

Persistent Intrinsic Volumes

Persistent homology as a tool for geometric inference



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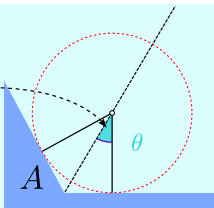


Problem. Recovering the boundary area of a set Y from an approximating set X as a function of $\varepsilon = d_H(X, Y)$.

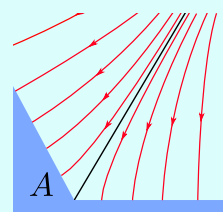
Some notations. d_A denotes the distance to any $A \subset \mathbb{R}^d$.

- $A^t := \{x \in \mathbb{R}^d \mid d_A(x) \leq t\}$ is the t -offset of A .
- $H_i(A)$ is the i -th homology vector space over an arbitrary field.
- $\dim H_i(A)$ is its i -th Betti number.

The μ -reach of a set A is the largest t such that for any $x \in A^t \setminus A$, the cosine of the half-angle between two closest points of x in X is less than μ .



Theorem. When $t < \text{reach}_\mu(A)$, for every $\varepsilon > 0$, there exists a flow whose trajectories, parametrized by the arc-length, make d_A decrease at speed $\geq \mu - \varepsilon$ on $A^t \setminus A$.

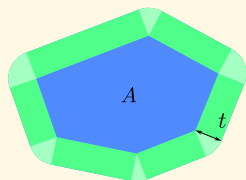


Smooth sets, polyhedra and most stratified sets have positive μ -reach.

Intrinsic volumes $V_0(A), \dots, V_d(A)$ are quantities defined for most subsets A of \mathbb{R}^d , and are related to their curvatures. For submanifolds, they coincide with the *Lipschitz-Killing curvatures*. For convex sets, they are defined by the volume of offsets.

$$\text{Vol}(A^t) =: \sum_{i=0}^d \omega_i t^i V_{d-i}(A)$$

volume of the unit ball of \mathbb{R}^i



More generally, they can be obtained via the **principal kinematic formula**.

$$\int_{\mathbb{R}^d} \chi(A \cap B(x, t)) dx = \sum_{i=0}^d \omega_i t^i V_{d-i}(A) =: Q_A(t)$$

Steiner's polynomial of A

Example. Up to scaling constants,

- $V_{d-1}(A)$ is the boundary area $\mathcal{H}^{d-1}(\partial A)$.
- $V_1(A)$ is the mean curvature of A .
- $V_0(A)$ is the Euler characteristic $\chi(A)$.

The **persistent Steiner polynomial** is defined as

$$Q_{X,\varepsilon}(t) := \int_{\mathbb{R}^d} \chi(\text{dgm}(X, \varepsilon, x)_t) dx$$

Final theorem. Recovering surrogate coefficient from $Q_{X,\varepsilon}$ yields quantities $V_i(X, \varepsilon, R)$ such that

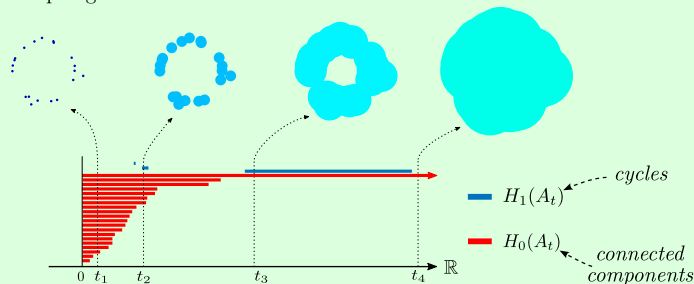
$$|V_i(X, \varepsilon, R) - V_i(Y^{2\varepsilon})| \leq \frac{P(i, d, R) M_R(Y^{2\varepsilon}) \varepsilon}{\mu} \quad \text{fraction in } i, d, R$$

When Y has bounded total curvatures, we further show that

$$|V_i(X, \varepsilon, R) - V_i(Y)| \leq (M_R(Y^{2\varepsilon}) + M_R(Y)) \frac{P(i, d, R) \varepsilon}{\mu}$$

A **filtration** is a family $(A_t)_{t \in \mathbb{R}}$ non-decreasing for the inclusion.

A **persistence diagram** associates to a filtration the values of birth and death of topological features.



Diagrams are made rigorous by their algebraic counterparts, called **persistent homology modules**. In dimension i , a bar is **born** (resp. **dies**) at time t when the dimension of $H_i(A_t)$ **increases** (resp. **decreases**) at t .

$$\begin{array}{ccc} A_t & \xleftarrow{\phi} & A_s \\ \downarrow \phi_* & \swarrow \text{inclusion-induced linear map} & \downarrow \\ H_i(A_t) & \xrightarrow{\phi_*} & H_i(A_s) \end{array}$$

If $(A_t), (B_t)$ are two filtrations such that for every t , $A_t \subset B_t$, the **image persistence module** $\text{dgm}(A, B)$ is given by the bars of the families $\iota_*(H_i(A_t))$.

$$\begin{array}{ccc} \longrightarrow \iota_*(H_i(A_t)) & \longrightarrow & \iota_*(H_i(A_s)) \longrightarrow \\ \uparrow \iota_* & & \uparrow \iota_* \\ \longrightarrow H_i(B_t) & \longrightarrow & H_i(B_s) \longrightarrow \\ \uparrow \iota_* & & \uparrow \iota_* \\ \longrightarrow H_i(A_t) & \longrightarrow & H_i(A_s) \longrightarrow \end{array}$$

Since $X^\varepsilon \subset Y^{2\varepsilon} \subset X^{3\varepsilon}$, for any $x \in \mathbb{R}^d$ we put $\text{dgm}(X, \varepsilon, x)$ the persistence image diagram induced by the two filtrations $(X^\varepsilon \cap B(x, t))_{t \in \mathbb{R}}$ and $(X^{3\varepsilon} \cap B(x, t))_{t \in \mathbb{R}}$, and $\text{dgm}(Y^{2\varepsilon}, x)$ for $(Y^{2\varepsilon} \cap B(x, t))_{t \in \mathbb{R}}$

$$\begin{array}{ccc} \longrightarrow H_i(X^{3\varepsilon} \cap B(x, t)) & \longrightarrow & H_i(X^{3\varepsilon} \cap B(x, s)) \longrightarrow \\ \uparrow & & \uparrow \\ \longrightarrow H_i(Y^{2\varepsilon} \cap B(x, t)) & \longrightarrow & H_i(Y^{2\varepsilon} \cap B(x, s)) \longrightarrow \\ \uparrow & & \uparrow \\ \longrightarrow H_i(X^\varepsilon \cap B(x, t)) & \longrightarrow & H_i(X^\varepsilon \cap B(x, s)) \longrightarrow \end{array}$$

χ -averaging Lemma : Let $\mu > 0$ such that $d_H(X, Y) \leq \varepsilon \leq \frac{1}{4} \text{reach}_\mu(X)$. Then

$$\int_0^R |\chi(\text{dgm}(X, \varepsilon, x)_t) - \chi(\text{dgm}(Y^{2\varepsilon}, x)_t)| dt \leq 2 \frac{\varepsilon}{\mu} N_0^R(\text{dgm}(Y^{2\varepsilon}, x))$$

Alternating sum of the Betti numbers at filtration value t Number of bars of the diagram inside $[0, R]$

Theorem. Let $\mu > 0$ be such that $d_H(X, Y) \leq \varepsilon \leq \frac{1}{4} \text{reach}_\mu(X)$. Then,

$$\int_0^R |Q_{X,\varepsilon}(t) - Q_{Y^{2\varepsilon}}(t)| dt \leq \frac{4\varepsilon}{\mu} \int_{\mathbb{R}^d} N_0^R(\text{dgm}(Y^{2\varepsilon}, x)) dx$$

Morse theory Lemma. Counting the critical points of $(d_x)_Y$ yields:

$$\int_{\mathbb{R}^d} N_0^R(\text{dgm}(Y^{2\varepsilon}, x)) dx \leq M_R(Y^{2\varepsilon}) \quad \leftarrow \text{depends on the total curvatures of } Y^{2\varepsilon}.$$