

# CS 6170: Computational Topology, Spring 2019

## Lecture 16

Topological Data Analysis for Data Scientists

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## Review on Kernel Methods

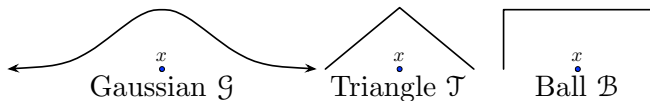
# An information introduction to kernel

- Informally, a *kernel*  $K$  is a (nonnegative) similarity measure between a pair of points in  $\mathbb{R}^d$ , where more similar points have higher value:

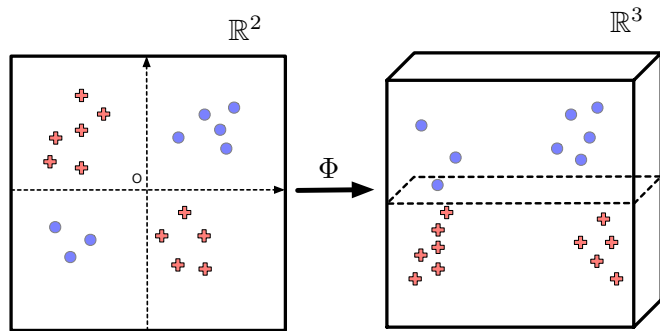
$$K : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}^+.$$

- Example: Gaussian kernel (positive definite)

$$K(p, x) = \exp(-\|p - x\|^2 / 2\sigma^2)$$



# Kernels and feature space: the kernel trick

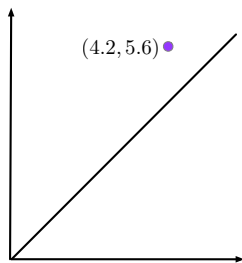
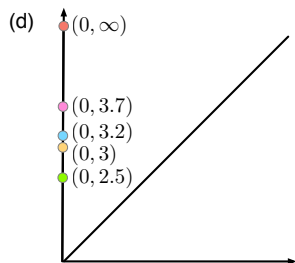
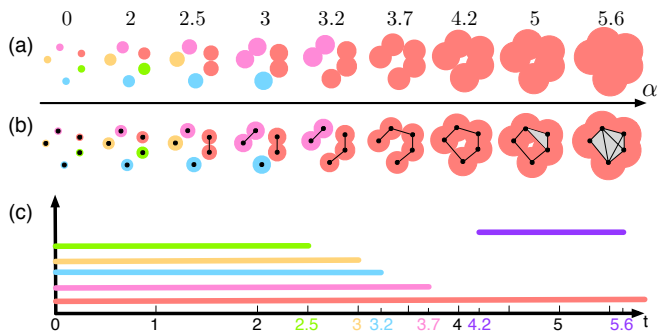


- There is no linear classifier in  $\mathbb{R}^2$
- Map points to a higher dimensional feature space such that there is a linear classifier.
- $\Phi(x) = [x_1 x_2 x_1 x_2] \in \mathbb{R}^3$

## Stability of Persistence Diagrams

Edelsbrunner and Harer (2010), C.VIII

# Barcode and persistence diagram



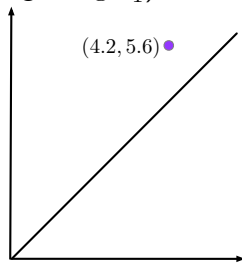
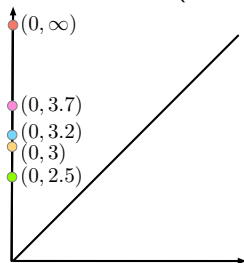


# Persistence diagram

## Definition

A *persistence diagram* is a finite multisite of points in the extended plane. To simplify the definitions and results, we add infinitely many points on the diagonal, each with infinite multiplicity.

- The extended plane:  $\bar{\mathbb{R}}^2 = (\mathbb{R} \cup \pm\infty)^2$ .
- Let  $\delta$  denote diagonal of the extended plane.
- For example see persistence diagrams in dimension 0 and dimension 1 below (left,  $\text{Dgm}_0$ ; right  $\text{Dgm}_1$ ).





- Let  $X, Y$  be two persistence diagrams
- $L_\infty$  norm: let  $x = (x_1, x_2) \in X$  and  $y = (y_1, y_2) \in Y$ , then

$$\|x - y\|_\infty := \max\{|x_1 - y_1|, |x_2 - y_2|\}.$$

- Let  $\eta : X \rightarrow Y$  be a bijection.

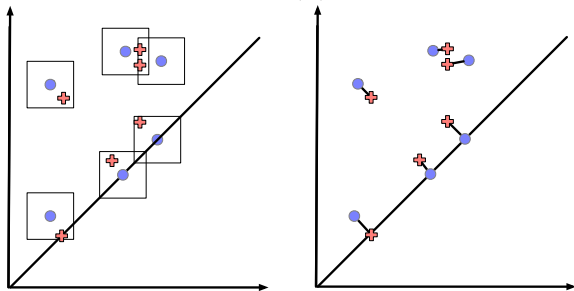
## Definition

The *bottleneck distance* between two persistence diagrams  $X$  and  $Y$  is defined to be

$$W_\infty(X, Y) = \inf_{\eta: X \rightarrow Y} \sup_{x \in X} \|x - \eta(x)\|_\infty.$$

# Bottleneck distance: an illustration

- Let  $\epsilon = W_\infty(X, Y)$
- Draw squares centered at  $x \in Y$  with  $2\epsilon$  sides such that each square contains its corresponding point  $\eta(x) \in Y$ .



# Bottleneck Stability for Filtrations

- $K$ : simplicial complex
- $f : K \rightarrow \mathbb{R}$  is a *monotonic function* on  $K$ , if for every  $\sigma < \tau$ ,  $f(\sigma) \leq f(\tau)$ .
- Let  $f, g : K \rightarrow \mathbb{R}$  be two monotonic functions, define

$$\|f - g\|_\infty = \sup_{\sigma \in K} |f(\sigma) - g(\sigma)|.$$

## Theorem (Stability Theorem for Filtrations)

*Let  $K$  be a simplicial complex and  $f, g : K \rightarrow \mathbb{R}$  two monotonic functions. For each dimension  $p$ , the bottleneck distance between the diagrams  $X = \text{Dgm}_p(f)$  and  $Y = \text{Dgm}_p(g)$  is bounded from above by the  $L_\infty$  distance between the functions (Edelsbrunner and Harer, 2010, Page 182), that is,*

$$W_\infty(X, Y) \leq \|f - g\|_\infty.$$

Edelsbrunner, H. and Harer, J. (2010). *Computational Topology: An Introduction*. American Mathematical Society, Providence, RI, USA.