

# Consistency of Persistent Homology

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We first recall the consistency:

Suppose we obtain a sample  $X_1, \dots, X_n \sim P$ . Let  $\theta(P)$  be a parameter, which is some function of  $P$ . Let  $\hat{\theta} = \hat{\theta}(X_1, \dots, X_n)$  denote an estimator for  $\theta$ , which is a function of a sample. Consistency is about, whether the estimator  $\hat{\theta}$  converge in probability to  $\theta$ , i.e.  $\hat{\theta} \xrightarrow{P} \theta$ . More precisely, can we find some function  $f(n)$  of the sample size  $n$  such that  $d(\hat{\theta}, \theta) = O_P(f(n))$ ? This is analogous to the Law of Large Number.

Let  $\mathbb{X} \subset \mathbb{R}^d$  be the target geometric structure, and  $P$  be a distribution on  $\mathbb{R}^d$  with  $\text{supp}(P) = \mathbb{X}$ . Let  $X_1, \dots, X_n$  be i.i.d. samples from  $P$  and  $\mathcal{X} = \{X_1, \dots, X_n\}$ . For the consistency of persistent homology, the distance is the bottleneck distance  $d_B$ , and  $\theta(P)$  and  $\hat{\theta}(\mathcal{X})$  should be appropriate persistent homologies of  $P$  and  $\mathcal{X}$ , respectively. We consider two cases:

1. Persistent homologies from Čech complexes and Vietoris-Rips complexes. Let  $\mathcal{PC}(\mathbb{X})$  and  $\mathcal{PC}(\mathcal{X})$  be the persistent homologies induced from Čech complexes  $\{H_k \check{\text{Cech}}_{\mathbb{R}^d}(\mathbb{X}, r)\}_{r \in \mathbb{R}}$  and  $\{H_k \check{\text{Cech}}_{\mathbb{R}^d}(\mathcal{X}, r)\}_{r \in \mathbb{R}}$ , respectively. Similarly, let  $\mathcal{PR}(\mathbb{X})$  and  $\mathcal{PR}(\mathcal{X})$  be the persistent homologies induced from Vietoris-Rips complexes  $\{H_k \text{Rips}(\mathbb{X}, r)\}_{r \in \mathbb{R}}$  and  $\{H_k \text{Rips}(\mathcal{X}, r)\}_{r \in \mathbb{R}}$ , respectively. We would like to know  $d_B(\mathcal{PC}(\mathbb{X}), \mathcal{PC}(\mathcal{X})) = O_P(f(n))$  and  $d_B(\mathcal{PR}(\mathbb{X}), \mathcal{PR}(\mathcal{X})) = O_P(f(n))$ .

## Consistency of Čech complexes and Vietoris-Rips complexes

Assume  $\mathbb{X}$  is compact. Recall the stability theorem for Čech complexes and Vietoris-Rips complexes:

**Corollary.** For a compact set  $\mathbb{X} \subset \mathbb{R}^d$  and  $\mathcal{X} \subset \mathbb{X}$ ,

$$\begin{aligned} d_B(\mathcal{PC}_{\mathbb{R}^d}(\mathbb{X}), \mathcal{PC}_{\mathbb{R}^d}(\mathcal{X})) &\leq d_H(\mathbb{X}, \mathcal{X}). \\ d_B(\mathcal{PR}(\mathbb{X}), \mathcal{PR}(\mathcal{X})) &\leq d_H(\mathbb{X}, \mathcal{X}). \end{aligned}$$

For a distribution  $P$ , we assume  $(a, b)$  assumption:

**Definition.**  $P$  satisfies  $(a, b)$  assumption if there exists  $r_0 > 0$  such that for all  $x \in \text{supp}(P)$  and for all  $r < r_0$ ,

$$P(\mathcal{B}(x, r)) \geq ar^b.$$

Recall that under  $(a, b)$  assumption, we have probabilistic bound on the Hausdorff distance between  $\mathbb{X}$  and  $\mathcal{X}$ :

**Proposition** ([2, Proposition 7.2][1, Theorem 2]). Let  $P$  be a distribution on  $\mathbb{R}^d$  with  $\text{supp}(P) = \mathbb{X}$ , and assume  $P$  satisfies  $(a, b)$  assumption with  $a, b > 0$ . Let  $X_1, \dots, X_n$  be i.i.d. samples from  $P$ , and let  $\mathcal{X} = \{X_1, \dots, X_n\}$ . Then there exists  $\epsilon_0 > 0$  such that for all  $\epsilon < \epsilon_0$ ,

$$P(d_H(\mathbb{X}, \mathcal{X}) < \epsilon) \geq 1 - a^{-1} \epsilon^{-b} \exp(-na\epsilon^b). \quad (1)$$

This directly implies that with probability  $1 - \delta$ , with large enough  $n$ ,

$$d_H(\mathbb{X}, \mathcal{X}) < C \left( \frac{\log n}{n} \right)^{1/b},$$

and hence

$$d_H(\mathbb{X}, \mathcal{X}) = O_P \left( \left( \frac{\log n}{n} \right)^{1/b} \right).$$

Then this implies both that

$$\begin{aligned} d_B(\mathcal{PC}_{\mathbb{R}^d}(\mathbb{X}), \mathcal{PC}_{\mathbb{R}^d}(\mathcal{X})) &= O_P \left( \left( \frac{\log n}{n} \right)^{1/b} \right), \\ d_B(\mathcal{PR}(\mathbb{X}), \mathcal{PR}(\mathcal{X})) &= O_P \left( \left( \frac{\log n}{n} \right)^{1/b} \right). \end{aligned}$$

(1) not only gives the probabilistic bound as above, but this also gives the bound on the expectation as well. Roughly speaking, this is deduced from

$$\mathbb{E} [d_H(\mathbb{X}, \mathcal{X})] = \int_0^\infty P(d_H(\mathbb{X}, \mathcal{X}) > \epsilon) d\epsilon.$$

**Theorem** ([1, Theorem 4]). *Let  $P$  be a distribution on  $\mathbb{R}^d$  with  $\text{supp}(P) = \mathbb{X}$ , and assume  $P$  satisfies  $(a, b)$  assumption with  $a, b > 0$ . Let  $X_1, \dots, X_n$  be i.i.d. samples from  $P$ , and let  $\mathcal{X} = \{X_1, \dots, X_n\}$ . Then,*

$$\mathbb{E} [d_H(\mathbb{X}, \mathcal{X})] \leq C \left( \frac{\log n}{n} \right)^{1/b},$$

where  $C$  only depends on  $a$  and  $b$ . And correspondingly,

$$\begin{aligned} \mathbb{E} [d_B(\mathcal{PC}_{\mathbb{R}^d}(\mathbb{X}), \mathcal{PC}_{\mathbb{R}^d}(\mathcal{X}))] &\leq C \left( \frac{\log n}{n} \right)^{1/b}, \\ \mathbb{E} [d_B(\mathcal{PR}(\mathbb{X}), \mathcal{PR}(\mathcal{X}))] &\leq C \left( \frac{\log n}{n} \right)^{1/b}. \end{aligned}$$

The convergence rate  $\left( \frac{\log n}{n} \right)^{1/b}$  of Čech complexes and Vietoris-Rips complexes is in fact minimax up to a logarithmic term.

**Theorem** ([1, Theorem 4]). *Let  $\mathcal{P}$  be a set of distributions  $P$  with  $\text{supp}(P)$  being compact and satisfying  $(a, b)$  assumption with fixed  $a, b > 0$ . Then for any estimator  $\hat{\text{dgm}}_n$  (that is, a function of data  $X_1, \dots, X_n$ ),*

$$\begin{aligned} \sup_{P \in \mathcal{P}} \mathbb{E}_P [d_B(\mathcal{PC}_{\mathbb{R}^d}(\mathbb{X}), \mathcal{PC}_{\mathbb{R}^d}(\mathcal{X}))] &\geq Cn^{-1/b}, \\ \sup_{P \in \mathcal{P}} \mathbb{E}_P [d_B(\mathcal{PR}(\mathbb{X}), \mathcal{PR}(\mathcal{X}))] &\geq Cn^{-1/b}. \end{aligned}$$

## References

- [1] Frédéric Chazal, Marc Glisse, Catherine Labruère, and Bertrand Michel. Convergence rates for persistence diagram estimation in topological data analysis. *J. Mach. Learn. Res.*, 16:3603–3635, 2015.
- [2] Partha Niyogi, Stephen Smale, and Shmuel Weinberger. Finding the homology of submanifolds with high confidence from random samples. *Discrete & Computational Geometry*, 39(1-3):419–441, 2008.