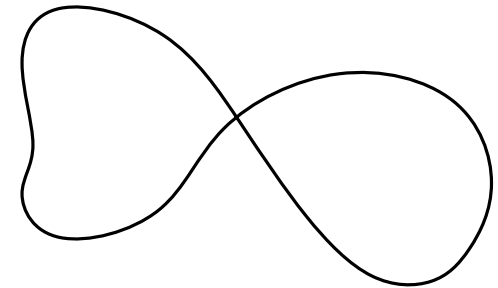
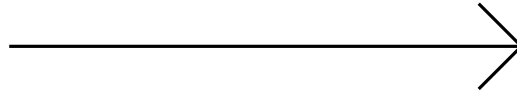
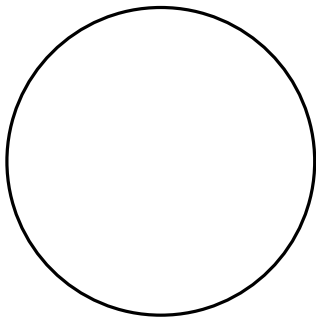


Introduction (1/3): Immersed manifolds

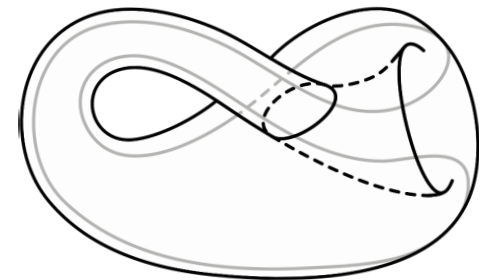
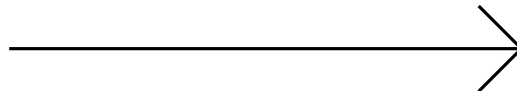
Abstract manifold \mathcal{M}_0 $\xrightarrow{\quad u \quad}$ $\mathcal{M} \subset \mathbb{R}^n$

Introduction (1/3): Immersed manifolds

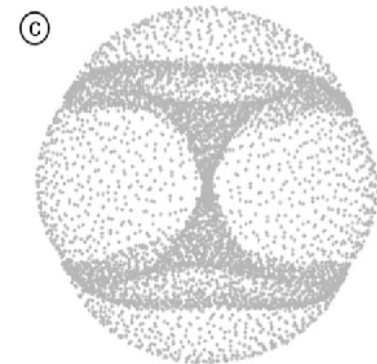
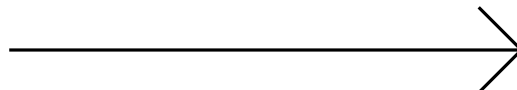
Abstract manifold \mathcal{M}_0 \xrightarrow{u} $\mathcal{M} \subset \mathbb{R}^n$



Klein bottle



Klein bottle $\cup \mathbb{S}_2$



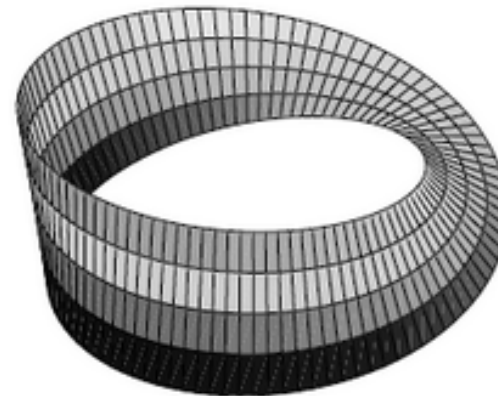
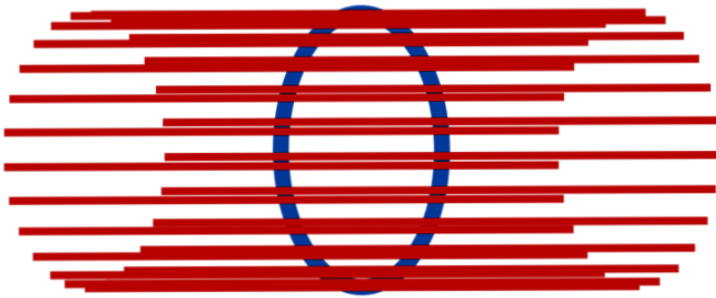
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Introduction (2/3): Vector bundles

Vector bundle over \mathcal{M}_0 :

it is a topological space E , with $p : E \rightarrow \mathcal{M}_0$ continuous surjective such that

- for all $x_0 \in \mathcal{M}_0$, $p^{-1}(\{x_0\})$ is given a structure of vector space, and $\forall x_0 \in \mathcal{M}_0, \exists U \subset \mathcal{M}_0$ neighborhood of x_0 , $\exists k \geq 0$, $\exists \phi : U \times \mathbb{R}^k \rightarrow p^{-1}(U)$ homeomorphism such that $\forall y_0 \in \mathcal{M}_0$,
- $v \mapsto \phi(y_0, v)$ is an isomorphism (of vector spaces)
- $\forall v \in \mathbb{R}^k, p \circ \phi(y_0, v) = y_0$

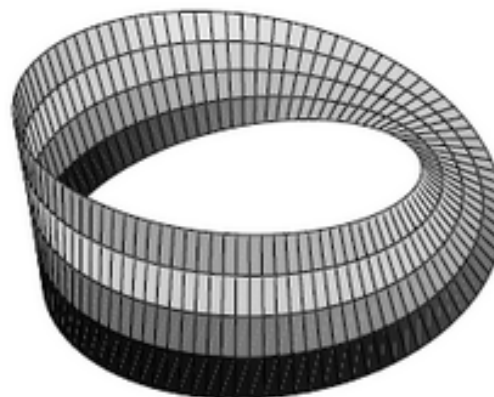
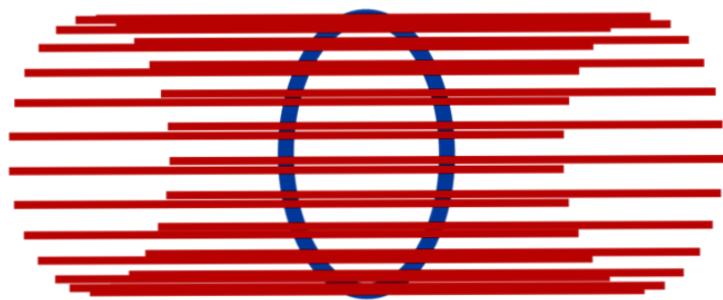


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Some vector bundles can be described by a map $\mathcal{M}_0 \rightarrow G_{d,n}$

Introduction (3/3): Tangent bundle

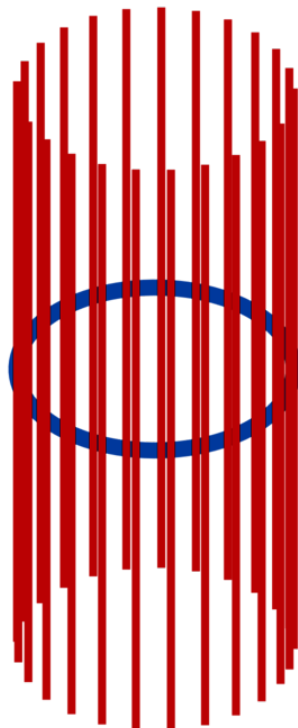
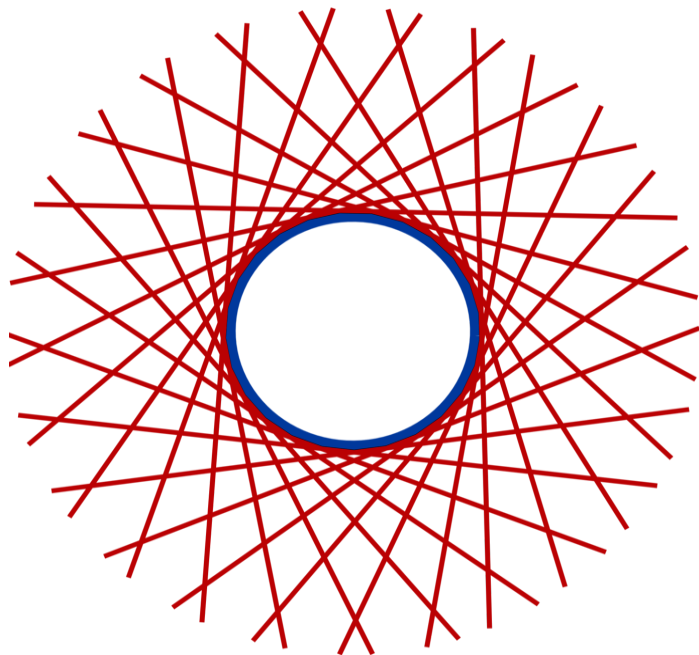
Tangent bundle:

(Definition for an abstract manifold): it is a vector bundle with total space the set $\bigcup_{x_0 \in \mathcal{M}_0} T_{x_0} \mathcal{M}_0$, endowed with a vector bundle structure...

(Definition for embedded manifold $\mathcal{M}_0 \rightarrow \mathcal{M}$): the tangent bundle is given by

$$T\mathcal{M} = \{(x, v), x \in \mathcal{M}, v \in T_x \mathcal{M}\}$$

and $p : (x, v) \in T\mathcal{M} \mapsto x \in \mathcal{M}$.



$T\mathcal{M}$ is a submanifold of dimension $2d$ of $\mathbb{R}^n \times \mathbb{R}^n$

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Another point of view:

Define $\check{\mathcal{M}} \subset \mathbb{R}^n \times G_{d,n}$ as

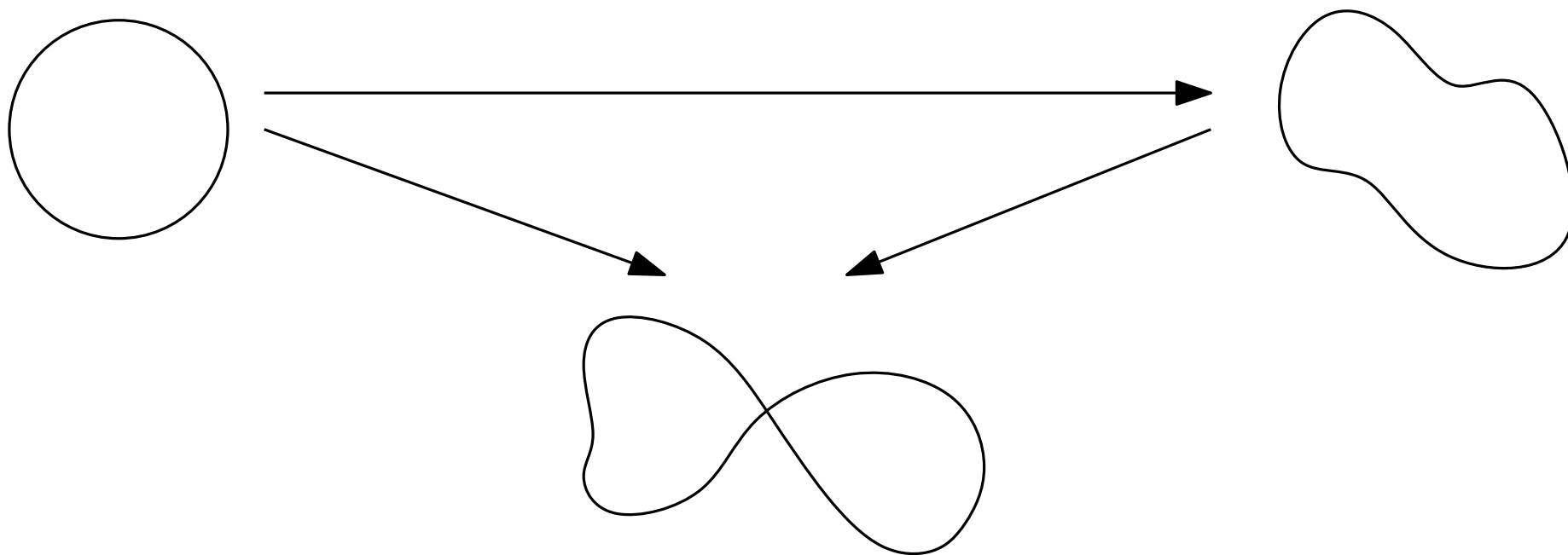
$$\check{\mathcal{M}} = \{(x, T_x \mathcal{M}), x_0 \in \mathcal{M}_0\}.$$

It is a submanifold of dimension d of $\mathbb{R}^n \times G_{d,n}$.

$\check{\mathcal{M}}$ can also be seen as a submanifold of $\mathbb{R}^n \times \mathbb{R}^{n^2}$.

Motivations (1/2): recover the diffeomorphism class of \mathcal{M}_0

$$\begin{array}{ccc} \mathcal{M}_0 & \xrightarrow{\check{u} : x_0 \mapsto (u(x_0), p_{T_x \mathcal{M}})} & \check{\mathcal{M}} \subset \mathbb{R}^n \times M_n(\mathbb{R}) \\ & \searrow u & \swarrow \pi : (x, T) \mapsto x \\ & \mathcal{M} \subset \mathbb{R}^n & \end{array}$$



Assumption: $\forall x_0, y_0 \in \mathcal{M}_0, T_{x_0} \mathcal{M} = T_{y_0} \mathcal{M} \implies x_0 = y_0$

\mathcal{M}_0 and $\check{\mathcal{M}}$ are diffeomorphic

Motivations (2/2): characteristic classes

The tangent bundle of \mathcal{M}_0 contains more precise information about the diffeomorphism class of \mathcal{M}_0 than just the cohomology ring $H^*(\mathcal{M}_0)$.

Example:

The Stiefel-Whitney class is a functor

$$\text{VectorBundles}_{\mathcal{M}_0} \longrightarrow H^*(\mathcal{M}_0, \mathbb{Z}_2)$$

If we consider the tangent bundle,

- the first Stiefel-Whitney class $w_1 \in H^1(\mathcal{M}_0, \mathbb{Z}_2)$ is zero iff \mathcal{M}_0 is orientable.
- the Stiefel-Whitney numbers are all zero iff \mathcal{M}_0 is the boundary of a compact manifold.

Model

Assumptions:

Differential geometric assumptions:

- \mathcal{M}_0 is a C^2 -manifold of dimension d
- $u : \mathcal{M}_0 \rightarrow \mathbb{R}^n$ is an immersion, $\mathcal{M} = u(\mathcal{M}_0)$
- $\forall x_0, y_0 \in \mathcal{M}_0, T_x \mathcal{M} = T_y \mathcal{M} \implies x_0 = y_0$

Riemannian geometric assumption:

- \mathcal{M}_0 is endowed (by pull-back) with a riemannian structure, and $\forall x_0 \in \mathcal{M}_0, ||II_{x_0}|| \leq \rho$

Measure assumption:

- μ_0 is a Radon measure on \mathcal{M}_0 with density f_0 which is L_0 -Lipschitz and bounded by $0 < f_{\min} \leq f_{\max}$

Model

Varifold:

Let $\check{u} : \mathcal{M}_0 \rightarrow \mathbb{R}^n \times M_n(\mathbb{R})$ be defined by $\check{u}(x_0) = (u(x_0), \frac{1}{d+2}p_{T_x}\mathcal{M})$.
Consider

$$\check{\mathcal{M}} = \check{u}(\mathcal{M}_0),$$

$$\check{\mu}_0 = \check{u}_*\mu_0.$$

Goal:

We observe a measure ν on \mathbb{R}^n such that $W_1(\mu, \nu) \leq \epsilon$.
Infer from ν the measure $\check{\mu}_0$.

Overview of the method

- For all $x \in \text{supp}(\nu)$, compute $\overline{\Sigma}_\nu(x)$ the normalized local covariance matrix.
- Consider the measure $\check{\nu} = \nu \otimes \delta_{\overline{\Sigma}_\nu(x)}$.
- Show that $W_1(\check{\nu}, \check{\mu}_0)$ small.
- Do persistent homology on $\check{\nu}$.

Reach (1/6)

Reach: Let $A \subset \mathbb{R}^n$, and $x \in \mathbb{R}^n \mapsto d(x, A) = \inf_{a \in A} \|x - a\|$ the distance function to A .

The medial axis of S is defined as

$$\text{med}(A) = \{x \in \mathbb{R}^n, \exists a, b \in A, \|x - a\| = \|x - b\| = d(x, A)\}$$

and the reach of A as

$$\text{reach}(A) = \inf_{a \in A} d(a, \text{med}(A)).$$

It is a scale at which A has good behaviour: with $r < \text{reach}(A)$ and $x \in A$,

- $A \cap \overline{\mathcal{B}}(x, r)$ is contractible
- the projection onto A is well defined on the neighborhood A^r
- tangent space estimation works [Arias-Castro et al, Spectral Clustering based on local PCA]

Reach (2/6)

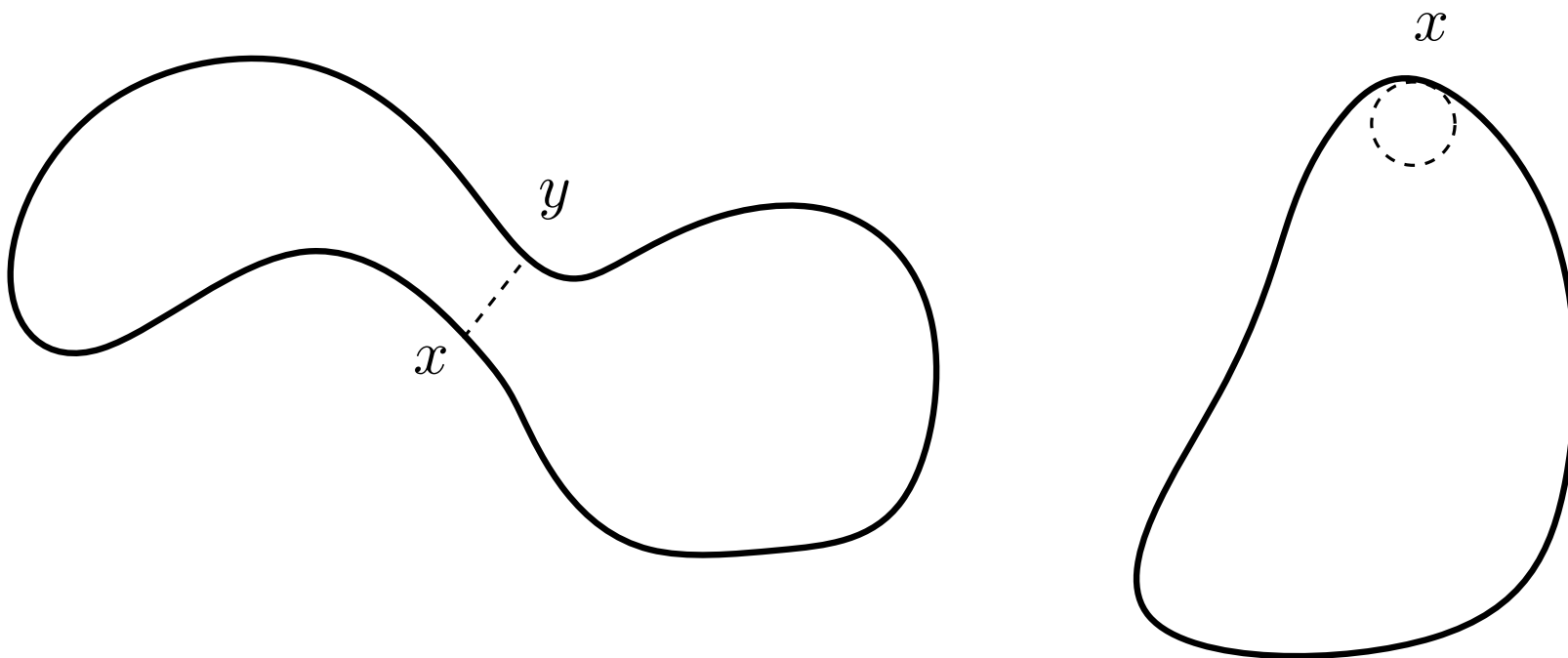
Theorem (Aamari):

Let M be a compact submanifold. Then

$$\text{reach}(M) = \frac{\lambda}{2} \wedge \frac{1}{\rho},$$

where

- $\lambda = \inf\{\|x - y\|, (x, y) \text{ bottleneck}\}$ (i.e. $\mathcal{B}(\frac{x+y}{2}, \frac{\|x-y\|}{2}) \cap \mathcal{M} = \emptyset$)
- $\rho = \sup_{x \in \mathcal{M}} \|II_x\|$

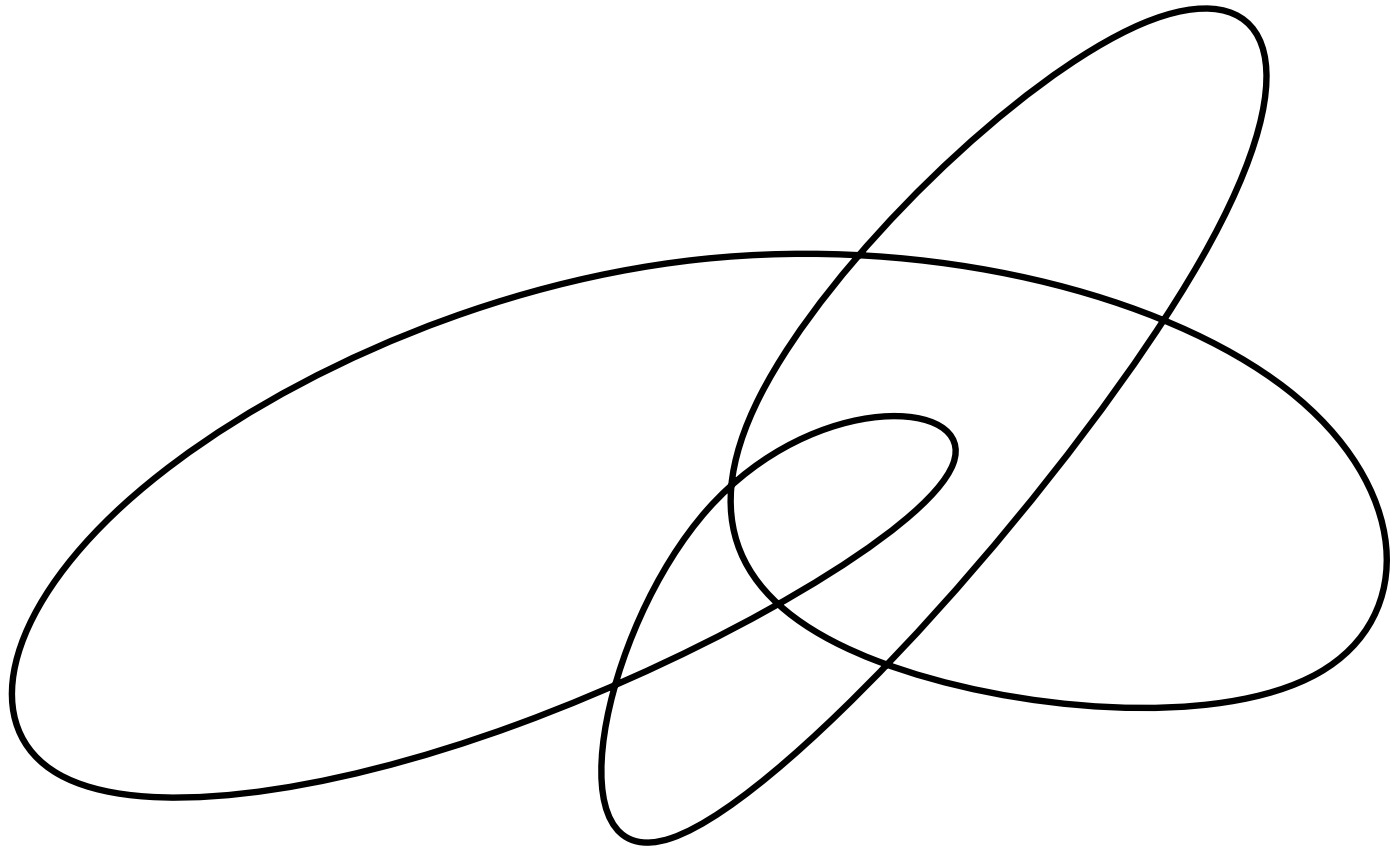


Reach (3/6)

Normal reach:

For every $x_0 \in \mathcal{M}_0$, we set

$$\lambda(x_0) = \inf\{\|x - y\|, y_0 \in \mathcal{M}_0, y_0 \neq x_0, x - y \perp T_y \mathcal{M}\}.$$

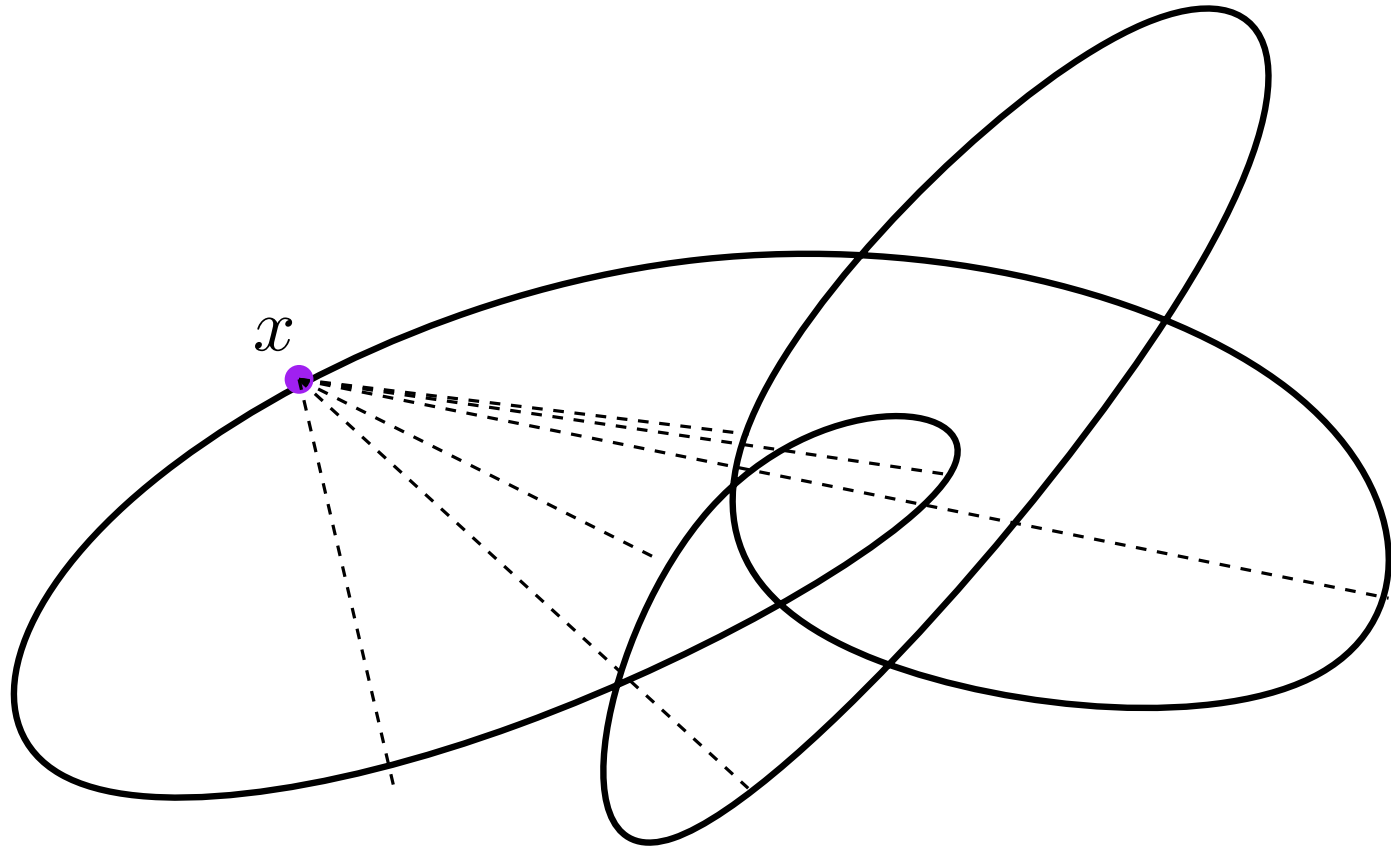


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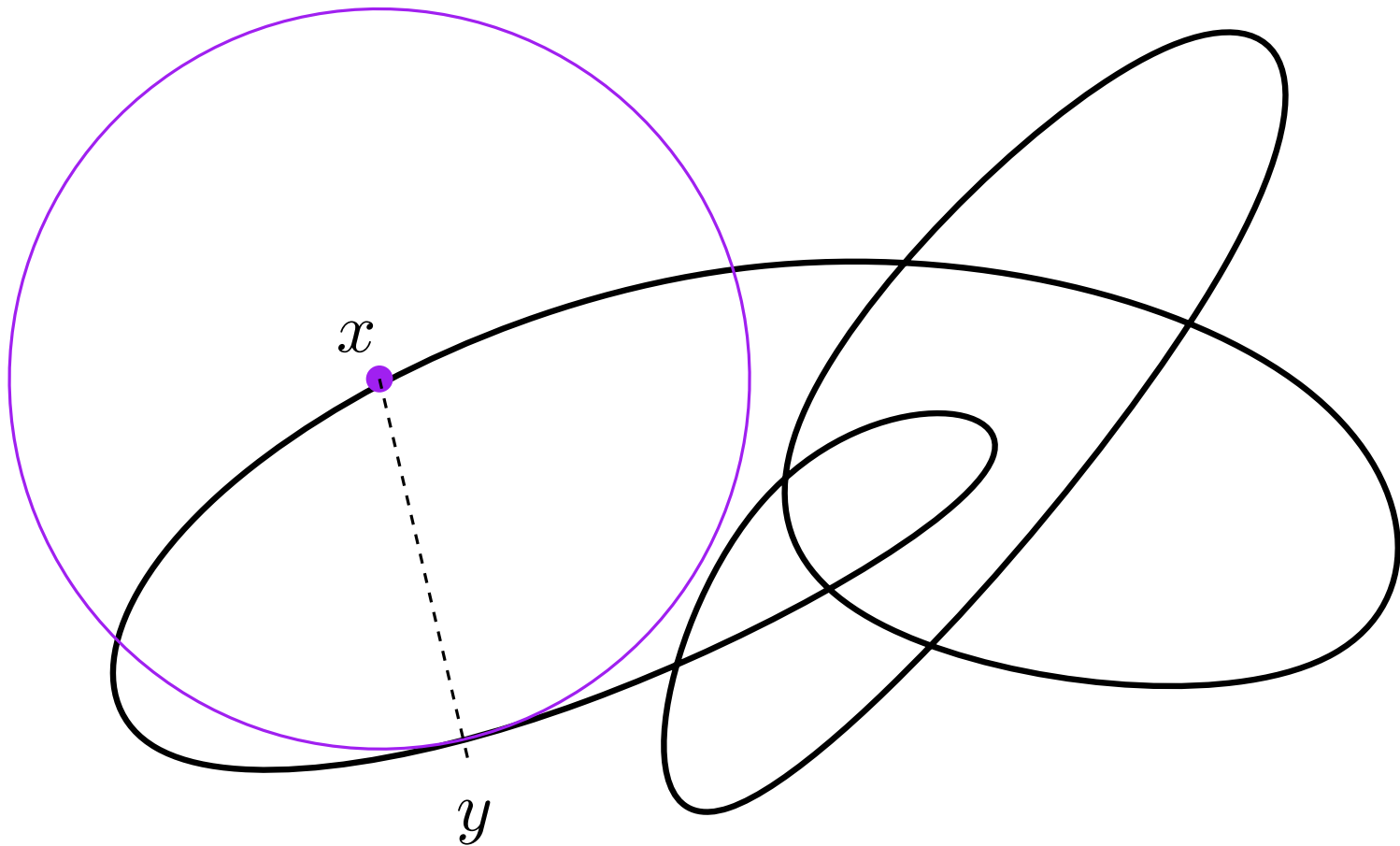


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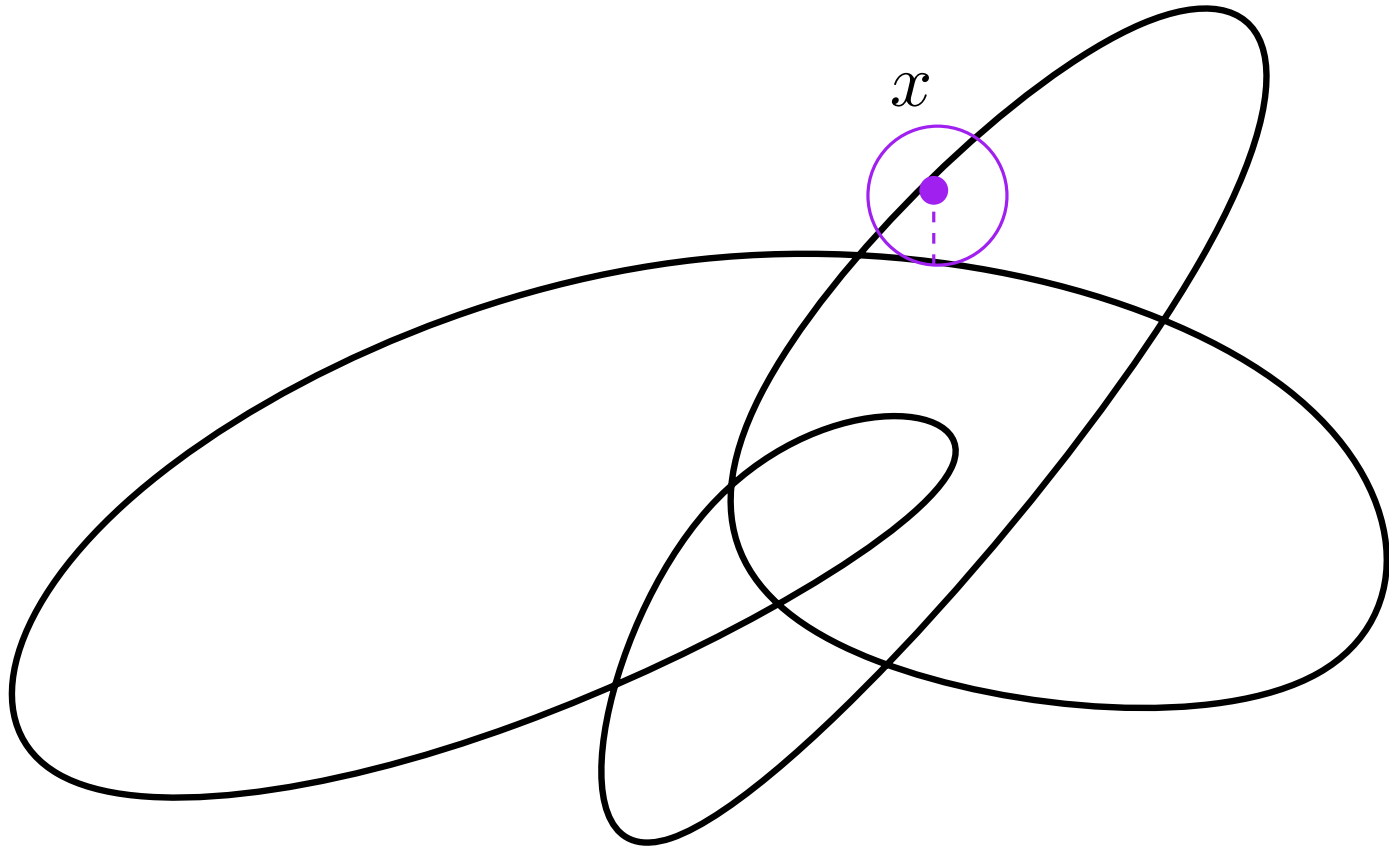


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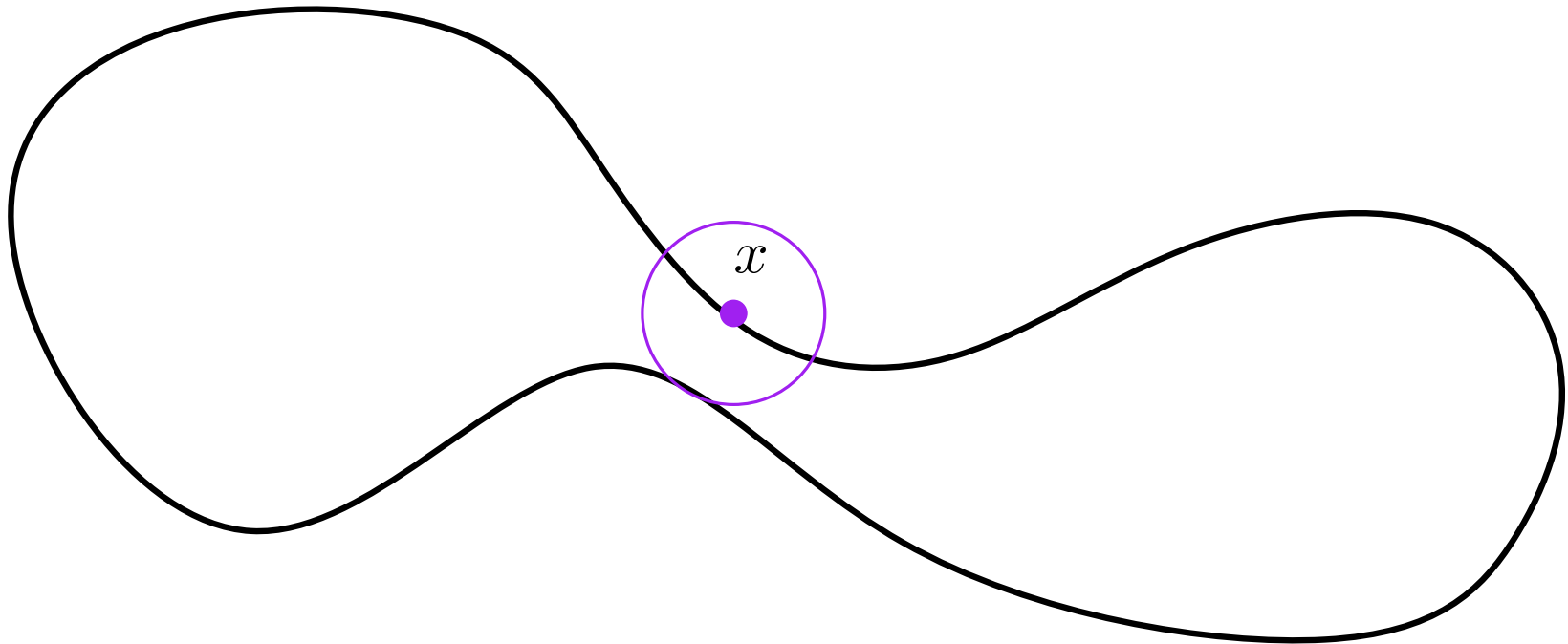
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Reach (4/6)

Proposition:

Let $x_0 \in \mathcal{M}_0$ and $r < \frac{1}{4\rho} \wedge \lambda(x_0)$. Then $\overline{\mathcal{B}}(x, r) \cap \mathcal{M}$ is a set of reach at least $\frac{1}{2\rho}$.



Reach (5/6)

Lemma:

Let $x_0, y_0 \in \mathcal{M}_0$. Denote $r = \|x - y\|$ and $\delta = d_{\mathcal{M}_0}(x_0, y_0)$. Suppose that $\|x - y\| < \frac{1}{2\rho} \wedge \lambda(x_0)$.

Then

$$\delta \leq c(\rho r)r \quad \text{where} \quad c(t) = \frac{1}{t}(1 - \sqrt{1 - 2t}).$$

Proof:

Show that $u^{-1}(\mathcal{M} \cap \mathcal{B}(x, r))$ is connected.

Show that $u^{-1}(\mathcal{M} \cap \mathcal{B}(x, r)) \cap \partial \mathcal{B}_{\mathcal{M}_0}(x_0, c(\rho r)r + \epsilon) = \emptyset$ for ϵ small enough, following [Niyogi et al., Finding the Homology of Submanifolds with High Confidence from Random Samples].

Reach (6/6)

Proposition:

Let $r \leq \frac{1}{4\rho} \wedge \lambda(x)$ and $x_0 \in \mathcal{M}_0$. We have

$$\mu(\mathcal{B}(x, r)) \geq ar^d$$

$$\left| \frac{\mu(\mathcal{B}(x, r))}{V_d r^d} - f(x) \right| \leq cr$$

Moreover, $s \geq 0$ is such that $s \leq r$, we also have

$$\mu(\mathcal{B}(x, r) \setminus \mathcal{B}(x, s)) \leq br^{d-1}(r - s)$$

(with $a = V_d(\frac{47}{48})^d f_{\min}$, $b = \frac{4d}{3}(\frac{17}{16})^d f_{\max}$, and
 $c = V_d(L + \frac{d}{4}f_{\max}\rho(\frac{17}{16})^{d-1} + 2df_{\max}\rho\frac{17}{16}(\frac{3}{2})^{d-1}).$)

Tangent space estimation (1/6)

Definition:

Let $r > 0$ and $x \in \text{supp}(\mu)$. The *local covariance matrix of μ around x at scale r* is the following matrix:

$$\Sigma_{\mu}(x) = \int_{\overline{\mathcal{B}}(x,r)} (x - y)^{\otimes 2} \frac{d\mu(y)}{\mu(\overline{\mathcal{B}}(x,r))}$$

and the *normalized local covariance matrix* is

$$\overline{\Sigma}_{\mu}(x) = \frac{1}{r^2} \Sigma_{\mu}(x).$$

Notation: for any $x \in \mathbb{R}^n$, $x^{\otimes 2} = x^t x \in M_n(\mathbb{R})$.

Tangent space estimation (2/6): Consistency

Proposition:

Let $x_0 \in \mathcal{M}_0$ and $r \leq \lambda(x_0) \wedge \frac{1}{4\rho}$. Denote $T = T_x \mathcal{M}$ and by p_T the linear projection on T . Then

$$\left\| \frac{\Sigma_\mu(x)}{r^2} - \frac{1}{d+2} p_T \right\|_F \leq cr$$

with $c = \left(\frac{8}{7}\right)^2 \rho + \frac{\text{lip}(f) J_{\max} + \frac{d}{4} \rho f_{\max}}{f_{\min} J_{\min}} + \frac{f_{\max} J_{\max}}{f_{\min} J_{\min}} d + \frac{C}{f_{\min} J_{\min}}$.

—————→ The tangent space at x is well estimated
provided that $r \leq \lambda(x_0)$

Tangent space estimation (3/6): Stability

Consider μ, ν close in Wasserstein distance.

The distance $\|\bar{\Sigma}_\mu(x) - \bar{\Sigma}_\nu(x)\|_F$ is defined only when $x \in \text{supp}(\mu) \cap \text{supp}(\nu)$.

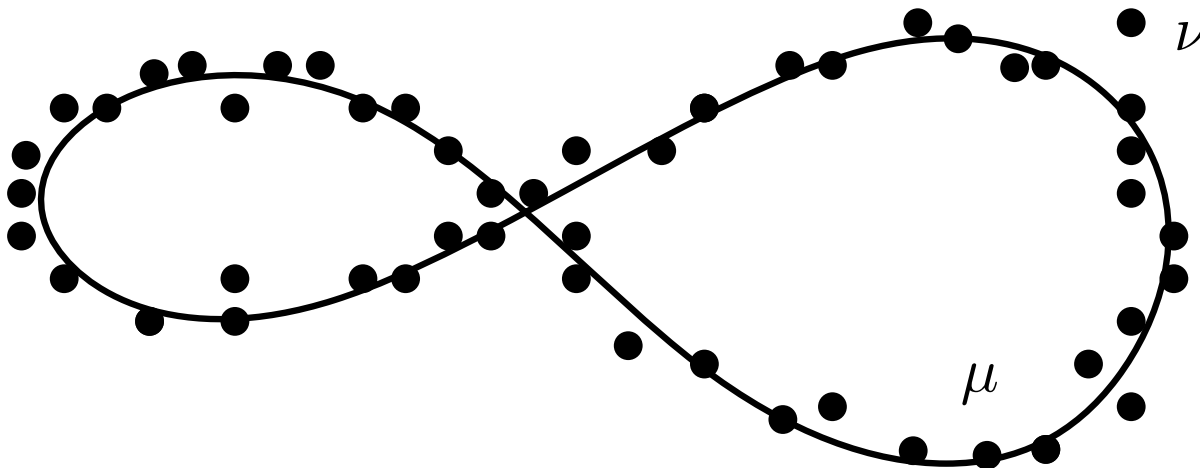
We will study the stability via the measures $\check{\mu}$ and $\check{\nu}$.

We define

$$\check{\mu} = \mu \otimes \delta_{\bar{\Sigma}_\mu(x)}$$

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We wish to bound $W_1(\check{\mu}, \check{\nu})$.



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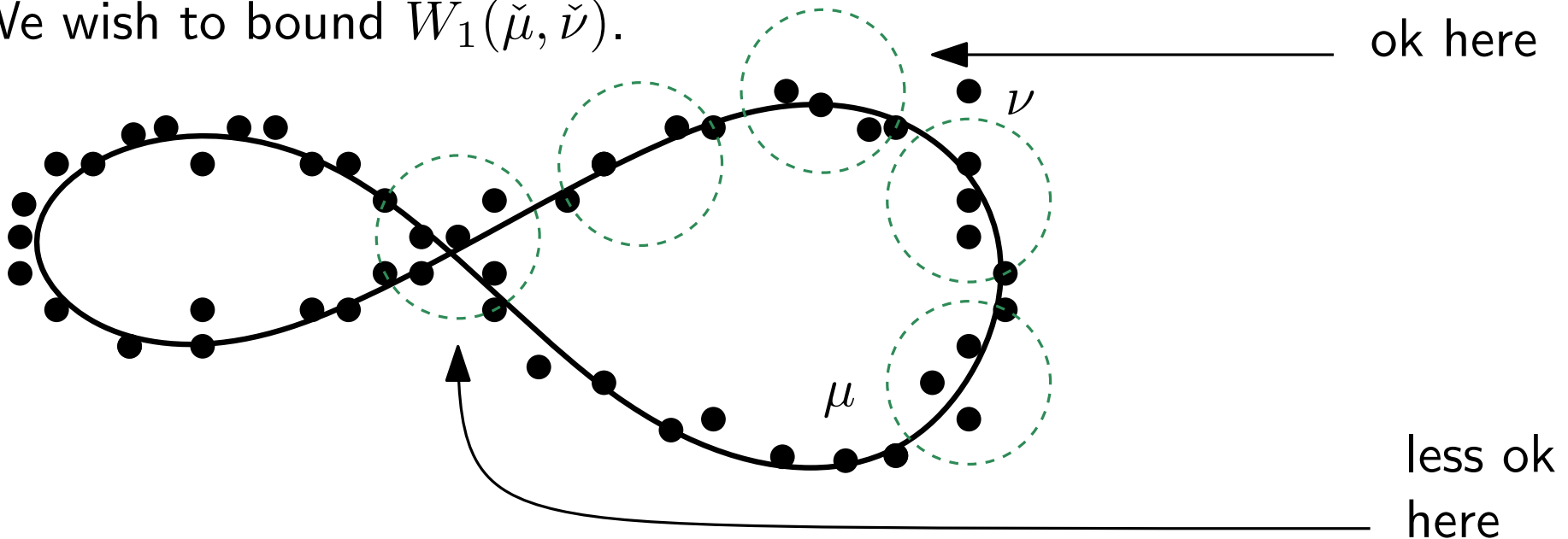
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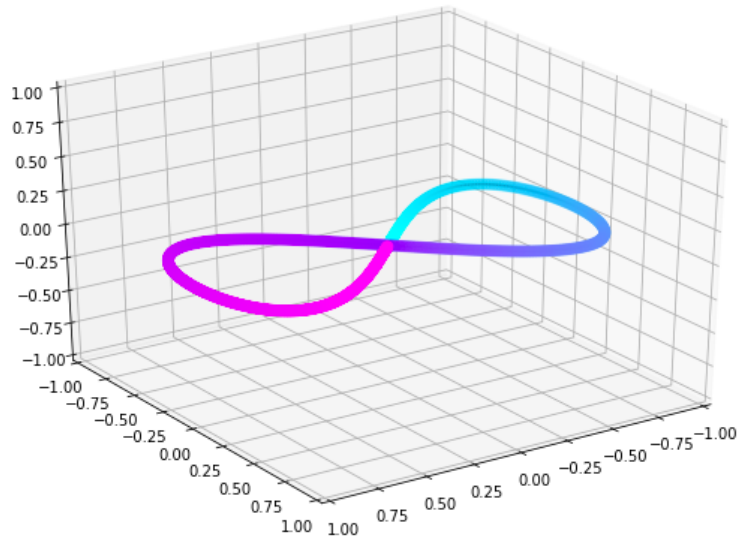


Tangent space estimation (4/6): Norm on $\mathbb{R}^n \times M_n(\mathbb{R})$

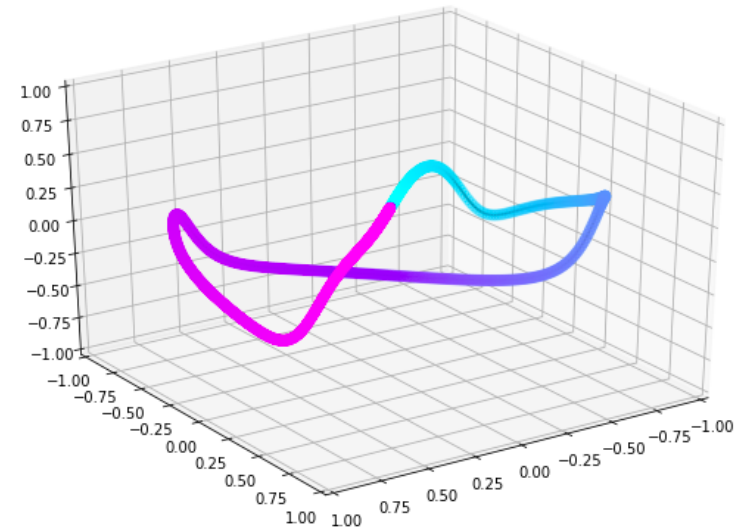
The Wasserstein norm $W_1(\check{\mu}, \check{\nu})$ is defined for the norm $\|\cdot\|_\gamma$ over $\mathbb{R}^n \times M_n(\mathbb{R})$, where

$$\|(x, A)\|_\gamma^2 = \|x\|^2 + \gamma^2 \|A\|_F^2.$$

The parameter γ controls how much is the tangent information important, compared to the spatial one.



$$\gamma = 0$$



$$\gamma = 1$$

Tangent space estimation (5/6): Stability

Proposition:

Consider ν any Radon measure on \mathbb{R}^n . μ is as before.

Let $w = W_1(\mu, \nu)$. Suppose that $r \leq \frac{1}{4\rho}$ and $w \leq \min(a^{\frac{d+1}{d}}, 1)(\frac{r}{4})^{d+1}$.

If we suppose that $w \leq 1$, then

$$W_1(\check{\mu}, \check{\nu}) \leq \frac{2\gamma}{r} c' \mu(\lambda^r) w^{\frac{1}{2(d+1)}} + \frac{2\gamma}{r} c w^{\frac{1}{d+1}} + w.$$

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Proof: Build an obvious transport plan $\check{\pi}$ between $\check{\mu}$ and $\check{\nu}$ from a transport plan π between μ and ν . Write

$$W_1(\check{\mu}, \check{\nu}) \leq \int \|x - y\| + \gamma \|\bar{\Sigma}_\mu(x) - \bar{\Sigma}_\nu(y)\|_F d\pi(x, y)$$

Show that, for every $x \in \text{supp}(\mu)$ and $y \in \text{supp}(\nu)$,

$$\|\bar{\Sigma}_\mu(x) - \bar{\Sigma}_\nu(y)\|_F \leq \frac{2}{r} (\|x - y\| + W_1(\overline{\mu_x}, \overline{\nu_y})),$$

where $\overline{\mu_x}$ is μ restricted to $\mathcal{B}(x, r)$ and normalized.

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Conclude

$$\begin{aligned} W_1(\check{\mu}, \check{\nu}) &\leq \int (1 + \frac{2\gamma}{r})\|x - y\| + \frac{2\gamma}{r} W_1(\bar{\mu}_x, \bar{\nu}_y) d\pi(x, y) \\ &= (1 + \frac{2\gamma}{r}) W_1(\mu, \nu) + \frac{2\gamma}{r} \int W_1(\bar{\mu}_x, \bar{\nu}_y) d\pi(x, y). \end{aligned}$$

Tangent space estimation (6/6): Approximation

Theorem:

Let ν be any Radon measure on \mathbb{R}^n , and μ as before. Choose $r < \frac{1}{4\rho}$.

Let $w = W_1(\mu, \nu)$. Suppose that $w \leq \min(a^{\frac{d+1}{d}}, 1)(\frac{r}{4})^{d+1}$ and $w \leq 1$.

Then

$$W_1(\check{\nu}, \check{\mu}_0) \leq 2\gamma(1 + \frac{c''}{r}w^{\frac{1}{2(d+1)}})\mu(\lambda^r) + \gamma c' r + \frac{2\gamma}{r}cw^{\frac{1}{d+1}} + w$$

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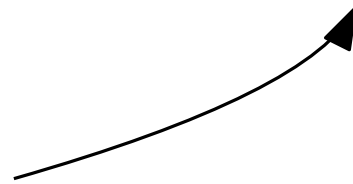
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Proof: Write

$$W_1(\check{\nu}, \check{\mu}_0) \leq W_1(\check{\nu}, \check{\mu}) + W_1(\check{\mu}, \check{\mu}_0)$$

Use stability



Use consistency



Tangent space estimation (6/6): Approximation

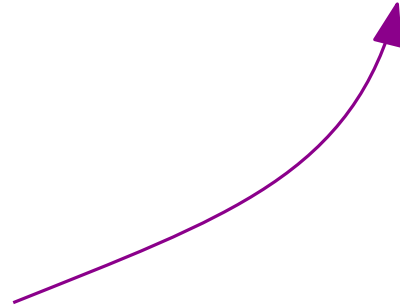
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Does this go to 0 as r does?

Quantification of λ

Proposition (not proven):

There exists a constant c such that for all $r < \dots$,

$$\mu_0(\lambda^r) \leq cr$$

As a consequence, the bound

$$W_1(\check{\nu}, \check{\mu}_0) \leq 2\gamma(1 + \frac{c''}{r}w^{\frac{1}{2(d+1)}})\mu(\lambda^r) + \gamma c' r + \frac{2\gamma}{r}cw^{\frac{1}{d+1}} + w$$

goes to 0 as r and w do (with $\frac{w}{r^{d+1}} \rightarrow 0$).

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DTM-based filtrations

Given $E = \mathbb{R}^N$ and μ a Radon measure on E , the 1-DTM-filtration is a filtration of the space E that is stable in Wasserstein distance.

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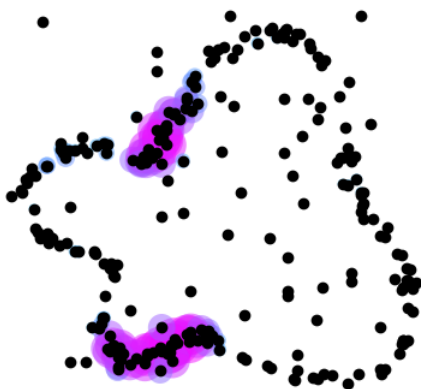
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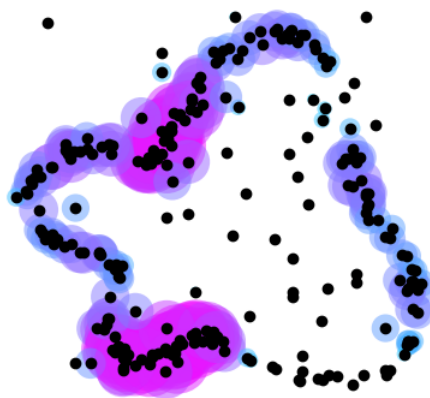
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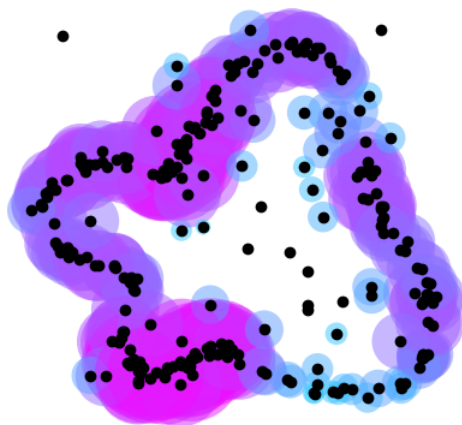
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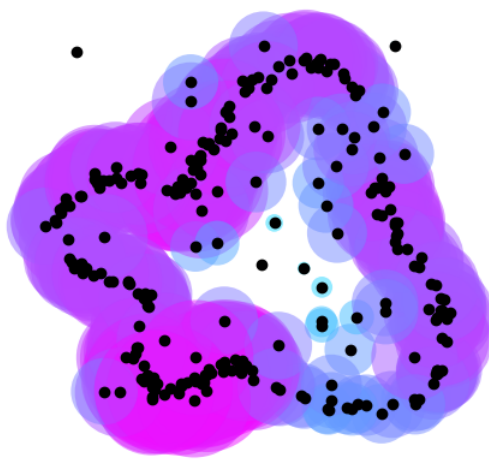
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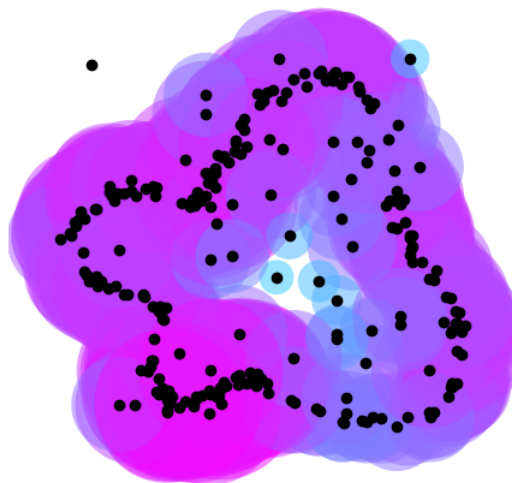
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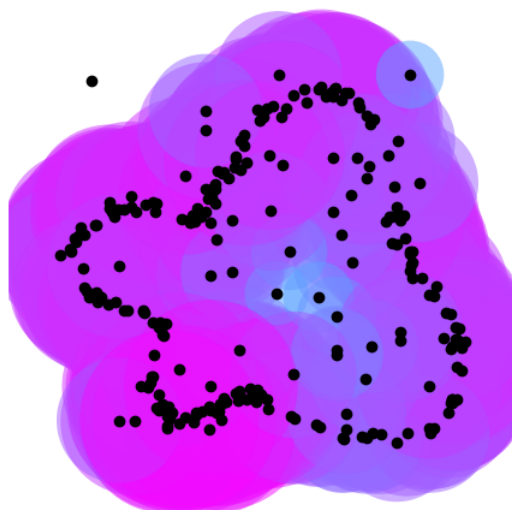
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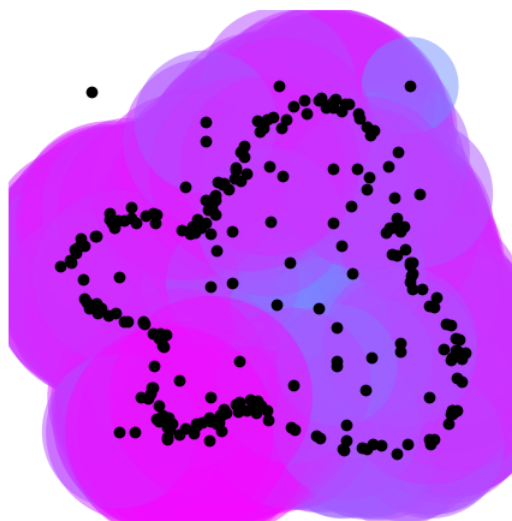
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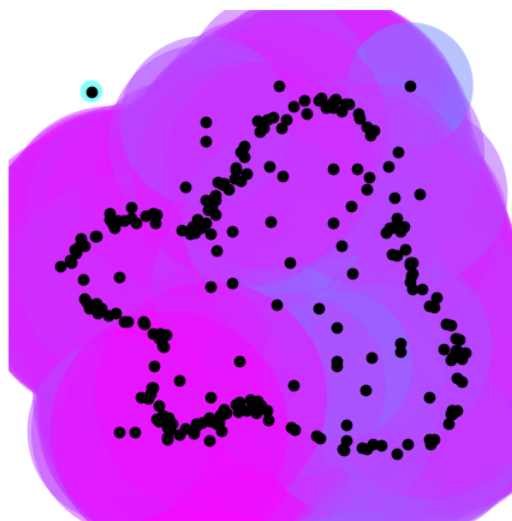
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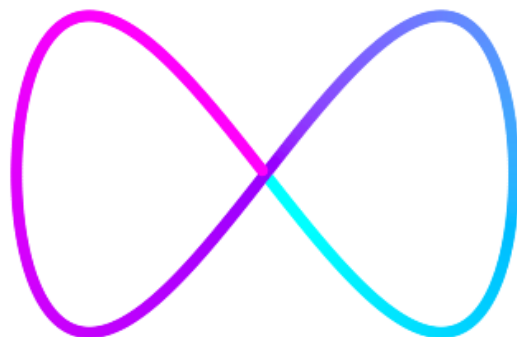
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Persistent homology of $\check{\mathcal{M}}$ (2/2)

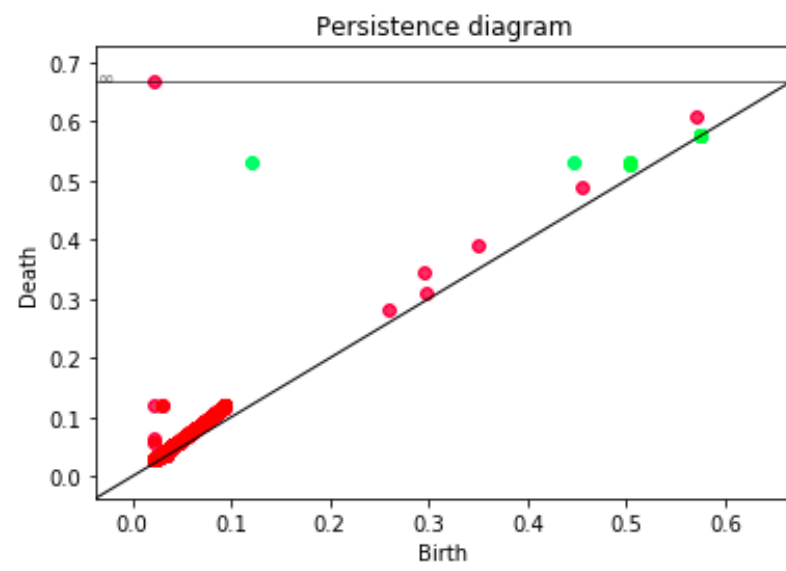
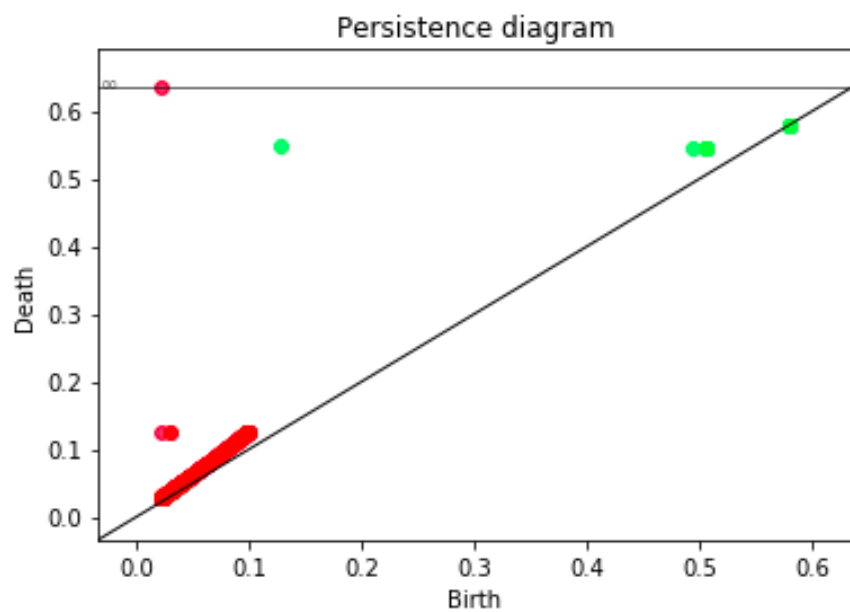
measures on
 $\mathbb{R}^n \times M_n(\mathbb{R})$

$\check{\mu}_0$

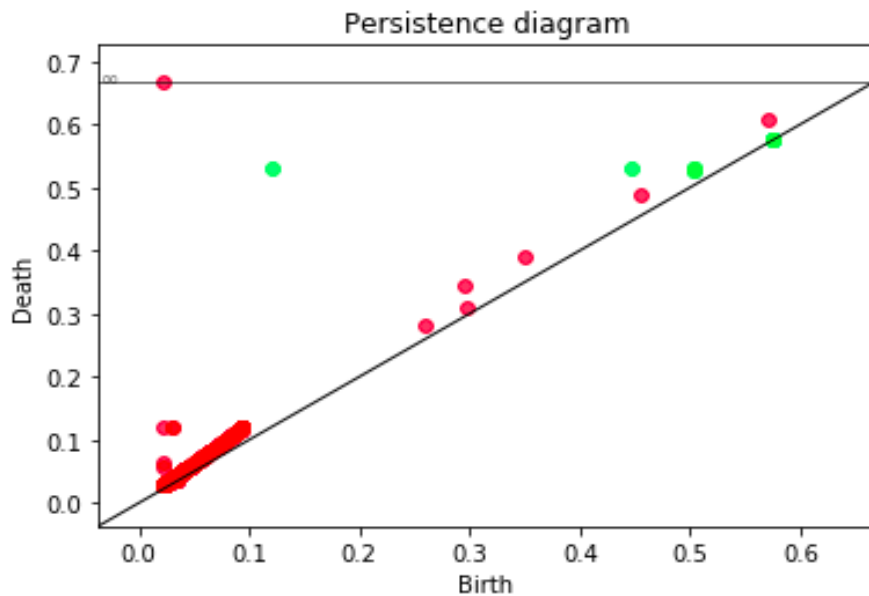
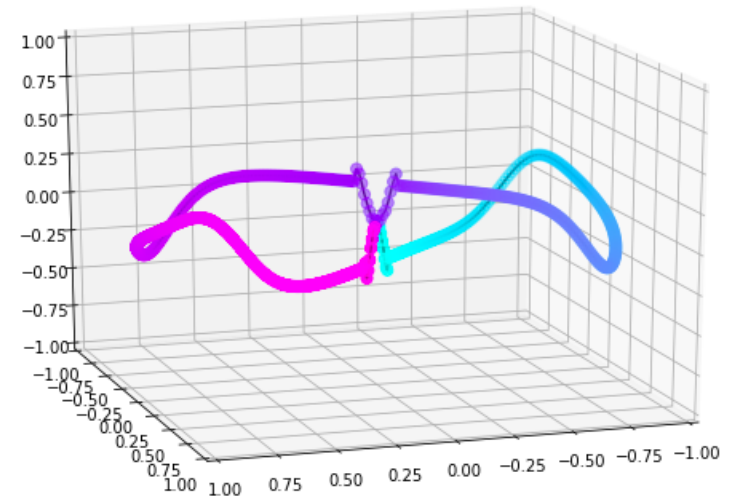
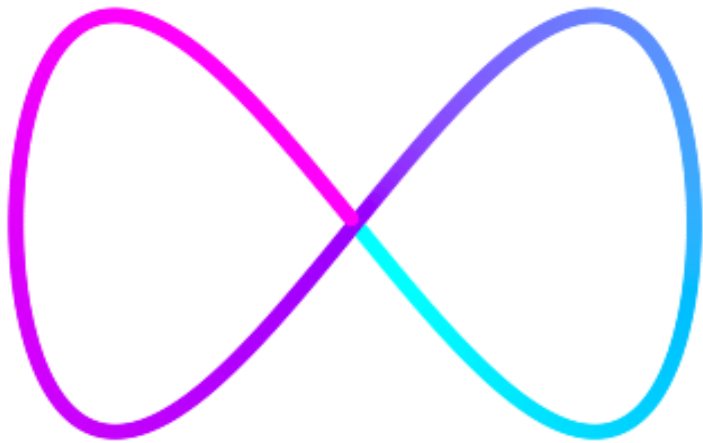


$\check{\nu}$

DTM-based
filtration



Thank you



Values of the DTM on X with parameter $m=0.01$

