

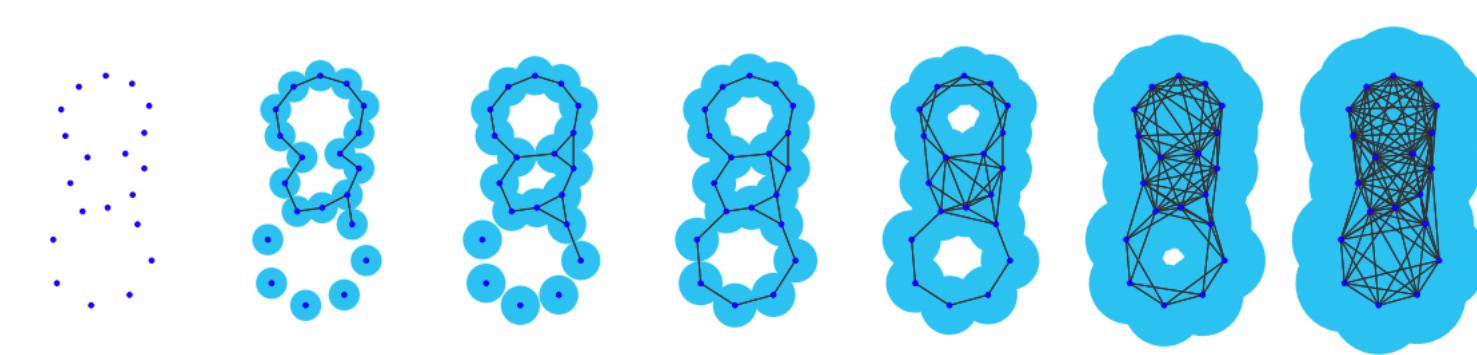
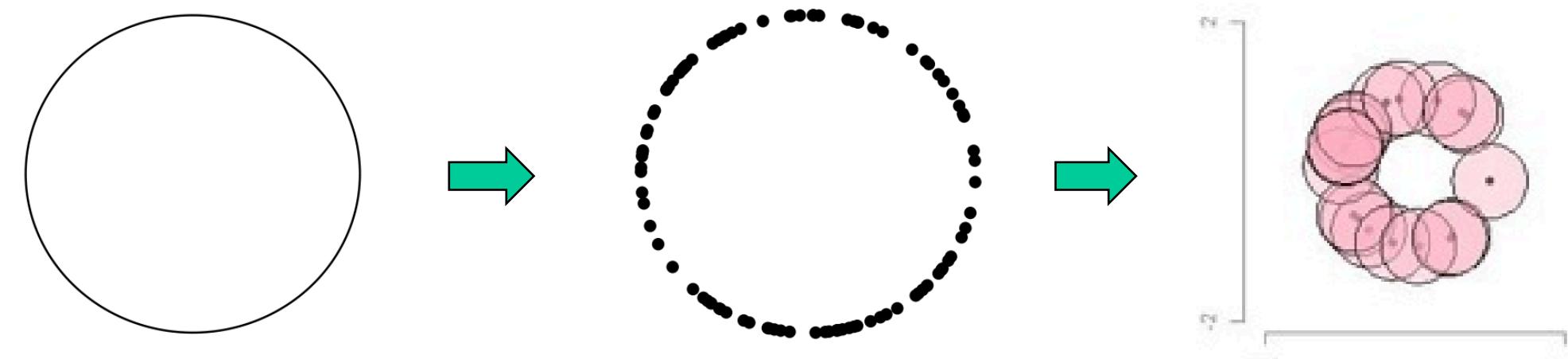
Kernel Methods in Topological Data Analysis

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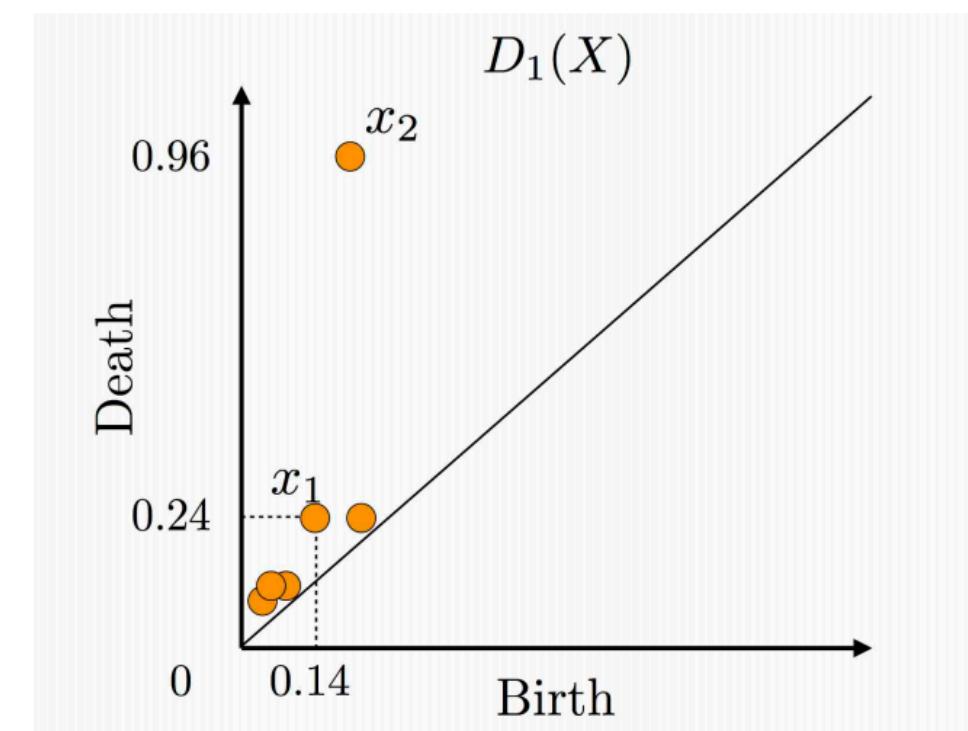
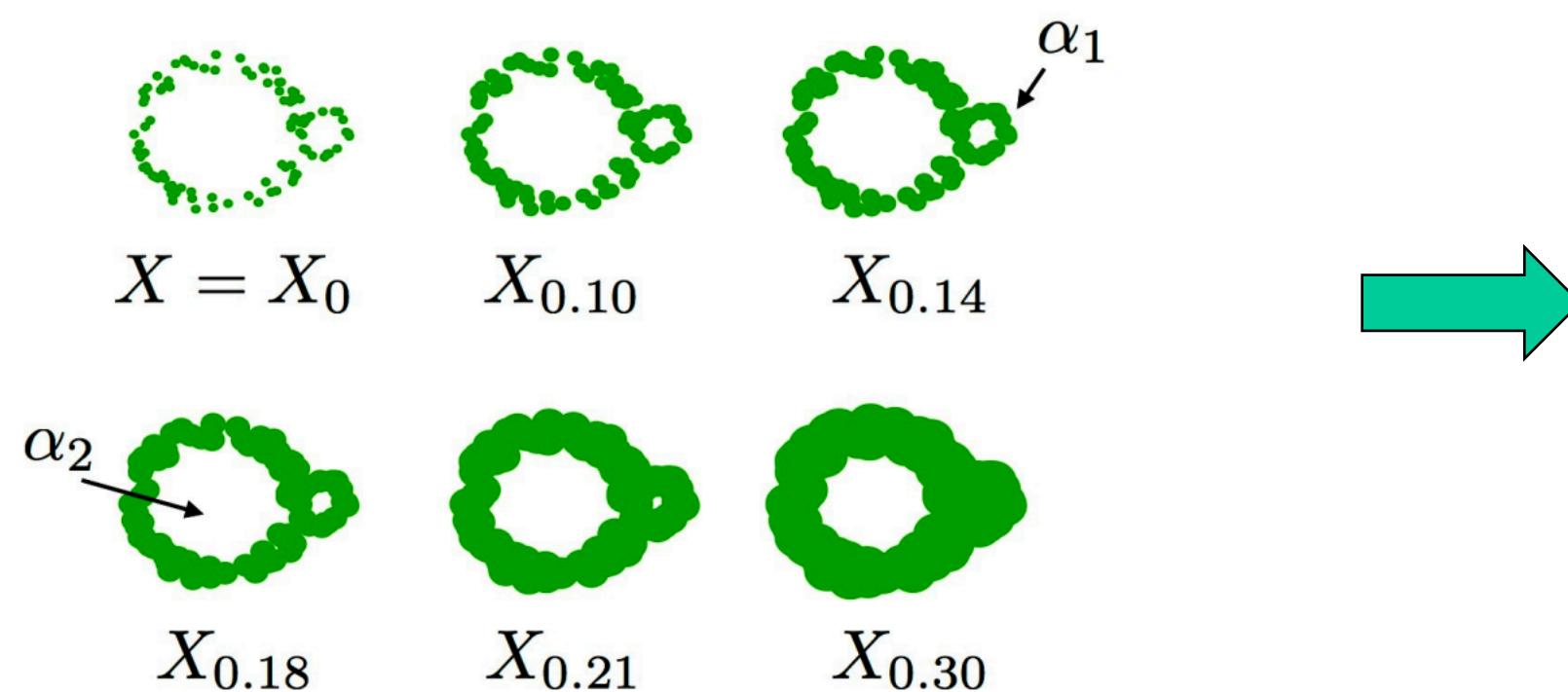
(ISM・福水健次先生との共同研究)

□ Topological Data Analysis

- Extract the geometrical and topological information of complex data set.
- High dimensional data -> finite sample (point cloud) -> Topological structure



- Persistent Diagram* (Birth-Death time)



□ Probability Measure on PD and Kernel Embedding

- Statistical inference on PD needs probability measures that support means, expectation, covariance...
- Wasserstein distance

$$W_p(d_1, d_2) = \left(\inf_{\gamma} \sum_{x \in d_1} \|x - \gamma(x)\|_{\infty}^p \right)^{\frac{1}{p}}$$

- Space of persistence diagrams

$$D_p = \{d \mid W_p(d, d_0) < \infty\} = \{d \mid \text{Pers}_p(d) < \infty\}$$

- Frechet expectation

$$\text{Var}_{\mathcal{P}} = \inf_{d \in D_p} \left[F_{\mathcal{P}}(d) = \int_{D_p} W_p(d, e)^2 d\mathcal{P}(e) < \infty \right]$$

- Or kernel way, PD can be embedded into RKHS by Bocher integral (vectorization)

$$E_k : M_b(\Omega) \rightarrow \mathcal{H}_k, \quad \mu \mapsto \int k(\cdot, x) d\mu(x)$$

- Uniqueness

If k is C_0 -universal, the mapping E_k is injective.

$$d_k(\mu, \nu) = \|E_k(\mu) - E_k(\nu)\|_{\mathcal{H}_k}$$

- defines a distance on M_b

□ Persistence weighted Gaussian kernel

- Distance measure

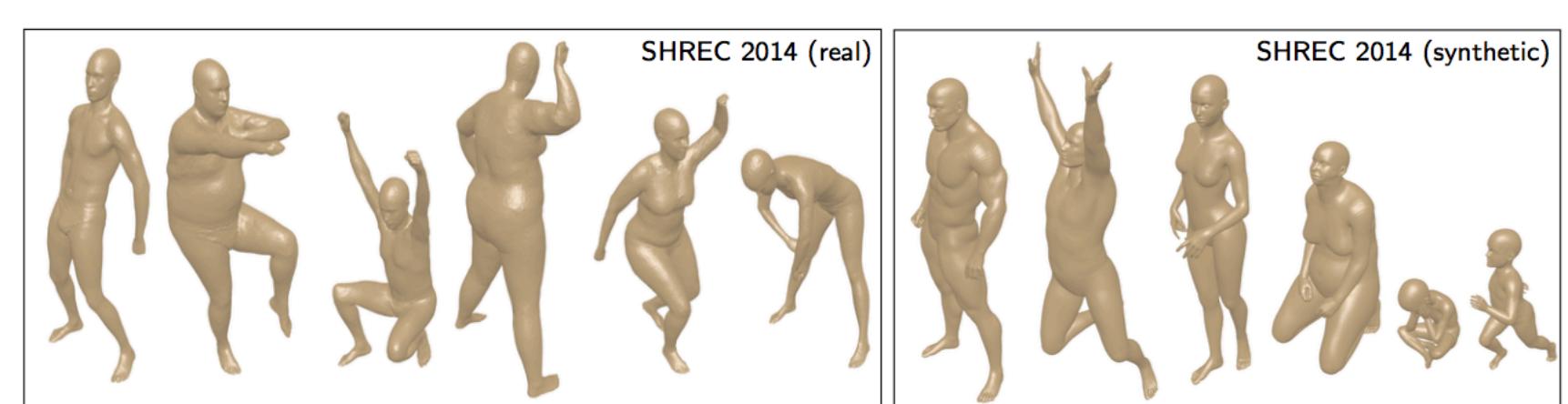
$$\begin{aligned} K_L(D, E) &= \langle E_{k_G}(\mu_D^{w_{\text{arc}}}), E_{k_G}(\mu_E^{w_{\text{arc}}}) \rangle_{\mathcal{H}_{k_G}} \\ &= \sum_{x \in D} \sum_{y \in E} w_{\text{arc}}(x) w_{\text{arc}}(y) k_G(x, y) \end{aligned}$$

- For $k(\cdot, \cdot)$, Gaussian kernel is used.
- Weight is defined by

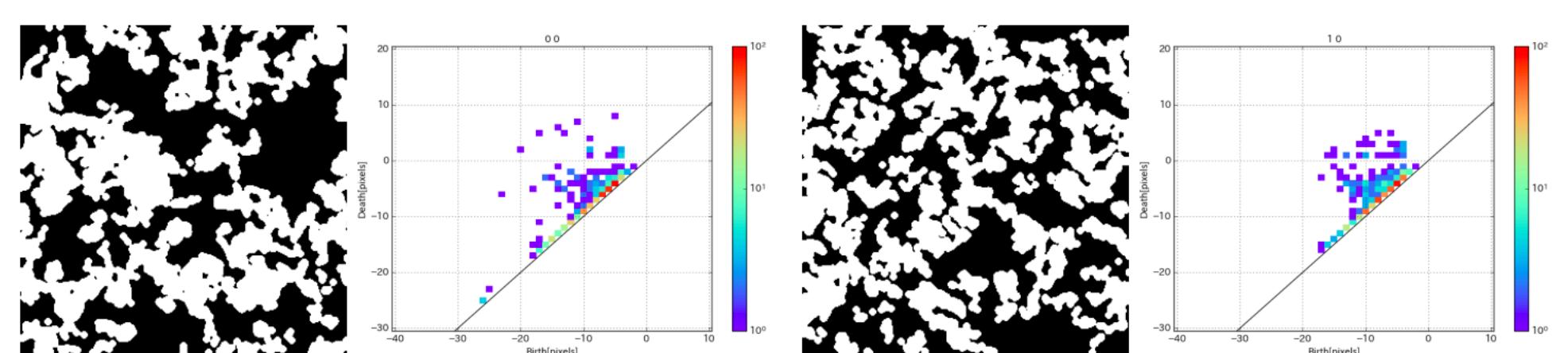
$$w_{\text{arc}}(x) = \arctan(C \text{pers}(x)^p) \quad (C, p > 0)$$

□ Applications

- Computer vision



- Material Science



Reference: 1.Persistence weighted Gaussian kernel for topological data analysis. Genki Kusano, Kenji Fukumizu, Yasuaki Hiraoka

*Images are taken from the above publication.

2.Probability measures on the space of persistence diagrams. Yuriy Mileyko, Sayan Mukherjee and John Harer