

7th Italian-Japanese summer school in Mathematics

Persistent Homology:

A roadmap in Topological Data Analysis

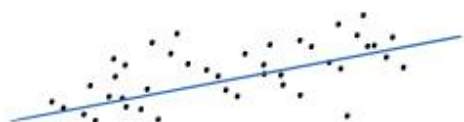
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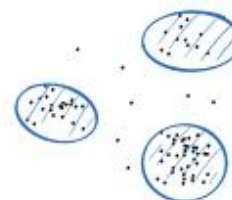
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Why topology for data analysis?

Topological data analysis (TDA) is a branch of data science that applies topology to extract hidden structures (*shape*) in data.



Correlation Methods



Clustering Methods

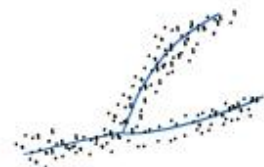
Premise of Topological Data Analysis

Data is noisy sample of some subspace in $\mathbb{R}^n$.

The “topology” of this subspace captures abstract (cor)relations in the data.



Loops
(and higher dim. generalizations)



Bifurcations

When is topology useful?

Many real datasets lie on or near nonlinear manifolds:

- periodic signals $\rightarrow$ circles
- motion trajectories $\rightarrow$ tori
- cell development $\rightarrow$ branching trees
- image and pose datasets $\rightarrow$ low-dimensional manifolds

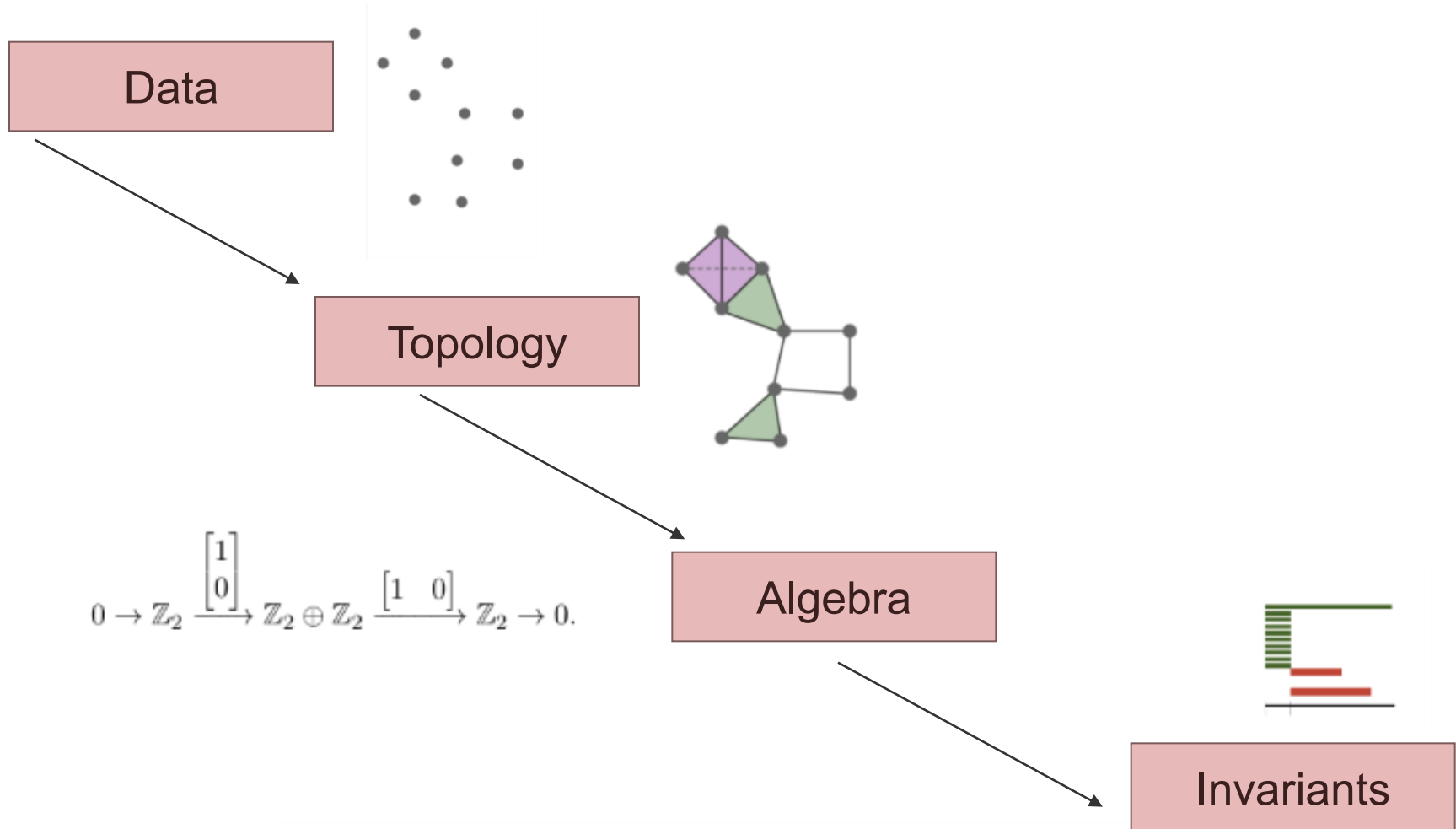
Traditional methods (PCA, clustering, regression) typically assume a linear structure.

TDA instead asks: How many connected pieces? How many loops? Any voids?

The goal is “seeing shape in data”

What will the course cover?

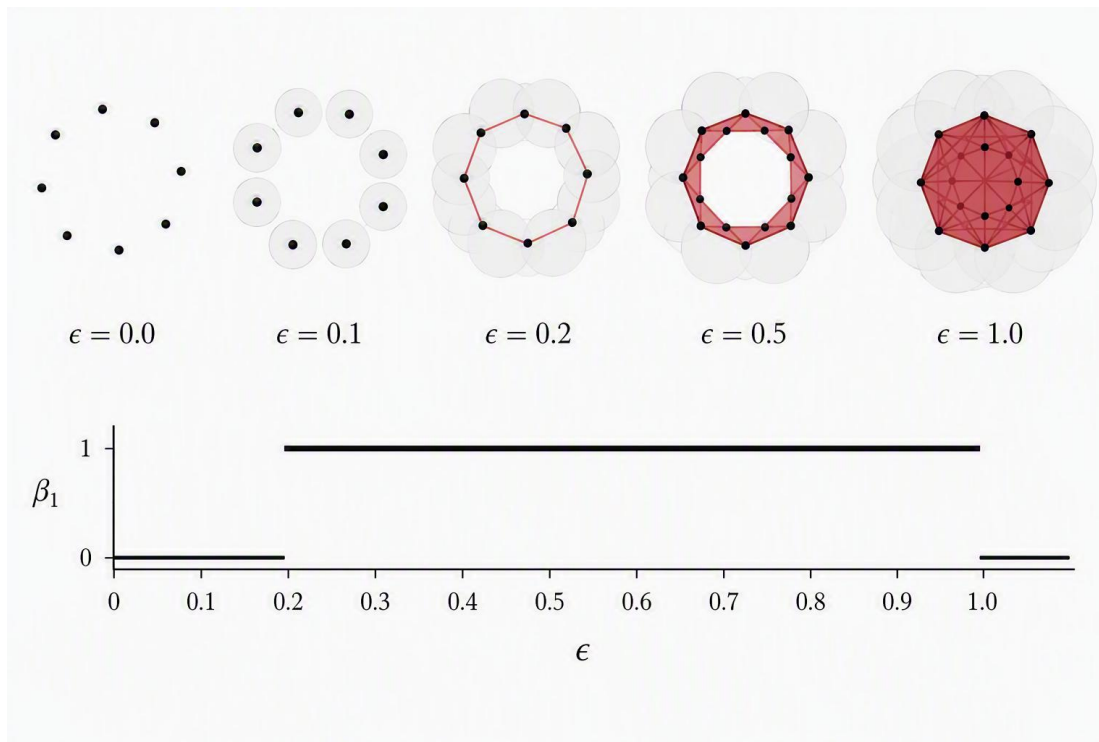
The whole **Persistent Homology** pipeline ...



What will the course cover?

... and its properties:

- Stability with respect to input data perturbation
- Topological inference



Syllabus and textbook

17/9: From data to simplicial complexes

- simplicial complexes, Čech and Vietoris–Rips complexes

18/9: From simplicial complexes to algebra:

- simplicial homology, filtrations, persistence modules

24/9: From persistence modules to invariants

- Barcodes, persistence modules

25/9: persistent homology stability, topological inference

Reference:

M.B. Botnan: Topological Data Analysis Mastermath, [lecture notes.pdf](#)