

The self-similar evolution of stationary point processes via persistent homology

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Based on: DS, Anna Wienhard, arXiv:2012.05751

Motivation

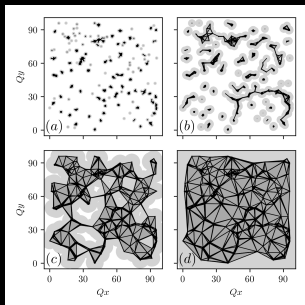
Point clouds from samples of functions $\psi_\omega(t) : \Lambda \rightarrow \mathbb{C}$, $\Lambda \subset \mathbb{R}^2$:

$$X_{\vec{\nu}, \omega}(t) := |\psi_\omega(t)|^{-1} [0, \vec{\nu}] \subset \Lambda. \quad (1)$$

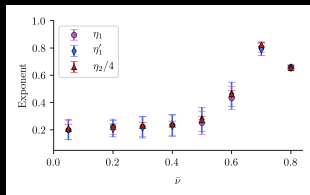
Scaling ansatz for density of persistence diagram measures:

$$\tilde{\rho}(t, \Lambda)(b, d) = (t/t')^{-\eta_2} \tilde{\rho}(t', \Lambda)((t/t')^{-\eta_1} b, (t/t')^{-\eta'_1} d). \quad (2)$$

Alpha complexes of different radii



[Reprinted from Spitz et al., arXiv:2001.02616]



[Reprinted from Spitz et al., arXiv:2001.02616]

Found $\eta_2/\eta_1 \simeq 4$ in $n = 2$. Recent simulations show $\eta_2/\eta_1 \simeq 5$ in $n = 3$. Geometric explanation?

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Random measures and point processes

\mathcal{M}^n : Space of boundedly finite measures on \mathbb{R}^n

\mathcal{N}^n : Integer-valued ones

$(\Omega, \mathcal{E}, \mathbb{P})$: Probability space

Random measure is a measurable map $\mu : (\Omega, \mathcal{E}, \mathbb{P}) \rightarrow (\mathcal{M}^n, \mathcal{B}(\mathcal{M}^n))$

Point process is a measurable map $\xi : (\Omega, \mathcal{E}, \mathbb{P}) \rightarrow (\mathcal{N}^n, \mathcal{B}(\mathcal{N}^n))$

To any realization ξ_ω , $\omega \in \Omega$, there exists countable set of *atoms*
 $\{x_1, x_2, \dots\} \subset \mathbb{R}^n$ such that $\forall A \in \mathcal{B}^n$:

$$\xi_\omega(A) = \sum_i \delta_{x_i}(A). \quad (3)$$

Point process *simple*, if a.s. all its atoms are distinct.

Point process *has all finite moments*, if $\forall k \geq 1, A \in \mathcal{B}_b^n: \mathbb{E}[\xi(A)^k] < \infty$.

For simple point process ξ define point clouds:

$$X_{\xi_\omega}(A) = \left\{ x_i \mid \xi_\omega(A) = \sum_i \delta_{x_i}(A) \right\}. \quad (4)$$

Filtered simplicial complexes and persistent homology

$X \subset \mathbb{R}^n$ point cloud. Čech complexes: $\check{C}_r(X) := \{\sigma \subseteq X \mid \bigcap_{x \in \sigma} B_r(x) \neq \emptyset\}$.

ℓ th persistence diagram of $H_\ell(\mathcal{C}(X))$ defined as finite multiset

$$\text{Dgm}_\ell(X) := \left\{ (b_i, d_i) \mid H_\ell(\mathcal{C}(X)) \cong \bigoplus_{i=1}^k I(b_i, d_i) \right\}. \quad (5)$$

Lemma

For ξ simple point process on \mathbb{R}^n , $D \circ X_\xi$ defines point process on space of pers. diags. \mathcal{D} endowed with the Wasserstein metric W_p , for any $p > n$, via

$$\omega \mapsto \left(A \mapsto \sum_{x \in D_A(X_{\xi_\omega}(A))} \delta_x \right), \quad D_A : X \mapsto \bigcup_{\ell=0}^{n-1} \text{Dgm}_\ell(X) \quad (6)$$

for all $\omega \in \Omega$, $A \in \mathcal{B}_b^n$. Equivalently, $D_A \circ X_\xi(A) : \Omega \rightarrow \mathcal{D}$ is a random variable for each $A \in \mathcal{B}_b^n$.

Bounded total persistence

M : Triangulable, compact metric space. K : Simp. complex giving triangulation ϑ of M , $\text{diam}(\sigma) := \max_{x,y \in \sigma} d(\vartheta(x), \vartheta(y))$,

$$\text{mesh}(K) := \max_{\sigma \in K} \text{diam}(\sigma), \quad N(r) := \min_{\text{mesh}(K) \leq r} \#K \quad (7)$$

Degree- k total persistence of filtration of sublevel sets of f :

$$\text{Pers}_k(f) := \sum_{\text{pers}(x) > t} \text{pers}(x)^k. \quad (8)$$

Proposition (Cohen-Steiner *et al.*, Found. Comp. Math. 10(2), 2010)

Assume that size of smallest triangulation of M grows polynomially with one over mesh, i.e. there exist C_0, m , such that $N(r) \leq C_0/r^m$ for all $r > 0$. Let $\delta > 0$ and $k = m + \delta$. Then,

$$\text{pers}(f) \leq \frac{m + 2\delta}{\delta} C_0 \text{Lip}(f)^m \text{Amp}(f)^\delta, \quad (9)$$

where $\text{Amp}(f) := \max_{x \in M} f(x) - \min_{y \in M} f(y)$.

M implies bounded degree- k total persistence, if $\exists C_M > 0$: $\text{Pers}_k(f) \leq C_M$ for every tame function $f : M \rightarrow \mathbb{R}$ with $\text{Lip}(f) \leq 1$.

Example: Compact Riemannian n -manifolds have bounded degree- n total pers.

Ergodicity of point processes

Random measure μ *stationary*, if $\forall x \in \mathbb{R}^n$ the probability distributions of μ and $\theta_x \mu$ coincide ($(\theta_x \mu)(A) := \mu(A + x)$)

Stationary random measure μ *ergodic*, if $\mathbb{P}(\mu \in \mathcal{A}) \in \{0, 1\}$ for all Borel sets \mathcal{A} of \mathcal{M}^n that are invariant under translation by x for all $x \in \mathbb{R}^n$.

$\{A_k\} \subset \mathcal{B}_b^n$ called a *convex averaging sequence* if

$$(i) \text{ each } A_k \text{ is convex,} \quad (ii) A_k \subseteq A_{k+1} \text{ for all } k, \quad (iii) \lim_{k \rightarrow \infty} r(A_k) = \infty, \quad (10)$$

where $r(A) := \sup\{r \mid A \text{ contains a ball of radius } r\}$.

Proposition (Daley & Vere-Jones 2007)

$\{A_k\}$ *convex averaging sequence*. When random measure ξ on \mathbb{R}^d is stationary, ergodic and has finite expectation measure with

$$m = \mathbb{E}[\xi([0, 1]^n)] = m \lambda_n([0, 1]^n), \quad (11)$$

then $\xi(A_k)/\lambda_n(A_k) \rightarrow m$ almost surely as $k \rightarrow \infty$.

Ergodicity in persistent homology

Definition

ξ stationary simple point process on \mathbb{R}^n with finite expectation measure, $\{A_k\}$ convex averaging sequence. Set for all k :

$$n(X_k) := \# \bigcup_{\ell=0}^{n-1} \text{Dgm}_\ell(X_\xi(A_k)). \quad (12)$$

Say that ξ is ergodic in persistence if a.s. for any $\epsilon > 0 \exists N \in \mathbb{N}$ such that for all $k, l \geq N$,

$$\left| \frac{n(X_k)}{n(X_l)} - \frac{\lambda_n(A_k)}{\lambda_n(A_l)} \right| < \epsilon. \quad (13)$$

Lemma

ξ simple point process on \mathbb{R}^n having all finite moments. If ξ is stationary and ergodic, then ξ is ergodic in persistence.

Persistence diagram measures

ξ simple point process, $A \in \mathcal{B}_b^n$, define for all $\omega \in \Omega$:

$$\rho_\omega(A) := \sum_{x \in D_A(X_{\xi_\omega}(A))} \delta_x. \quad (14)$$

Defines a point process $\rho_\cdot(A)$ on $\Delta := \{(b, d) \mid 0 \leq b < d \leq \infty\}$.

Persistence diagram measure: $\rho_\cdot : \Omega \times \mathcal{B}_b^n \rightarrow \mathcal{N}(\Delta)$.

If they exist, its first moment measures define $\mathfrak{p}(A) := \mathbb{E}[\rho_\omega(A)]$.

Persistence diagram expectation measure: $\mathfrak{p} : \mathcal{B}_b^n \rightarrow \mathcal{R}(\Delta)$, $A \mapsto \mathfrak{p}(A)$

Lemma

*ξ simple stationary and ergodic point process on \mathbb{R}^n having all finite moments.
Then the corresponding persistence diagram expectation measure exists.*

Geometric quantities

Number of persistent homology classes:

$$n_\omega(A) := \int_\Delta \rho_\omega(A)(dx) = \rho_\omega(A)(\Delta) \quad (15)$$

Expectation: $n(A) := \int_\Delta p(A)(dx) = p(A)(\Delta)$

Let $q > 0$. *Degree- q persistence:*

$$l_{q,\omega}(A) := \left[\frac{1}{n_\omega(A)} \int_\Delta \text{pers}(x)^q \rho_\omega(A)(dx) \right]^{1/q} \quad (16)$$

Expectation: $l_q(A) := \left[\frac{1}{n(A)} \int_\Delta \text{pers}(x)^q p(A)(dx) \right]^{1/q}$

Maximum death:

$$d_{\max,\omega}(A) := \max\{d \mid (b, d) \in D_A(X_{\rho_\omega}(A))\} = \lim_{p \rightarrow \infty} \left[\int_\Delta d(x)^p \rho_\omega(A)(dx) \right]^{1/p} \quad (17)$$

Expectation: $\vartheta_{\max}(A) := \lim_{p \rightarrow \infty} \left[\int_\Delta d(x)^p p(A)(dx) \right]^{1/p}$

Geometric quantities

Proposition

\mathfrak{p} persistence diagram expectation measure, $A \in \mathcal{B}_b^n$. Find

$$n(A) = \mathbb{E}[n_\omega(A)], \quad \mathfrak{d}_{\max}(A) = \mathbb{E}[d_{\max,\omega}(A)]. \quad (18)$$

If $\mathfrak{p}(A)$ is boundedly finite, then $n(A) < \infty$. If $\text{supp}(\mathfrak{p}(A)) \subset \Delta$ is bounded, then $\mathfrak{d}_{\max}(A) < \infty$.

Proposition

ξ stationary, ergodic simple point process on \mathbb{R}^n having all finite moments, \mathfrak{p} persistence diagram expectation measure from ξ , $\{A_k\}$ convex averaging sequence, $\epsilon > 0$. Find for all $q > 0$ and k suff. large:

$$|l_q(A_k) - \mathbb{E}[l_{q,\omega}(A_k)]| < \epsilon. \quad (19)$$

Let $A \in \mathcal{B}_b^n$. If $\mathfrak{p}(A)$ is boundedly finite and $\text{supp}(\mathfrak{p}(A)) \subset \Delta$ is bounded, then $l_q(A) < \infty$ for all $q > 0$.

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Functional summaries

Problem: Persistence diagrams do not naturally lead to statistical goals!

$\mathcal{F}(T)$: Collection of functions on compact metric space T , $f : T \rightarrow \mathbb{R}$

Functional summary F is a map from the space of persistence diagrams to such a collection of functions, $F : \mathcal{D} \rightarrow \mathcal{F}(T)$

F *additive*, if for any two persistence diagrams $D, E \in \mathcal{D}$:

$\mathcal{F}(D + E) = \mathcal{F}(D) + \mathcal{F}(E)$ as sum of multisets

F *uniformly bounded*, if a constant $U < \infty$ exists, such that

$$\sup_{f \in \text{im}(F)} \sup_{s \in T} |f(s)| \leq U \quad (20)$$

Proposition (Berry *et al.*, arXiv:1804.01618)

F *uniformly bounded functional summary*, $D_i \in \mathcal{D}$ for $i \in \mathbb{N}$, sampled from a probability space $(\mathcal{D}, \mathcal{B}(\mathcal{D}), \mathbb{P}_{\mathcal{D}})$, $f_i := F(D_i)$. If $\text{im}(F)$ equicontinuous, then a.s. for $n \rightarrow \infty$

$$\sup_{s \in T} \left| \frac{1}{m} \sum_{i=1}^m f_i(s) - \mathbb{E}[F(D)(s)] \right| \rightarrow 0. \quad (21)$$

Functional summaries and persistence diagram measures

For additive func. summary \mathcal{A} , $s \in T$:

$$\mathcal{A}(D_A(X_{\xi\omega}(A)))(s) = \sum_{x \in D_A(X_{\xi\omega}(A))} \mathcal{A}(\{x\})(s) = \int_{\Delta} \mathcal{A}(\{x\})(s) \rho_{\omega}(A)(dx). \quad (22)$$

\implies Persistence diagram measures show up.

Proposition

Assume that pers. diag. expectation measure \mathfrak{p} exists. Then, for any functional summary $\mathcal{F} \in \mathcal{F}(K)$, $s \in K$, $A \in \mathcal{B}_b^n$ have

$$\mathbb{E} \left[\int_{\Delta} \mathcal{F}(\{x\})(s) \rho_{\omega}(A)(dx) \right] = \int_{\Delta} \mathcal{F}(\{x\})(s) \mathfrak{p}(A)(dx). \quad (23)$$

Intensive functional summaries

ξ stationary, ergodic simple point process on \mathbb{R}^n , $\mathcal{F} : \mathcal{D} \rightarrow \mathcal{F}(T)$ functional summary.

Inspired by statistical mechanics:

\mathcal{F} ξ -intensive, if for any convex averaging sequence $\{A_k\}$, $\epsilon > 0$ and k sufficiently large:

$$\lim_{l \rightarrow \infty} \|\mathcal{F}(D_l) - \mathcal{F}(D_k)\|_\infty < \epsilon, \quad (24)$$

$\|\cdot\|_\infty$: sup-norm, $D_k := \bigcup_{\ell=0}^{n-1} \text{Dgm}_\ell(X_{\xi_\omega}(A_k))$.

Lemma

$\mathcal{A} \in \mathcal{A}(K)$ additive functional summary, ξ stationary, ergodic simple point process on \mathbb{R}^n , D_k as before. Then, $\mathcal{A}(D_k)/\lambda_n(A_k)$ a.s. defines a ξ -intensive functional summary.

\implies Intuition for ergodicity on level of persistence diagrams.

Corollary

$\mathcal{A}(D_k)/n(D_k)$ is a ξ -intensive functional summary

Intensive functional summaries

Proposition

ξ stationary, ergodic simple point process on \mathbb{R}^n , \mathcal{F} a ξ -intensive functional summary, $\{A_k\}$ convex averaging sequence. Pick samples $\omega_i \in \Omega$, $i \in \mathbb{N}$, according to the probability distribution \mathbb{P} , define

$$D_{k,i} := \bigcup_{\ell=0}^{n-1} \text{Dgm}_\ell(X_{\xi\omega_i}(A_k)). \quad (25)$$

Then for any $j \in \mathbb{N}$, $\epsilon > 0$ and sufficiently large k find a.s.:

$$\sup_{s \in K} \left| \lim_{l \rightarrow \infty} \mathcal{F}(D_{l,j})(s) - \lim_{m \rightarrow \infty} \frac{1}{m} \sum_{i=1}^m \mathcal{F}(D_{k,i})(s) \right| < \epsilon. \quad (26)$$

\implies For intensive functional summaries ensemble-average and infinite-volume average converge to each other.

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Self-similar scaling

Inspiration: How to describe behavior as it typically appears in quantum-physical time evolutions in persistent-homological quantities?

$(\mathfrak{p}(t))_{t \in (T_0, T_1)}$ family of persistence diagram expectation measures, $0 < T_0 < T_1$. For $t \in (T_0, T_1)$, $A \in \mathcal{B}_b^n$ set $\mathfrak{p}(t, A) := \mathfrak{p}(t)(A)$, $\{A_k\}$ convex averaging sequence.

$(\mathfrak{p}(t))_{t \in (T_0, T_1)}$ scales self-similarly between times T_0 and T_1 with exponents $\eta_1, \eta_2 \in \mathbb{R}$, if for all $t, t' \in (T_0, T_1)$, $B \in \mathcal{B}(\Delta)$, k suff. large:

$$\mathfrak{p}(t, A_k)(B) = (t/t')^{-\eta_2} \mathfrak{p}(t', A_k)((t/t')^{-\eta_1} B), \quad (27)$$

where $\kappa B := \{(\kappa b, \kappa d) \mid (b, d) \in B\}$ for $\kappa \in [0, \infty)$.

Intuition based on pers. hom. with length scales as filtration parameter: Any persistence length scale blows up as a power-law t^{η_1} , find e.g.:

$$\mathfrak{n}(t, A_k) = (t/t')^{-\eta_2} \mathfrak{n}(t', A_k), \quad (28a)$$

$$\mathfrak{l}_q(t, A_k) = (t/t')^{\eta_1} \mathfrak{l}_q(t', A_k), \quad (28b)$$

$$\mathfrak{d}_{\max}(t, A_k) = (t/t')^{\eta_1} \mathfrak{d}_{\max}(t', A_k). \quad (28c)$$

Packing relation

Can we find a relation between the occurring exponents based on geometrical arguments?

Based on notion of bounded total persistence obtain:

Lemma (Packing lemma)

ξ simple point process on \mathbb{R}^n , ρ persistence diagram measure computed from ξ . Then there exists $c > 0$, s.t. for any $\delta > 0$, $\omega \in \Omega$ and $A \in \mathcal{B}_b^n$:

$$n_\omega(A) \leq \frac{c(n+2\delta)}{\delta} \frac{d_{\max, \omega}(A)^\delta}{l_{n+\delta, \omega}(A)^{n+\delta}}. \quad (29)$$

Theorem (Packing relation)

$(\xi(t))_{t \in (T_0, T_1)}$ family of stationary, ergodic simple point processes on \mathbb{R}^n having all finite moments, $(p(t))_t$ family of persistence diagram expectation measures computed from $(\xi(t))_t$. Assume that the family scales self-similarly between T_0 and T_1 with exponents $\eta_1, \eta_2 \in \mathbb{R}$. Then, if the interval (T_0, T_1) is sufficiently extended find a.s.:

$$\eta_2 = n\eta_1. \quad (30)$$

\implies Confirms exponents found in numerical simulations.

Example: Poisson process with power-law scaling intensity

Point processes $(\xi(t))_{t \in [0, \infty)}$ on \mathbb{R}^n form *time-dependent Poisson process*, if there exists $\gamma : [0, \infty) \rightarrow (0, \infty)$, s.t. $\forall t \in [0, \infty)$,

(i) $\mathbb{E}[\xi(t, A)] = \gamma(t)\lambda_n(A)$,

(ii) for every $m \in \mathbb{N}$ and all pairwise disjoint Borel sets $A_1, \dots, A_m \in \mathcal{B}^n$ the random variables $\xi(t, A_1), \dots, \xi(t, A_m)$ are independent.

Proposition

$(\xi(t))_{t \in [0, \infty)}$ *time-dependent Poisson process with intensity function*

$$\gamma(t) = \gamma_0 t^{-n\eta_1}, \quad (31)$$

$\gamma_0 > 0$ and $\eta_1 \geq 0$, $\{A_k\}$ *convex averaging sequence*. Then the family $(\mathfrak{p}(t, A_k)/\lambda_n(A_k))_t$ *converges vaguely for $k \rightarrow \infty$ to a family of Radon measures $(\mathfrak{P}(t))$ which scales self-similarly between 0 and ∞ with exponents η_1 and $n\eta_1$.*

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Strong law of large numbers

For the example of the time-dependent Poisson process needed to show the following generalization to the same results for cubes [Hiraoka, Shirai, Trinh, Ann. Appl. Prob. 28(5), 2018].

Theorem (Strong law of large numbers for persistent Betti numbers)

ξ simple point process on \mathbb{R}^n having all finite moments, $\{A_k\}$ convex averaging sequence. If ξ is stationary, then for any $0 \leq r \leq s < \infty$ and $\ell \in \mathbb{Z}$ there exists a constant $\hat{\beta}_\ell^{r,s}$, s.t.

$$\frac{\mathbb{E}[\beta_\ell^{r,s}(\mathcal{C}(X_k))]}{\lambda_n(A_k)} \rightarrow \hat{\beta}_\ell^{r,s} \quad \text{for } k \rightarrow \infty. \quad (32)$$

Additionally, if ξ is ergodic, then a.s.

$$\frac{\beta_\ell^{r,s}(\mathcal{C}(X_k))}{\lambda_n(A_k)} \rightarrow \hat{\beta}_\ell^{r,s} \quad \text{for } k \rightarrow \infty. \quad (33)$$

Proof is lengthy, using techniques from convex and integral geometry.

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Conclusions

- ▶ Theory of point processes well-suited to describe persistent homology in a probabilistic setting.
- ▶ Ergodicity, functional summaries and persistence diagram measures lead to diverse geometric quantities with well-defined expectations.
- ▶ Self-similar scaling approach yields interesting insights into geometry (packing relation).
- ▶ Strong law of large numbers for persistent Betti numbers generalized to arbitrary convex averaging sequences.

Further questions

- ▶ Can the packing relation be extended to non-ergodic point processes and their persistent homology?
- ▶ Do point processes exist, for which correlations do not reveal self-similar scaling but persistence diagram measures do?
- ▶ Can we generalize the results to weighted simplicial complexes?