

Geometric and Topological
Methods in Machine Learning
2021-2022

Computational Topology: Simplicial Complexes and (Persistent) Homology

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Class outline: taking a step back...

My classes are about

Topological Data Analysis (TDA)

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Goal: Study geometric data sets with techniques coming from *topology*.

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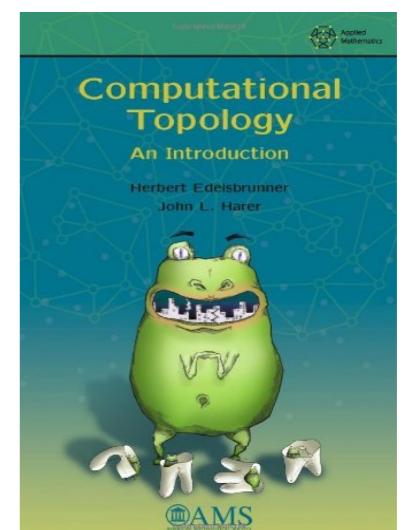
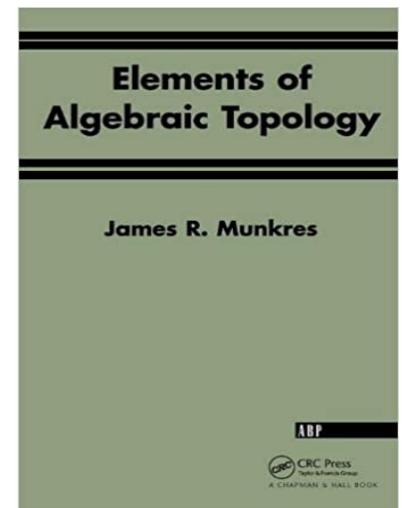
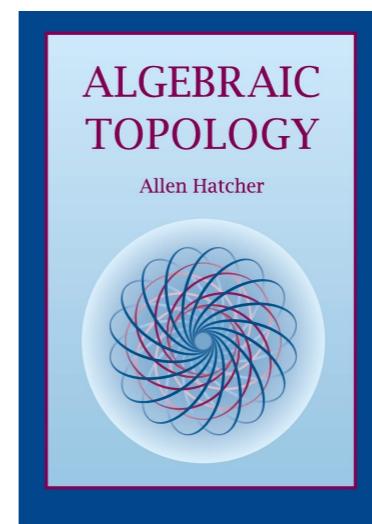
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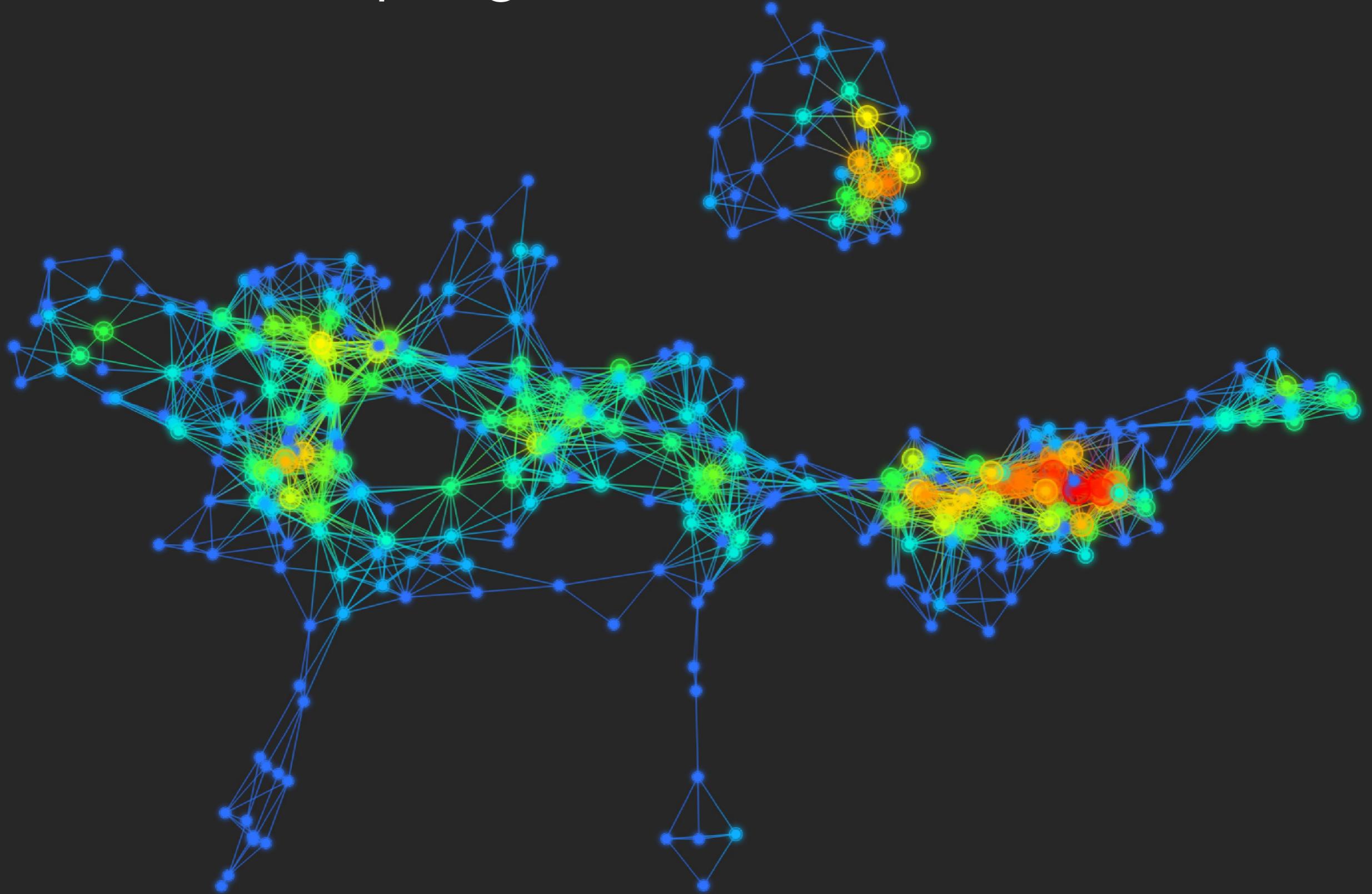
[*Elements of Algebraic Topology*,
Munkres, CRC Press, 1984]

[*Algebraic Topology*, Hatcher, Cambridge University Press, 2002]

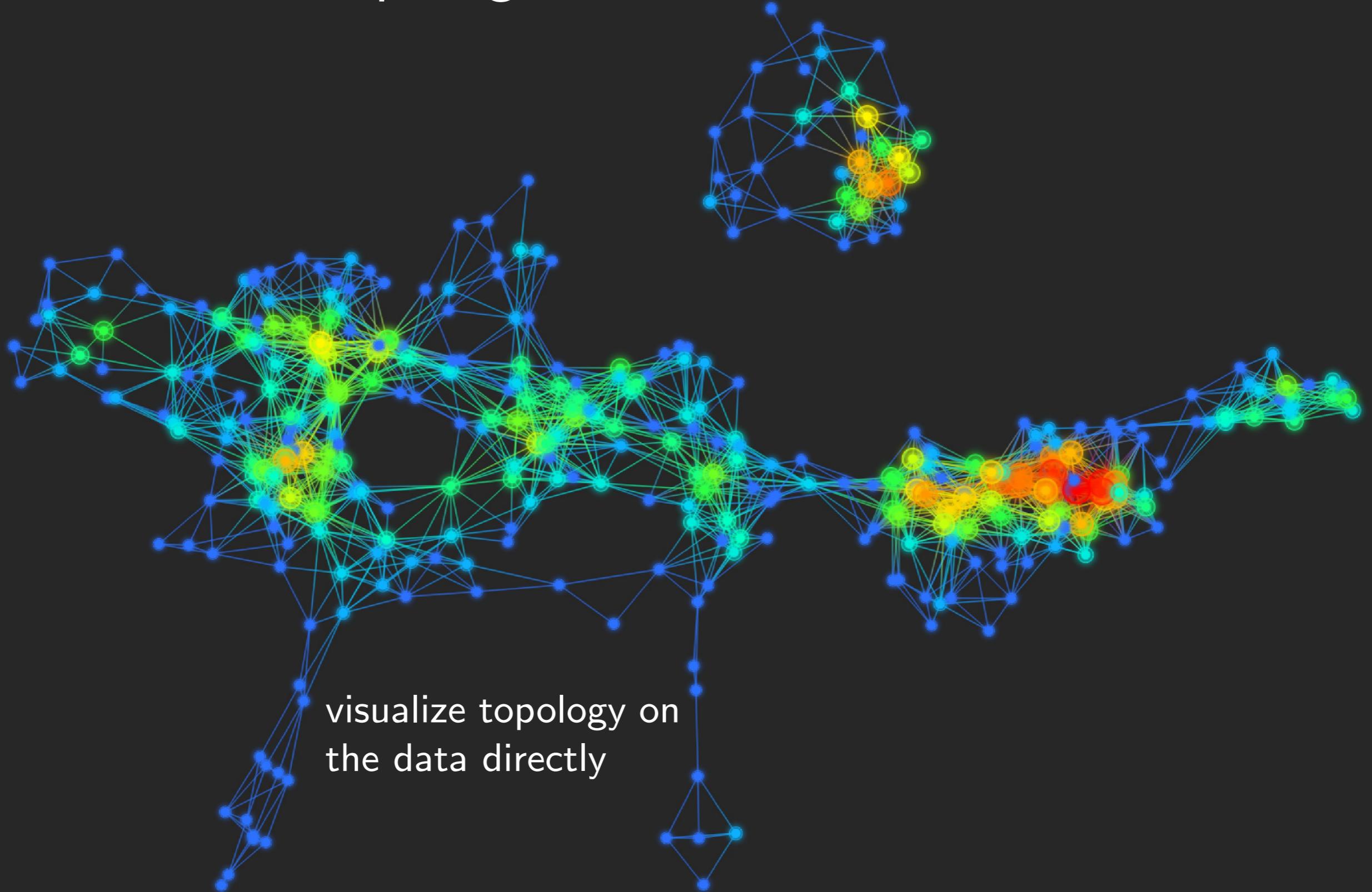
[*Computational Topology: an introduction*, Edelsbrunner, Harer, AMS, 2010]



Introduction: topological visualization



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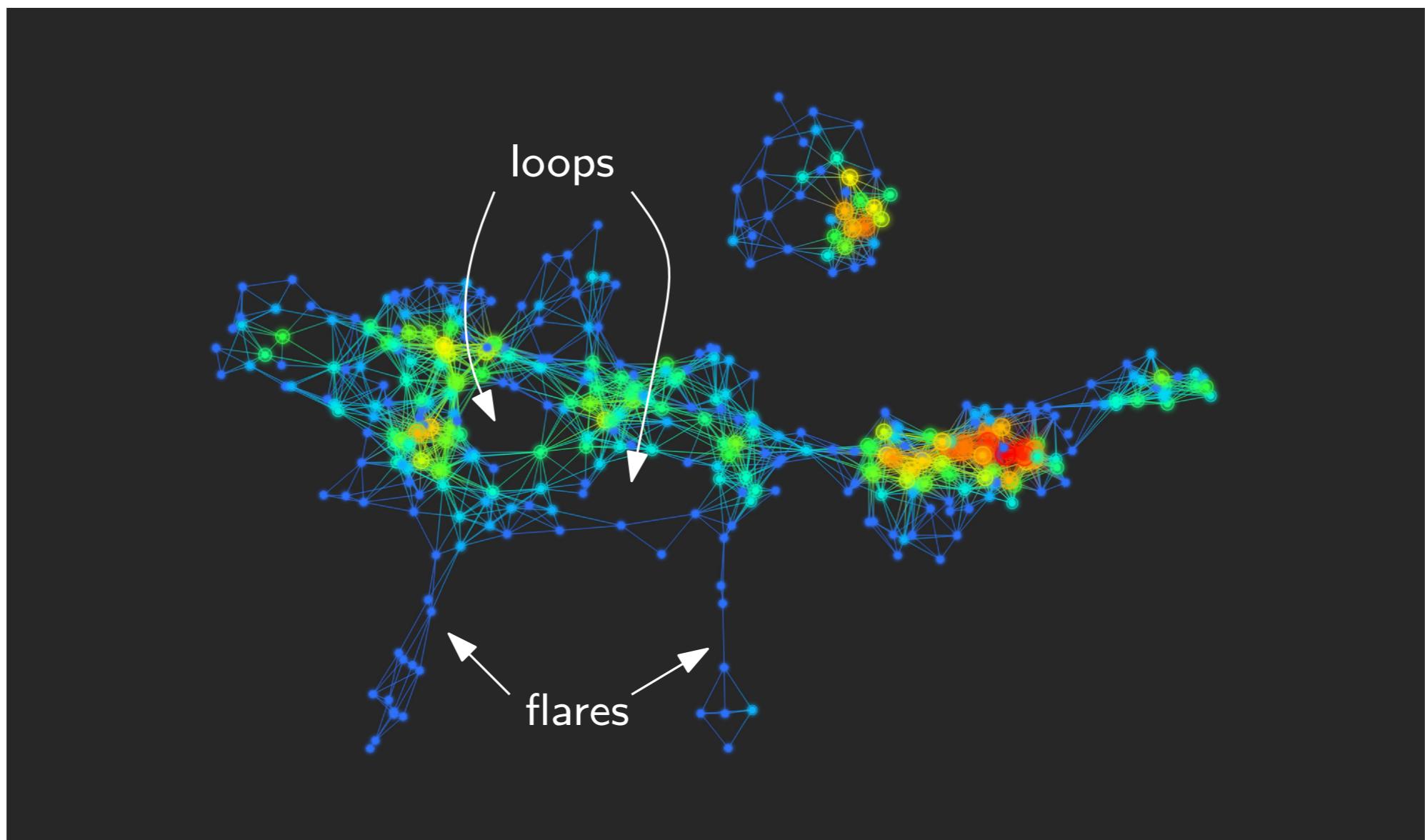


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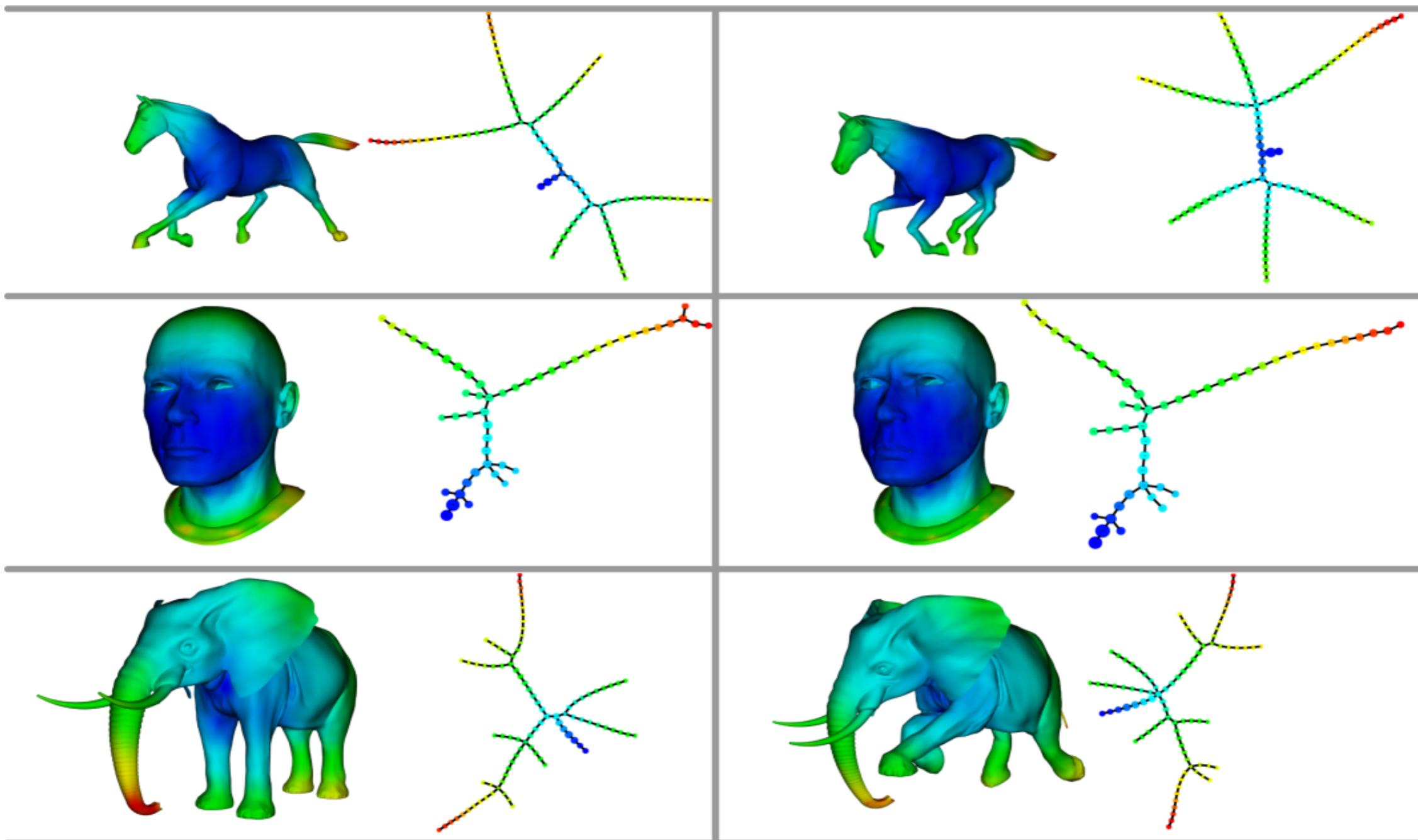
Two types of applications:

- clustering
- feature selection

Principle: identify statistically relevant subpopulations through **topological patterns** (flares, loops).



Introduction: topological visualization



3d shapes classification

[*Topological Methods for the Analysis of High Dimensional Data Sets and 3D Object Recognition*, Singh, Mémoli, Carlsson, Symp. Point based Graphics, 2007]

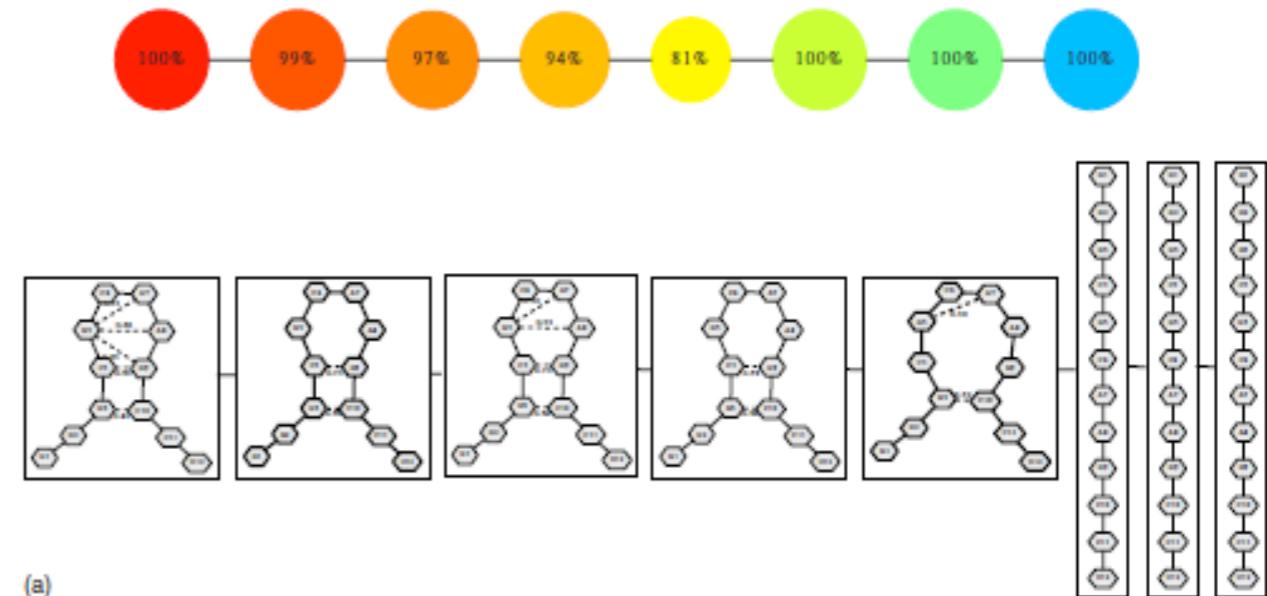
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[*Topological Methods for Exploring Low-density States in Biomolecular Folding Pathways*, Yao et al., J. Chemical Physics, 2009]

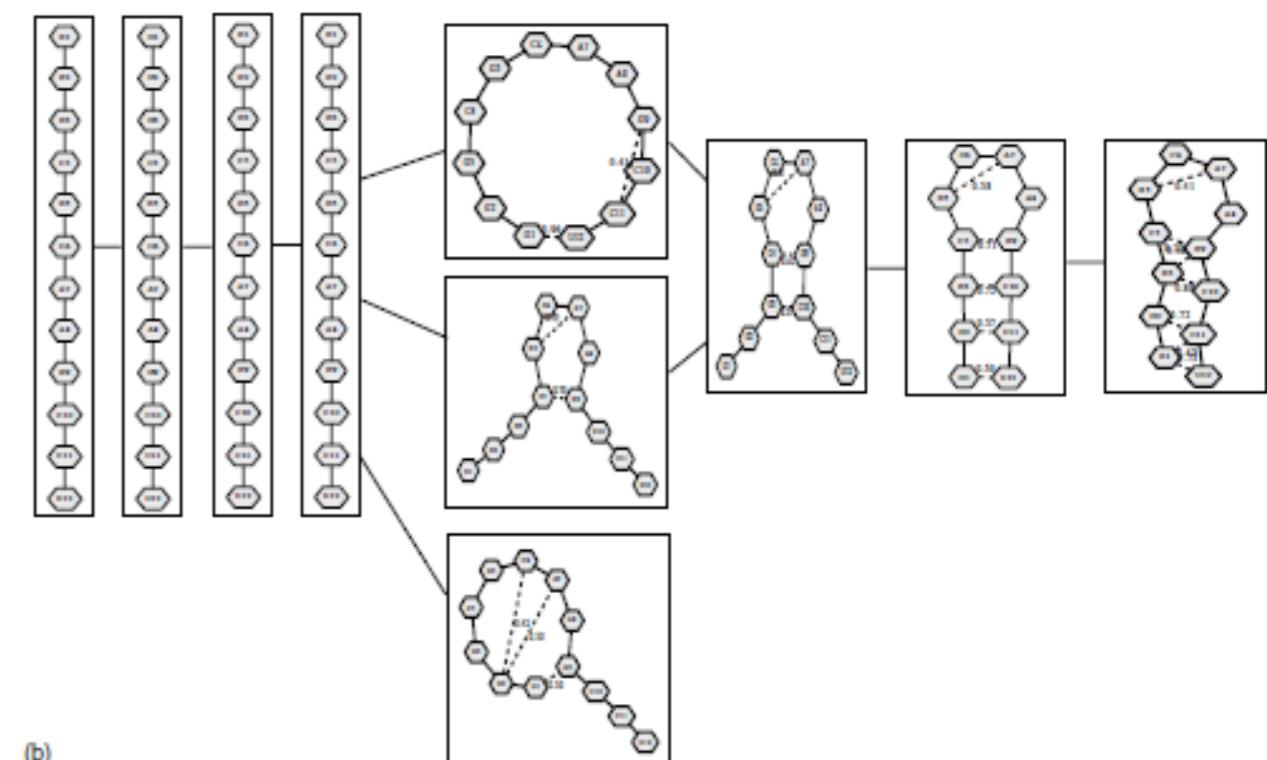
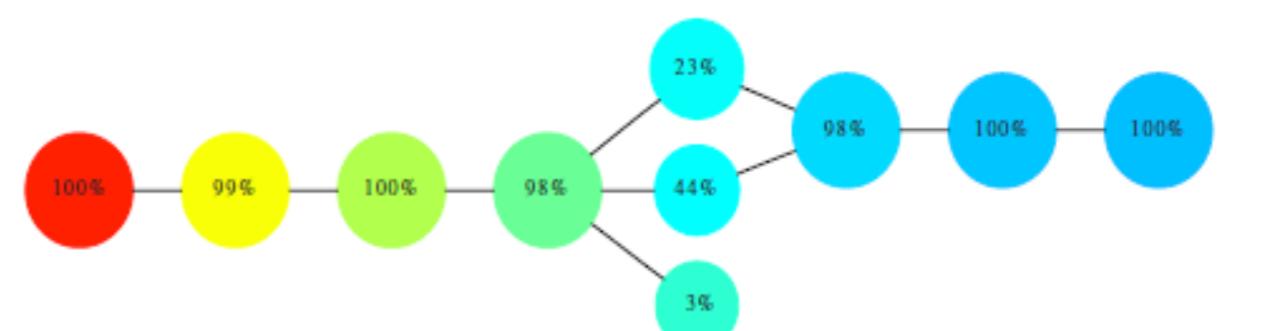
Data: conformations of molecules.

Goal: detect folding pathways.

Idea: 1 loop = 2 pathways.



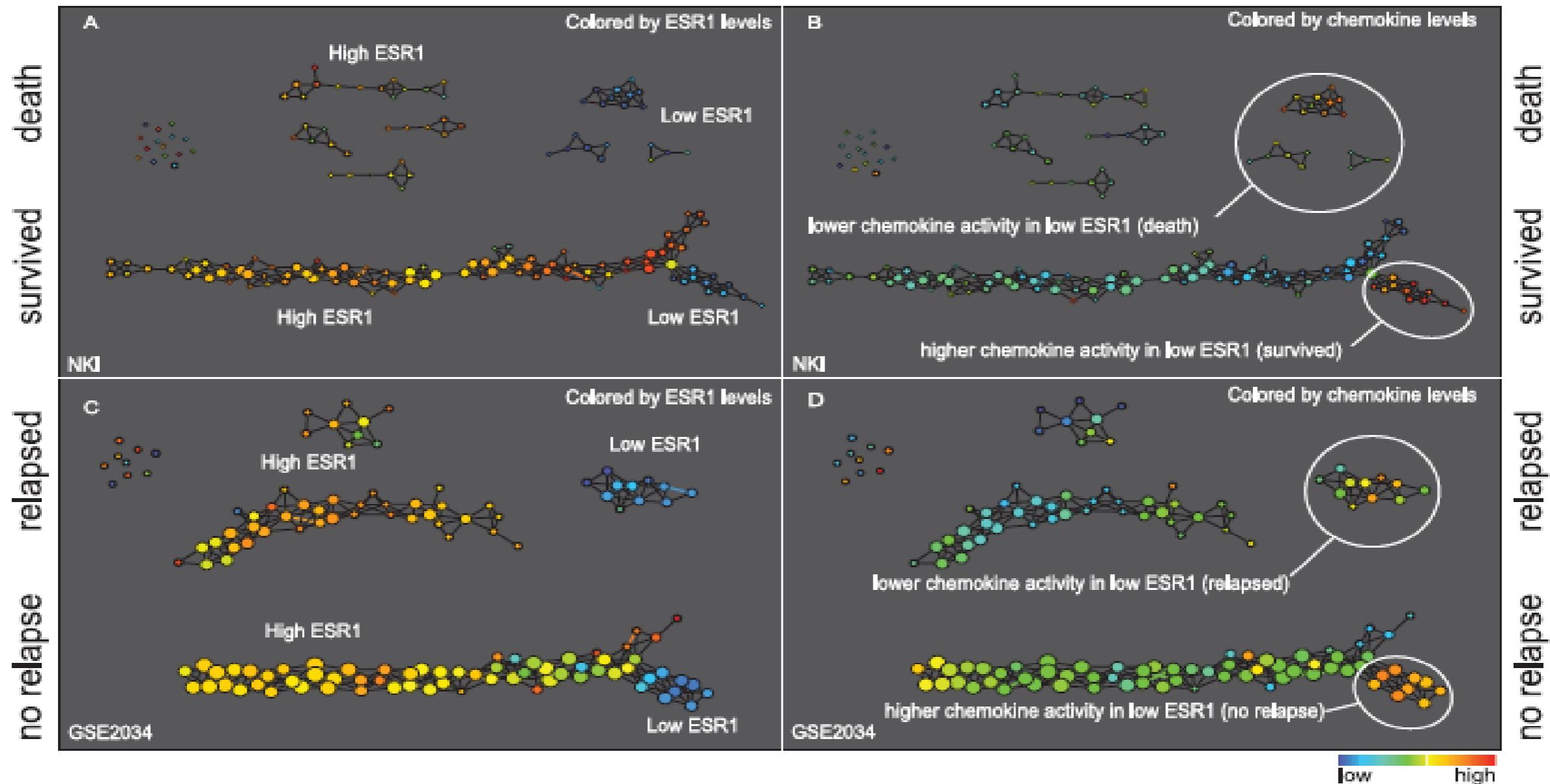
(a)



(b)

Introduction: topological visualization

[Extracting insights from the shape of complex data using topology, Lum et al., Nature, 2013]



Data: breast cancer patients that went through specific therapy.

Goal: detect variables that influence survival after therapy in breast cancer.

Introduction: topological descriptors built from data

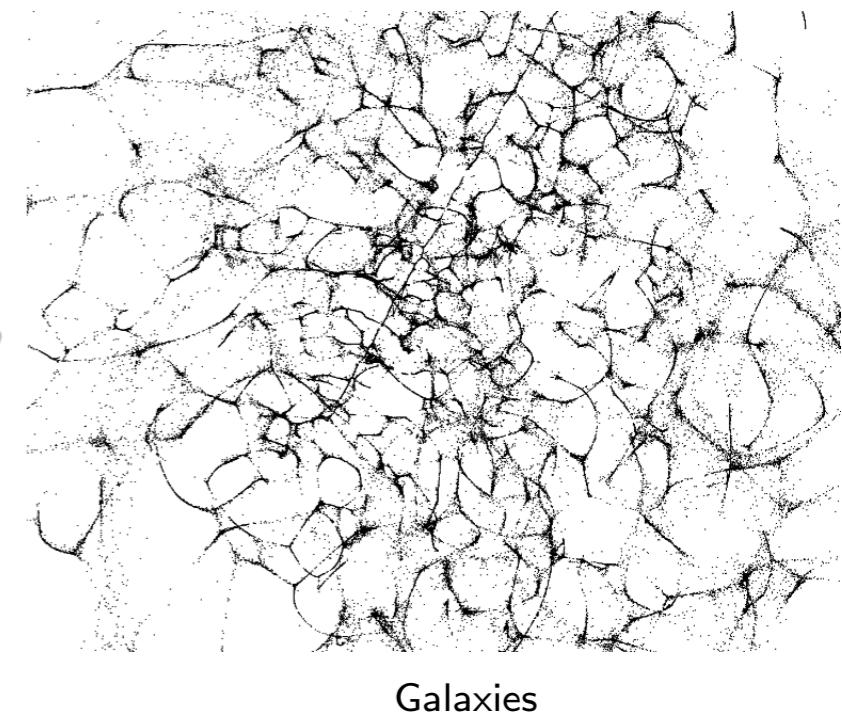
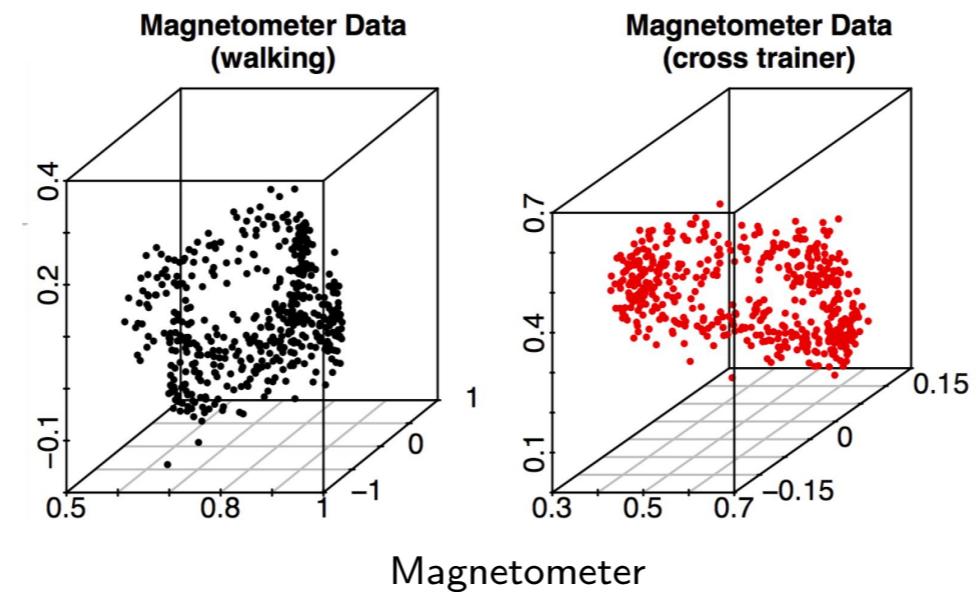
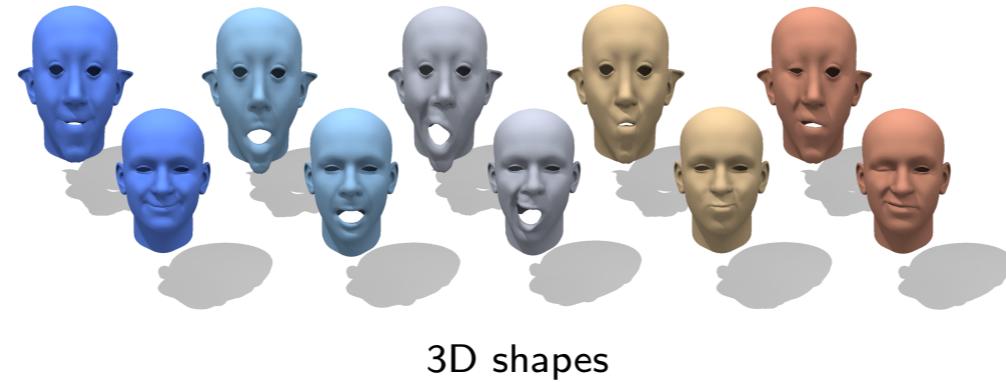
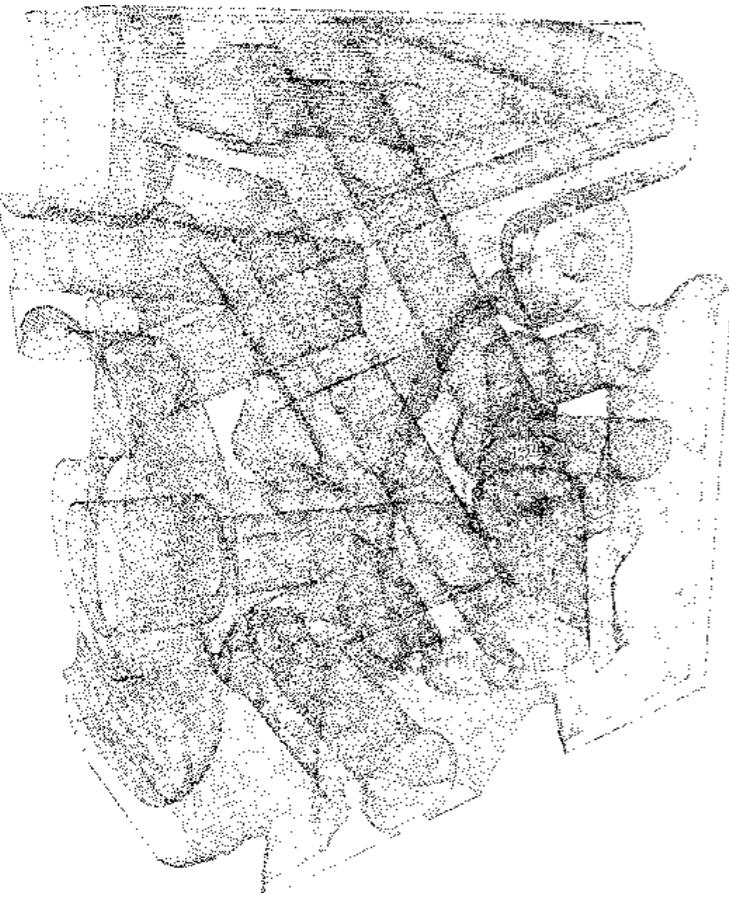
We will see how to build new topological features from data sets...

Introduction: topological descriptors built from data

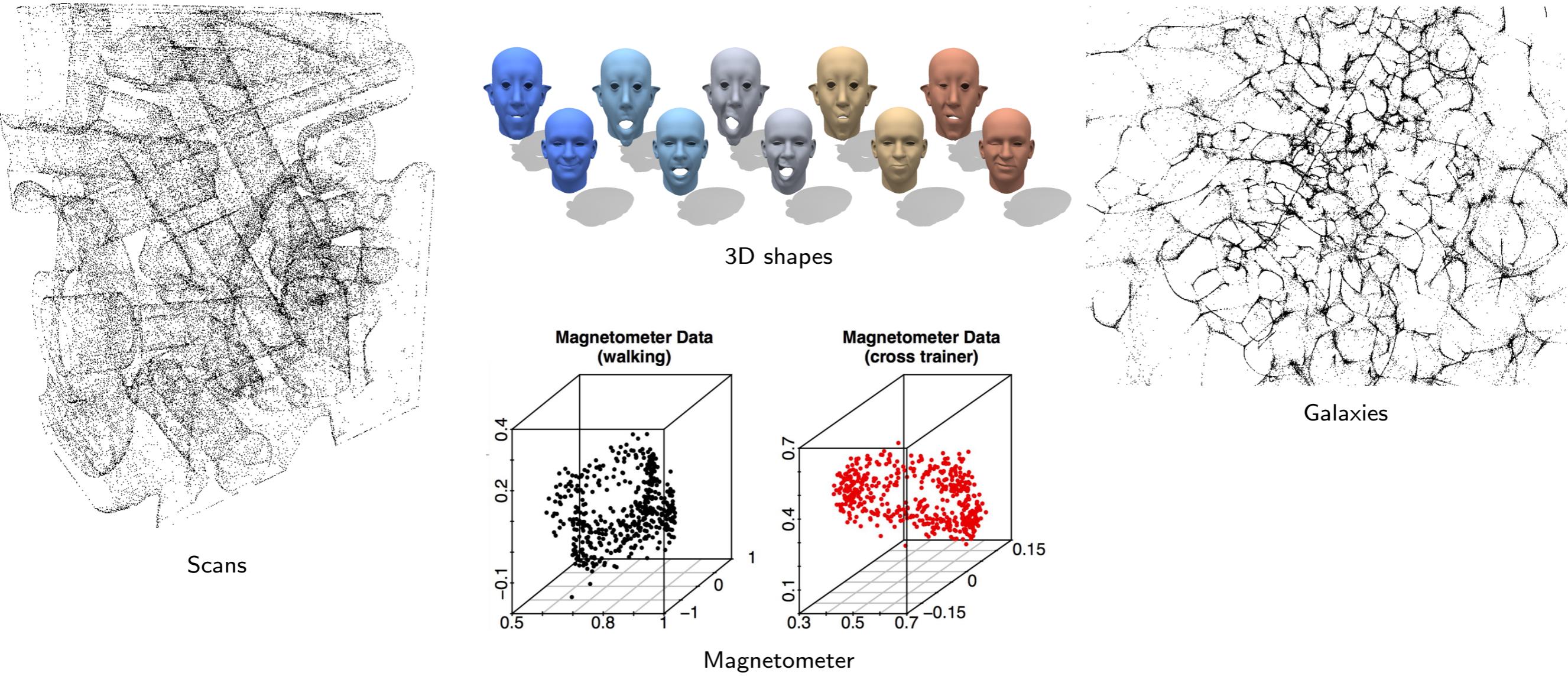
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...but why is that interesting?

Introduction: topological descriptors built from data



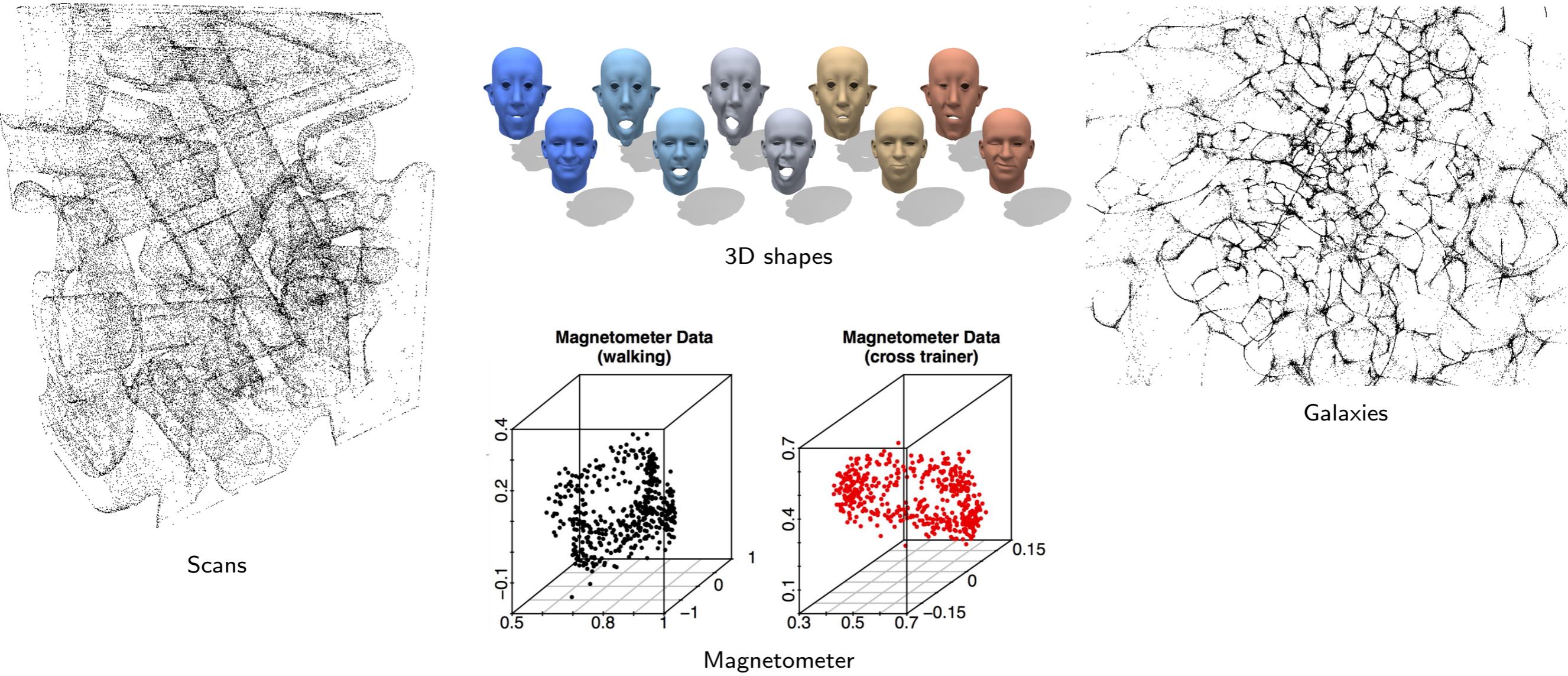
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Data often come as (sampling of) metric spaces or sets/spaces endowed with a similarity measure with, possibly complex, topological/geometric structure.

Data carrying geometric information is usually high dimensional.

Introduction: topological descriptors built from data

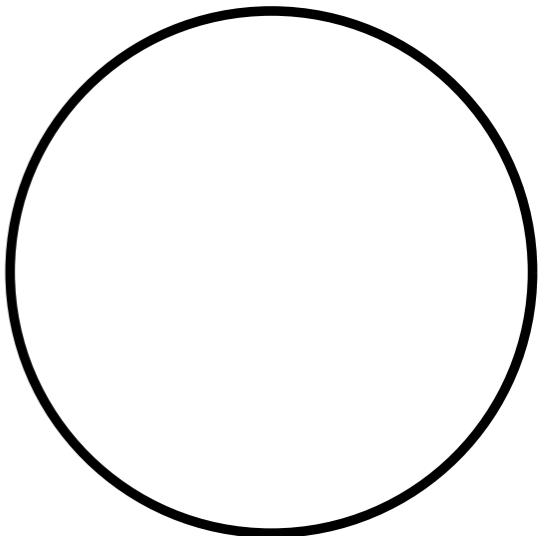
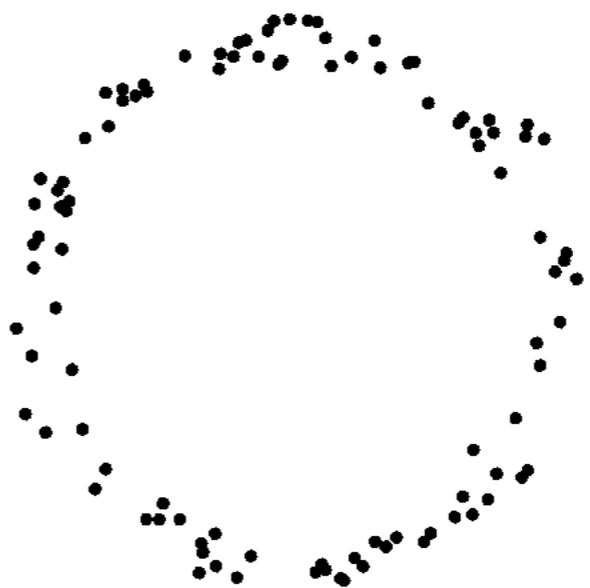


Features from [Topological Data Analysis](#) allow to:

- infer relevant topological and geometric features of these spaces.
- take advantage of topol./geom. information for further processing of data (classification, recognition, learning, clustering, parametrization...).

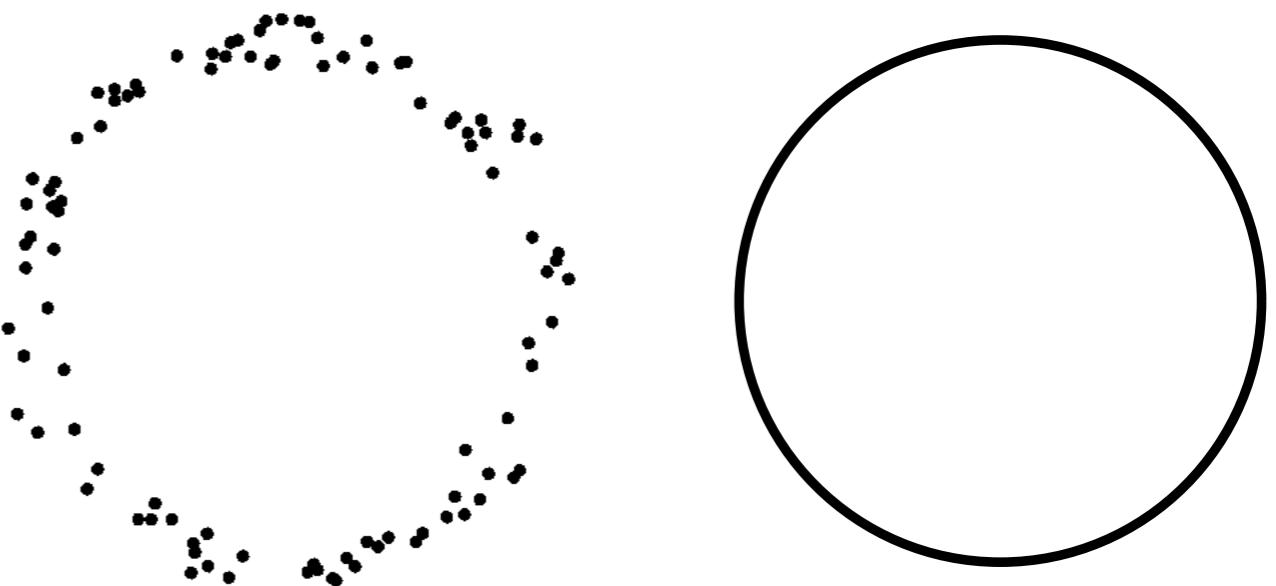
Challenges and advantages

Problem: how to define the *topology* of a data set?



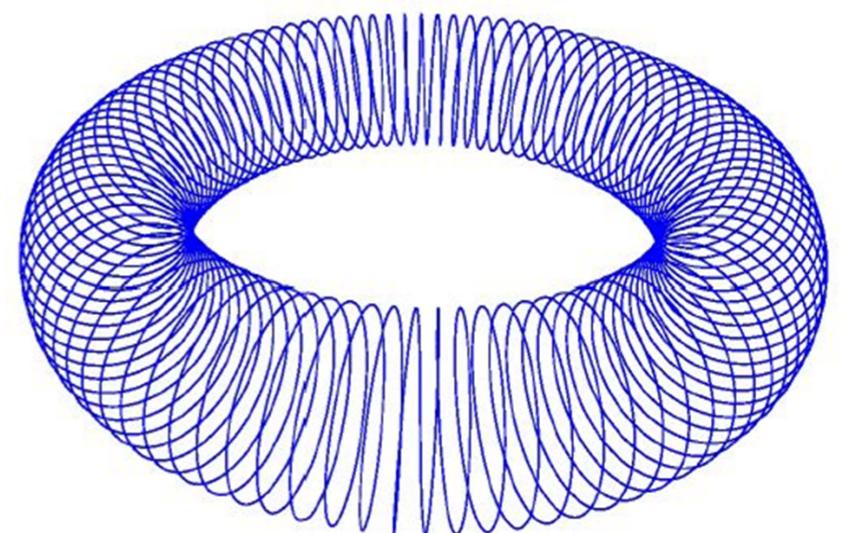
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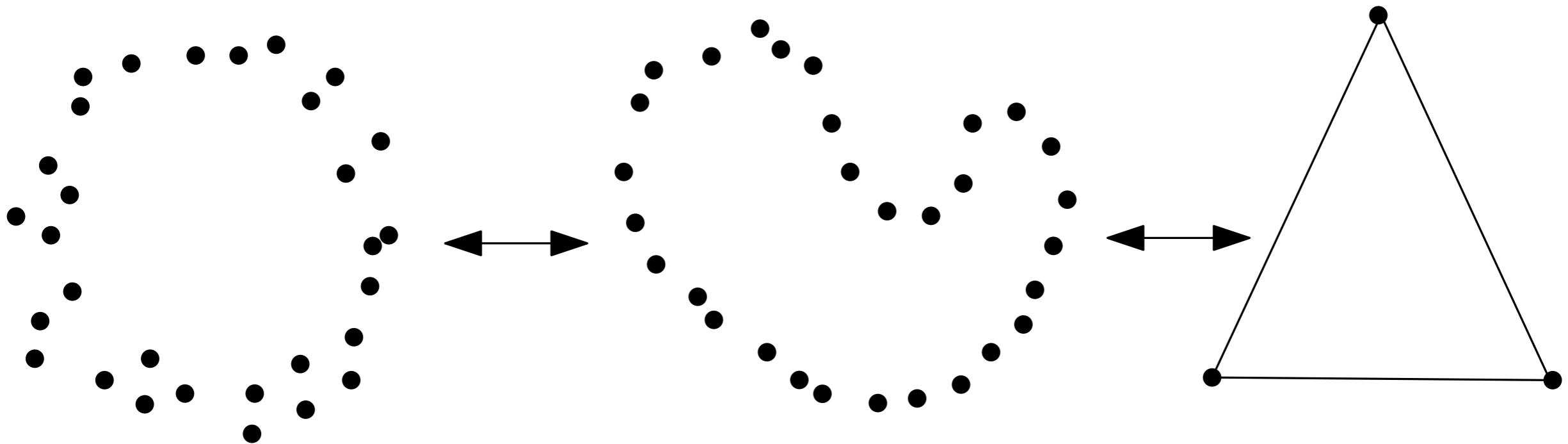


Challenges and goals:

- no direct access to topological/geometric information: need of intermediate constructions with *simplcial complexes*;
- distinguish topological “signal” from noise;
- topological information may be multiscale;
- statistical analysis of topological information.



Challenges and advantages



Advantages:

- **coordinate invariance:** topological features/invariants do not rely on any coordinate system ⇒ no need to have data with coordinates, or to embed data in spaces with coordinates... but the metric (distance/similarity between data points) is important.
- **deformation invariance:** topological features are invariant under homeomorphism and reparameterization.
- **compressed representation:** topology offers a set of tools to summarize the data in compact ways while preserving its topological structure.

A brief look at topology

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A: The so-called *topological spaces*.

Def: A *topological space* is a set X equipped with a *topology*, i.e., a family \mathcal{O} of subsets of X , called the *open sets* of X , such that:

- (i) the empty set \emptyset and X are elements of \mathcal{O} ,
- (ii) any union of elements of \mathcal{O} is an element of \mathcal{O} ,
- (iii) any finite intersection of elements of \mathcal{O} is an element of \mathcal{O} .

Open sets are the tools that allow to define *continuity*, which is the primary notion that allow to compare spaces in topology.

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Def: a map $f : X \rightarrow Y$ is *continuous* if and only if the pre-image $f^{-1}(O_Y) = \{x \in X : f(x) \in O_Y\}$ of any open set $O_Y \subseteq Y$ is an open set of X .

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A very common family of topological spaces is comprised of the *metric spaces*.

Def: A metric (or distance) on X is a map $d : X \times X \rightarrow [0, +\infty)$ such that:

- (i) for any $x, y \in X$, $d(x, y) = d(y, x)$,
- (ii) for any $x, y \in X$, $d(x, y) = 0$ if and only if $x = y$,
- (iii) for any $x, y, z \in X$, $d(x, z) \leq d(x, y) + d(y, z)$.

The set X together with d is a metric space.

The smallest topology containing all the open balls $B(x, r) = \{y \in X : d(x, y) < r\}$ is called the metric topology on X induced by d .

Ex: the standard topology in an Euclidean space is the one induced by the metric defined by the norm: $d(x, y) = \|x - y\|$.

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Def: Here are the main comparison tools of topology:

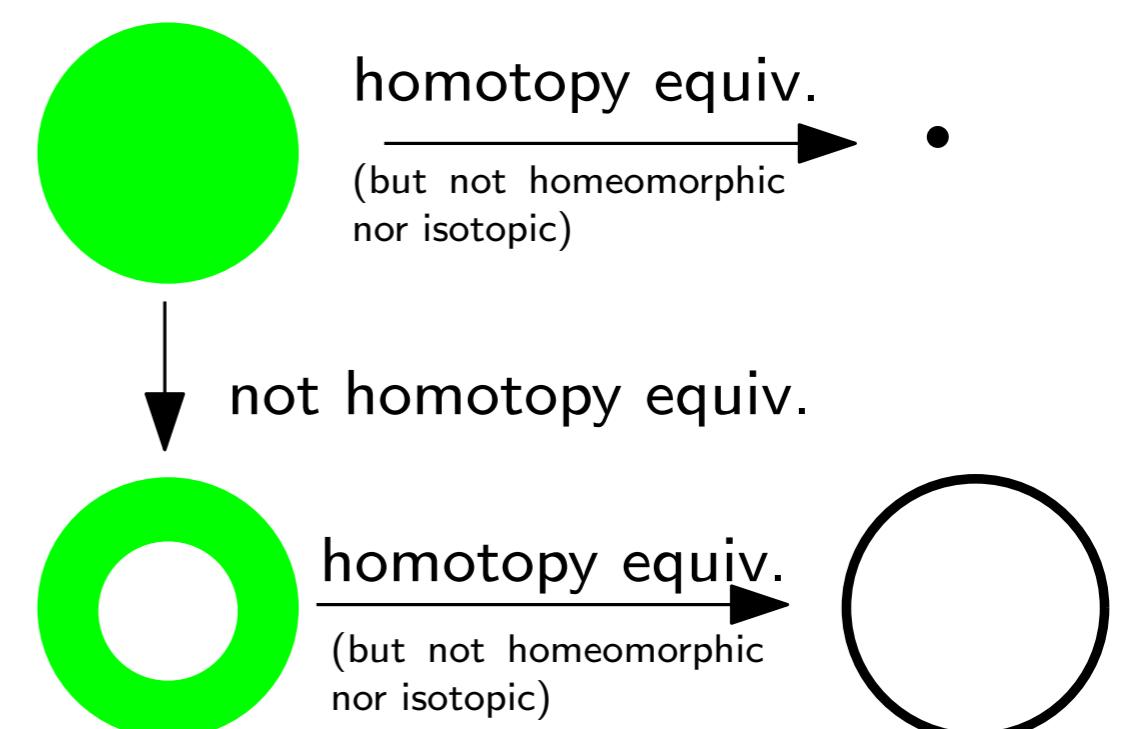
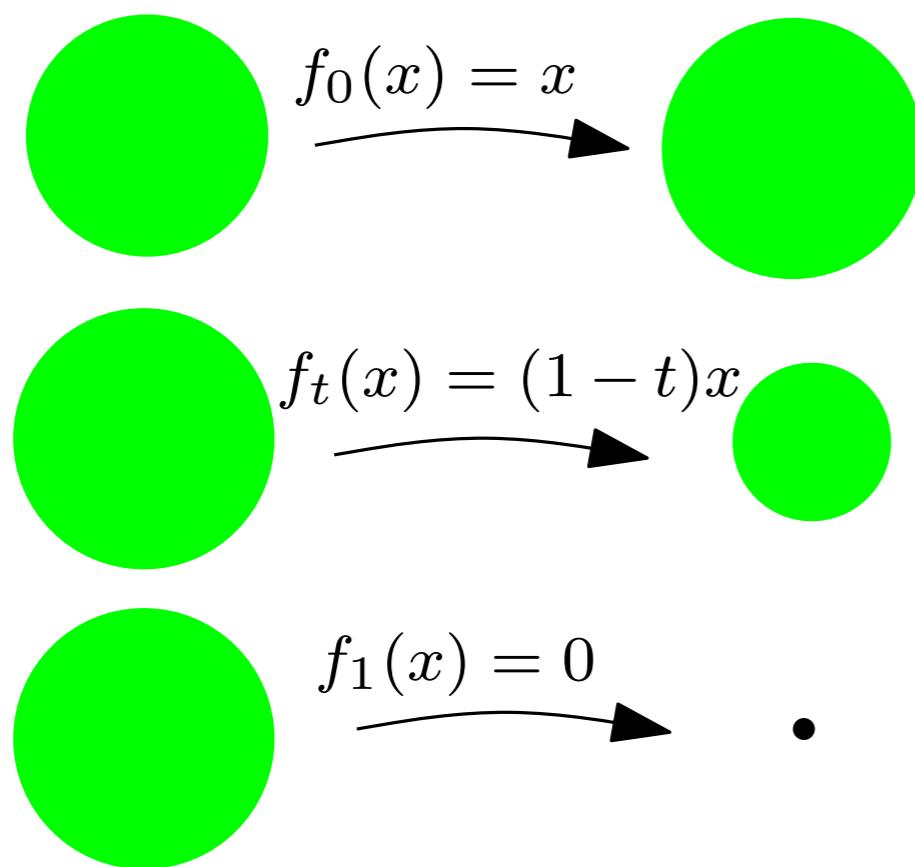
- Two maps $f_0 : X \rightarrow Y$ and $f_1 : X \rightarrow Y$ are *homotopic* if \exists a continuous map $F : [0, 1] \times X \rightarrow Y$ s.t. $\forall x \in X, F(0, x) = f_0(x)$ and $F_1(1, x) = f_1(x)$. X and Y are *homotopy equivalent* if \exists continuous maps $f : X \rightarrow Y$ and $g : Y \rightarrow X$ s.t. $g \circ f$ is homotopic to id_X and $f \circ g$ is homotopic to id_Y .
- X and Y are *homeomorphic* if \exists a bijection (homeomorphism) $h : X \rightarrow Y$ s.t. h and h^{-1} are continuous.
- X and Y are *isotopic* if \exists a continuous map (isotopy) $F : X \times [0, 1] \rightarrow Y$ s.t. $F(., 0) = \text{id}_X$, $F(X, 1) = Y$ and $\forall t \in [0, 1], F(., t)$ is an homeomorphism.

Q: Which notion is stronger/weaker?

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Previous examples are particular homotopy equivalences called *deformation retracts*.

Def: If $Y \subseteq X$ and if there exists a continuous map $F : [0, 1] \times X \rightarrow X$ s.t.:

- (i) $\forall x \in X, F(0, x) = x$
- (ii) $\forall x \in X, F(1, x) \in Y$
- (iii) $\forall y \in Y, \forall t \in [0, 1], F(t, y) \in Y$

then X and Y are homotopy equivalent. If one replaces condition (iii) by $\forall y \in Y, \forall t \in [0, 1], H(t, y) = y$ then H is a **deformation retract** of X onto Y .

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Q: Can you find two spaces that are homeomorphic but not isotopic?

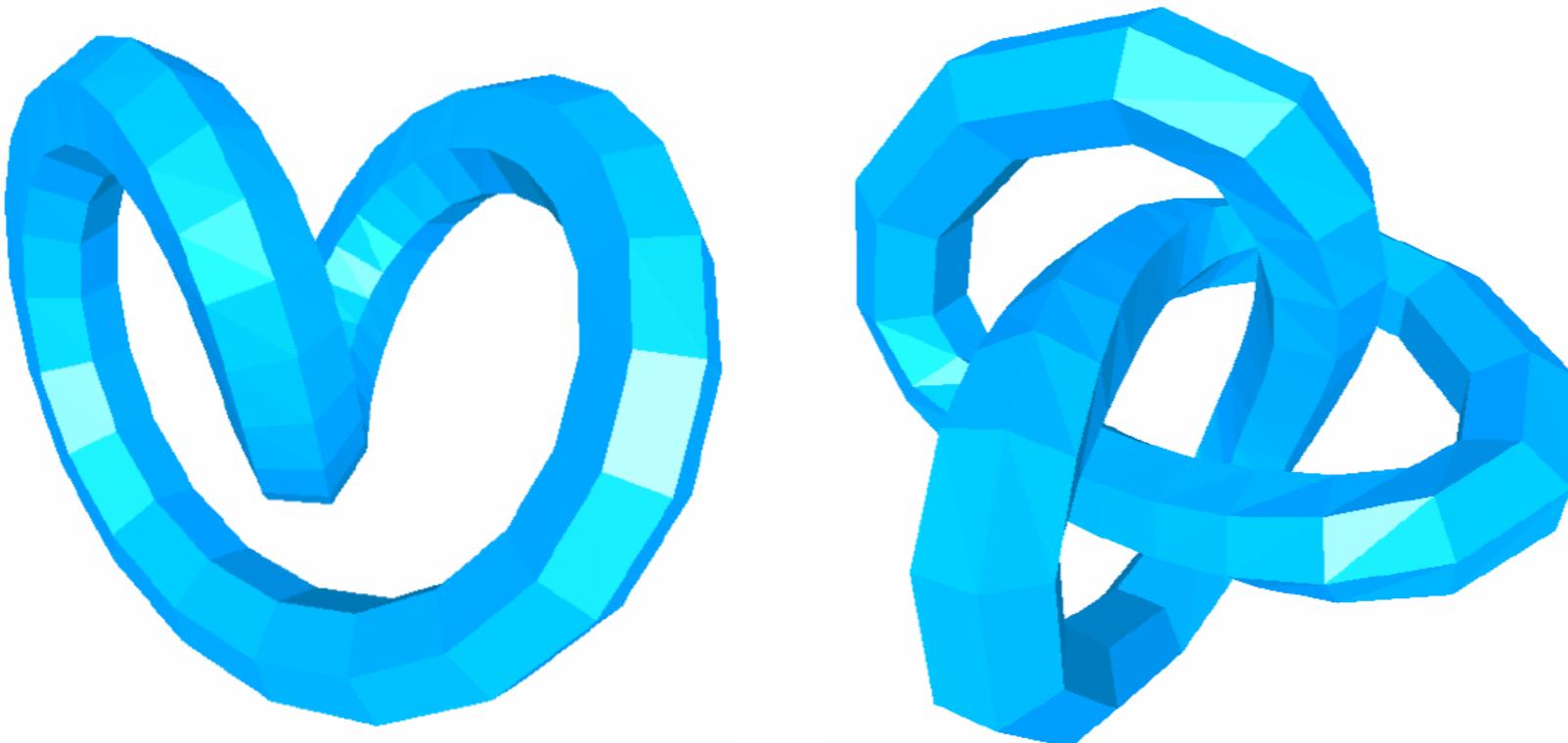
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A: Torus and trefoil knot.

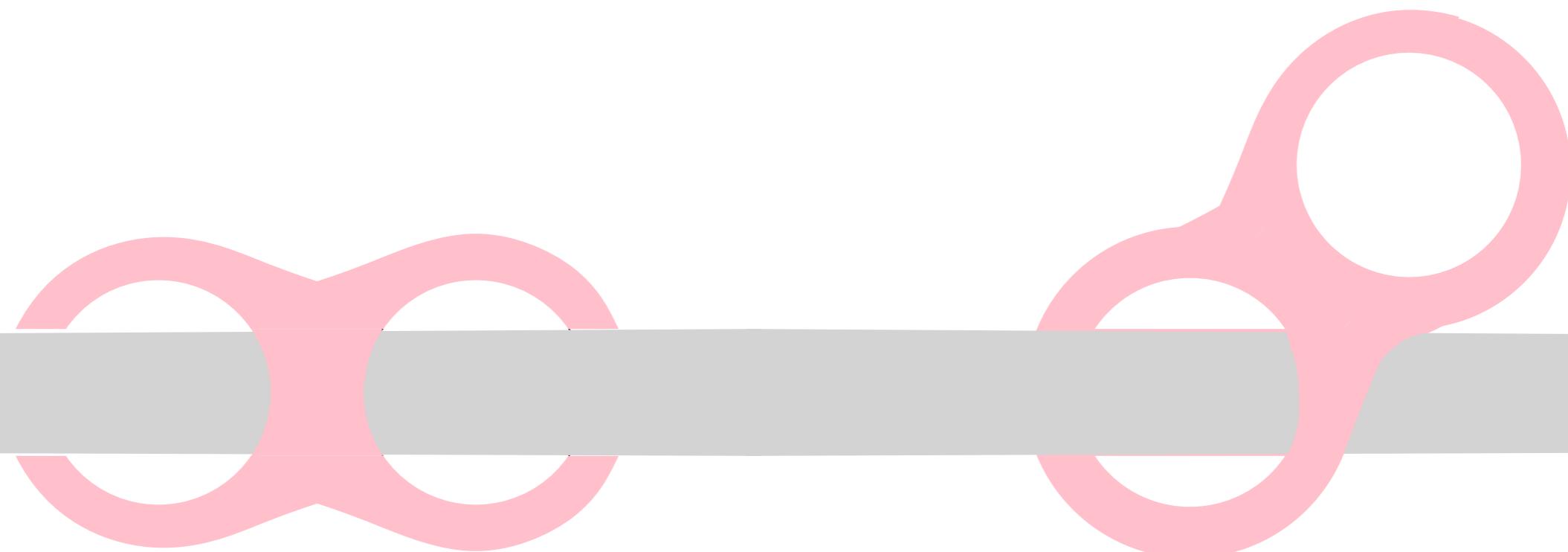


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Q: Can you find an isotopy between these guys?



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Pb 1: How to encode topological spaces for computational purposes?

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Pb 1: How to encode topological spaces for computational purposes?

Pb 2: Looking for homotopy equivalences/homeomorphisms/isotopies is extremely difficult. Are there mathematical quantities that are invariant to homotopy equivalences **and** easy to compute?

A topological space fit for computation

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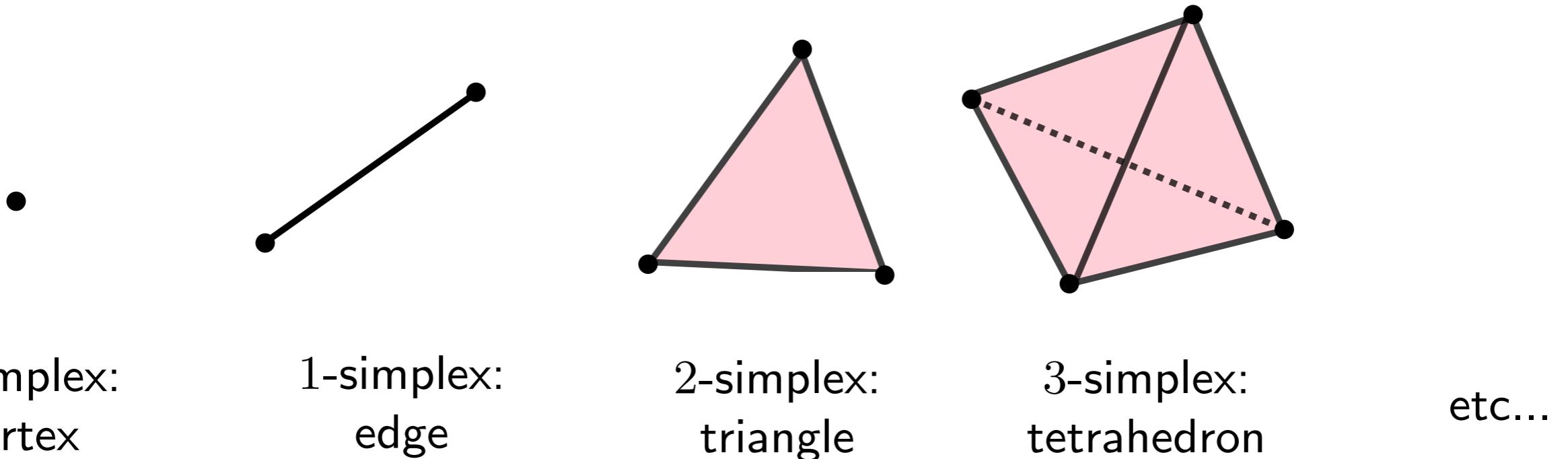
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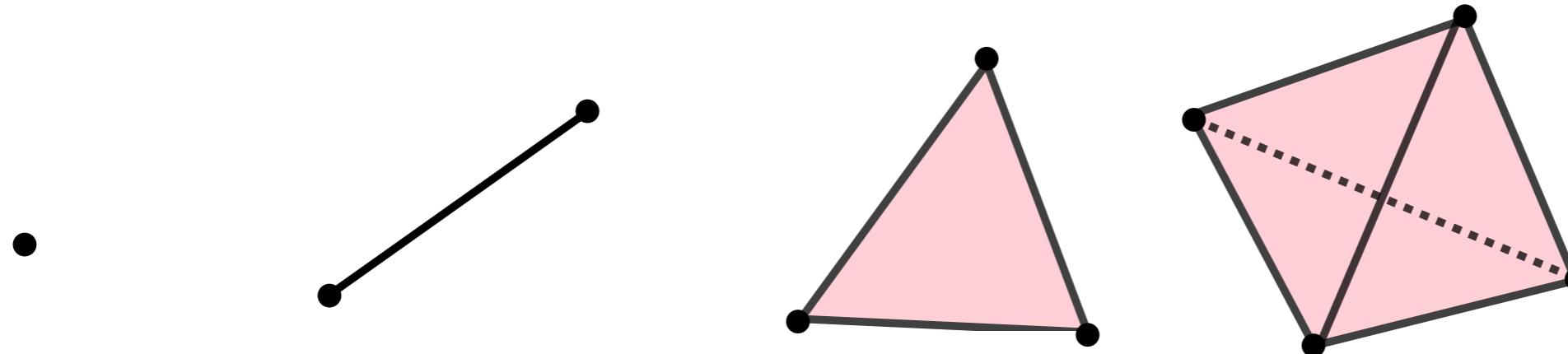
A: Using spaces made of small convex bricks, namely the *simplcial complexes* made of *simplices*.

Simplex and simplicial complex

Simplex and simplicial complex



Simplex and simplicial complex



0-simplex:
vertex

1-simplex:
edge

2-simplex:
triangle

3-simplex:
tetrahedron

etc...

Def: Given a set $P = \{p_0, \dots, p_k\} \subset \mathbb{R}^d$ of $k+1$ affinely independent points, the k -dimensional simplex σ (or k -simplex for short) spanned by P is the set of convex combinations

$$\sum_{i=0}^k \lambda_i p_i, \quad \text{with} \quad \sum_{i=0}^k \lambda_i = 1 \quad \text{and} \quad \lambda_i \geq 0.$$

The points p_0, \dots, p_k are called the **vertices** of σ .

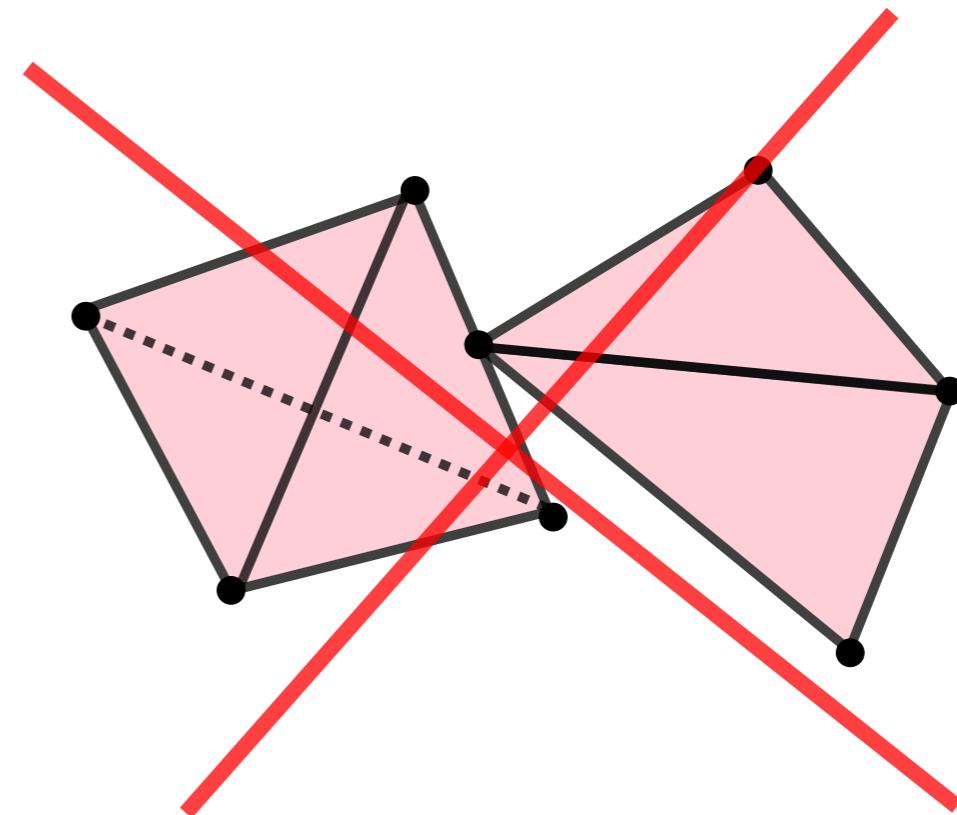
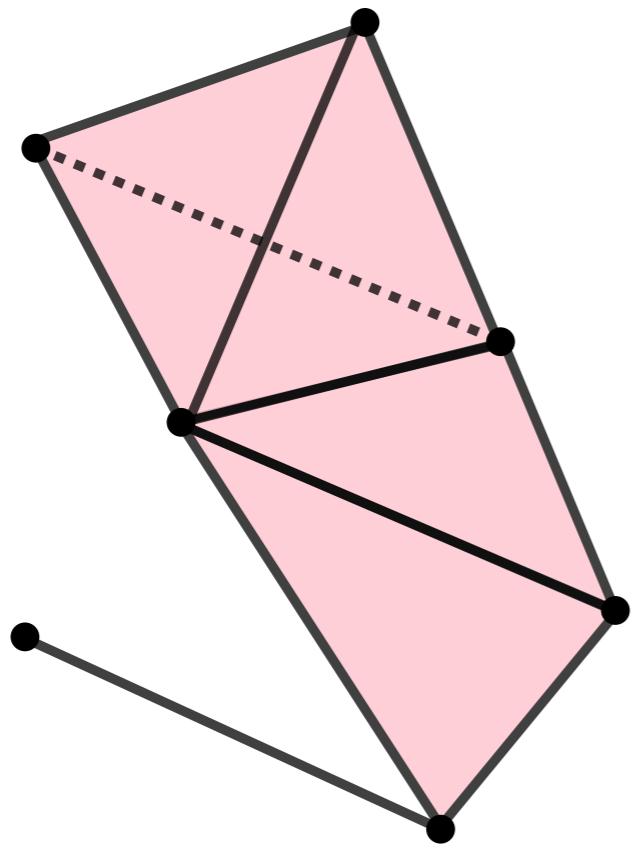
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Def: A **simplicial complex** K in \mathbb{R}^d is a collection of simplices s.t.:

- (i) any face of a simplex of K is a simplex of K ,
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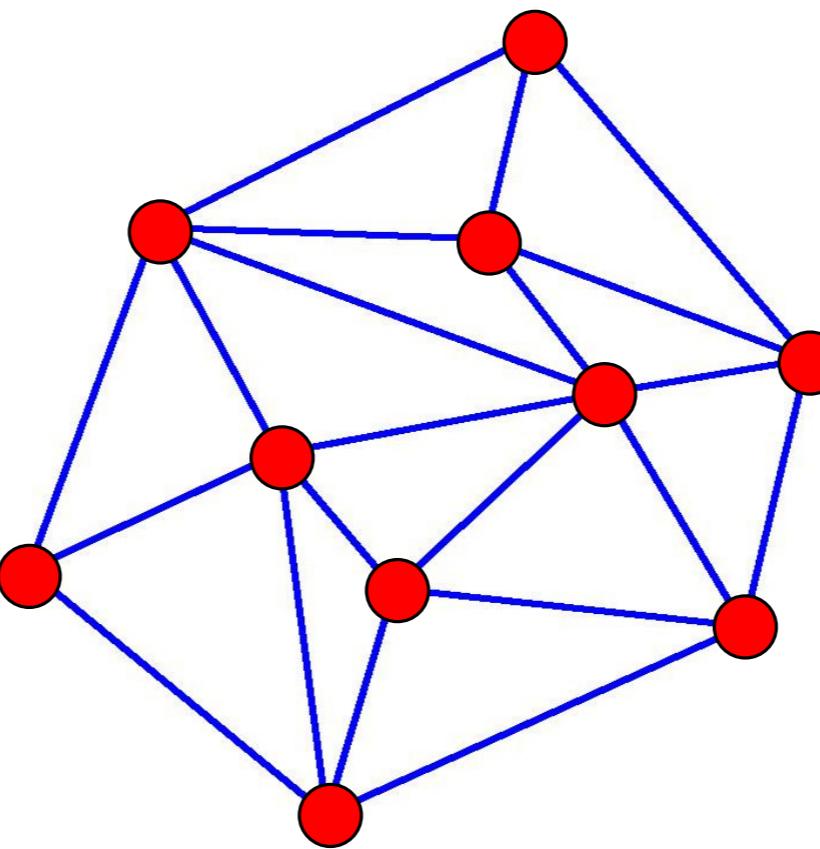
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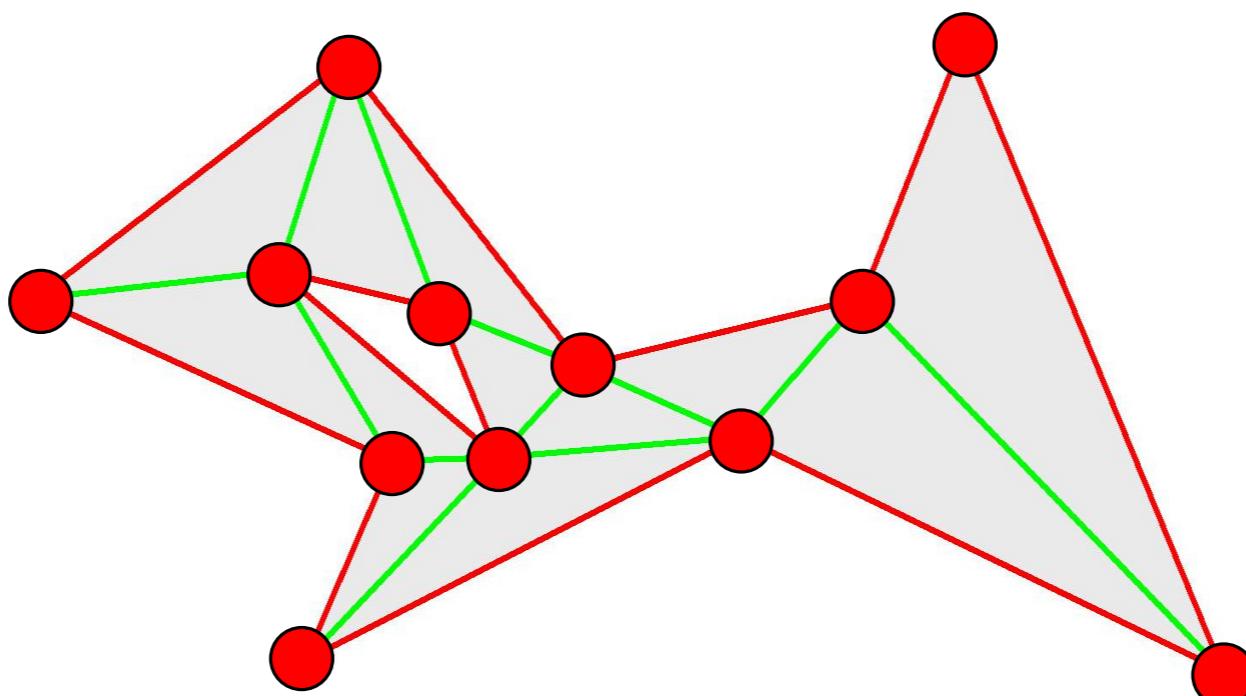


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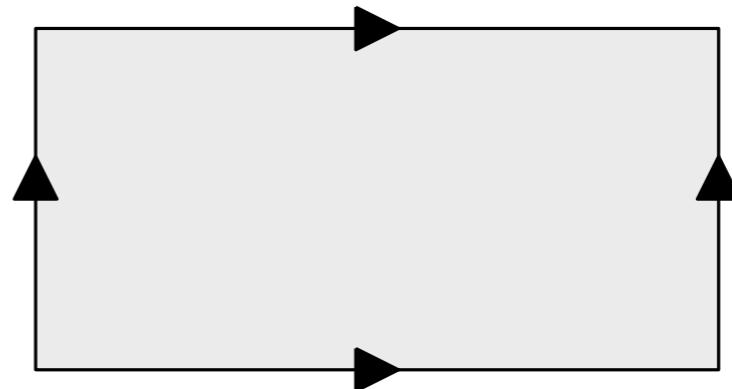
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Q: Triangulate

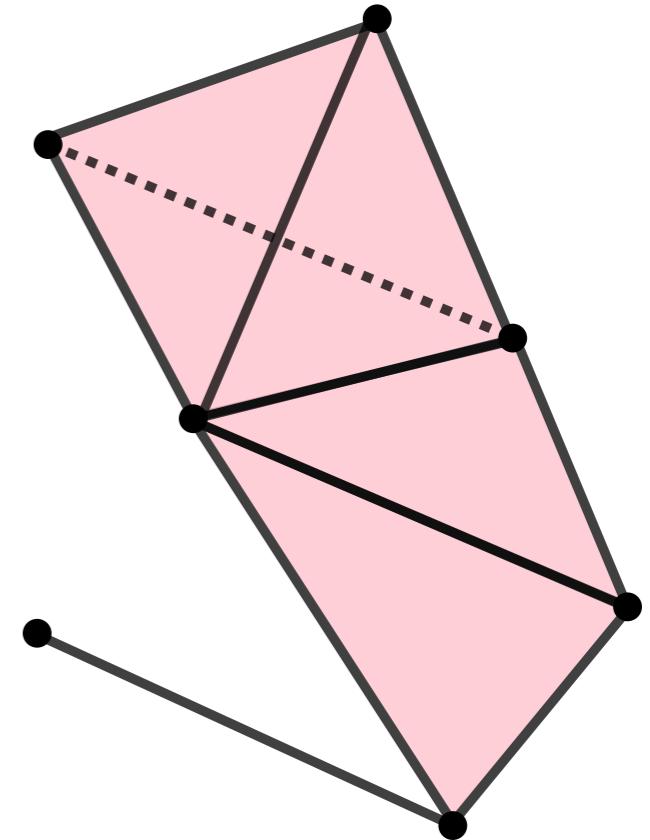


Abstract simplex and simplicial complex

Def: Let $P = \{p_1, \dots, p_n\}$ be a (finite) set. An **abstract simplicial complex** K with vertex set P is a set of subsets of P satisfying the two conditions:

- (i) the elements of P belong to K ,
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The elements of K are the **simplices**.

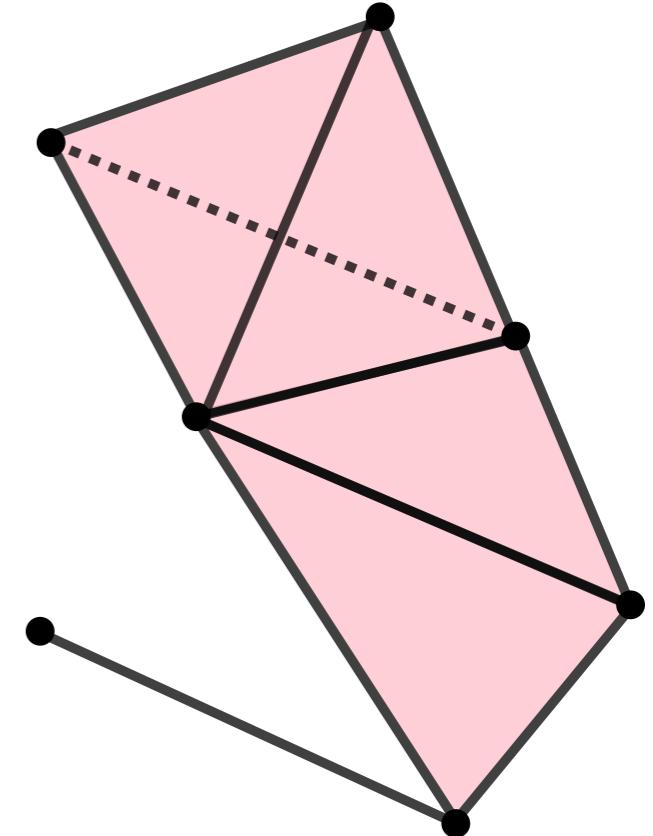


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IMPORTANT

Simplicial complexes can be seen at the same time as geometric/topological spaces (good for topological/geometrical inference) and as combinatorial objects (abstract simplicial complexes, good for computations).

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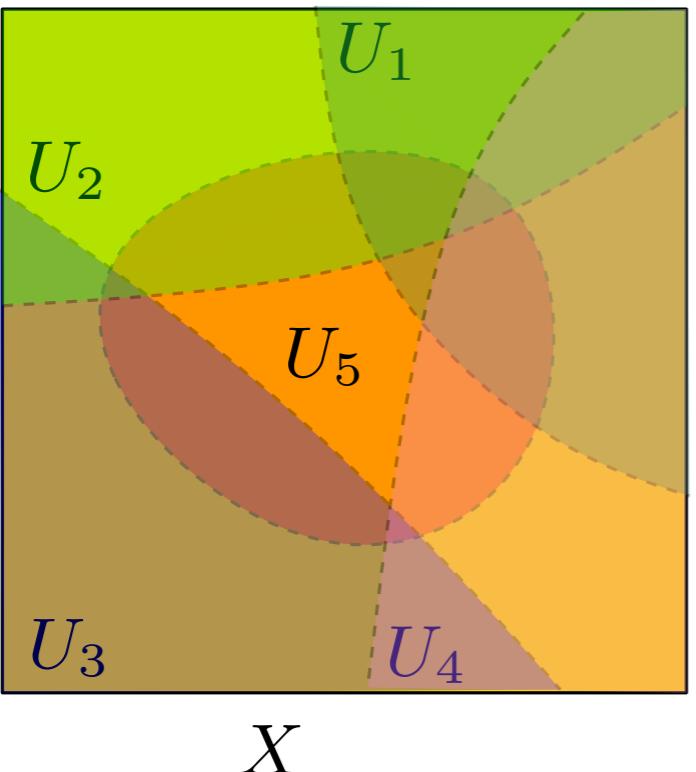
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Abstract simplicial complexes and their realizations are *homeomorphic*.

Nerve complex

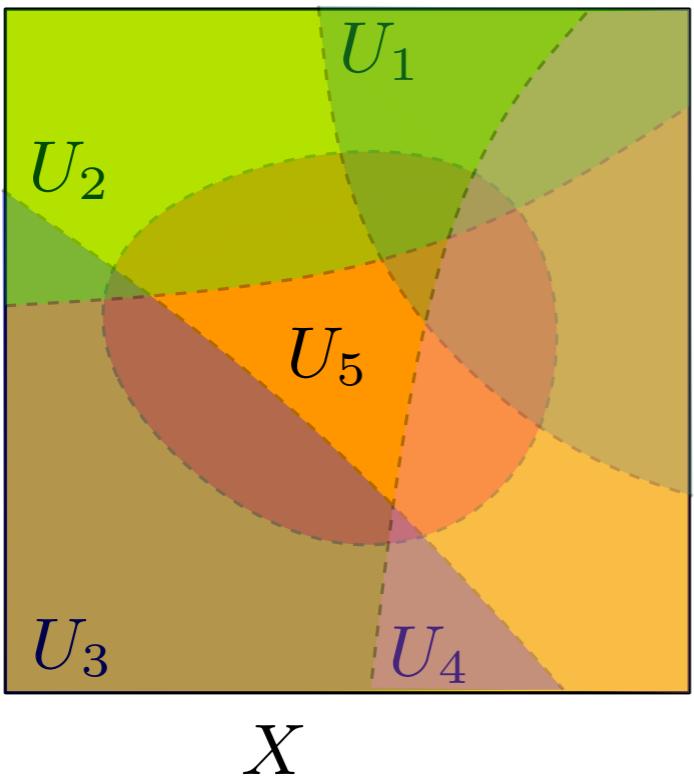
Def: An **open cover** of a topological space X is a collection $\mathcal{U} = (U_i)_{i \in I}$ of open subsets $U_i \subseteq X$, $i \in I$ where I is a set, such that $X \subseteq \bigcup_{i \in I} U_i$.

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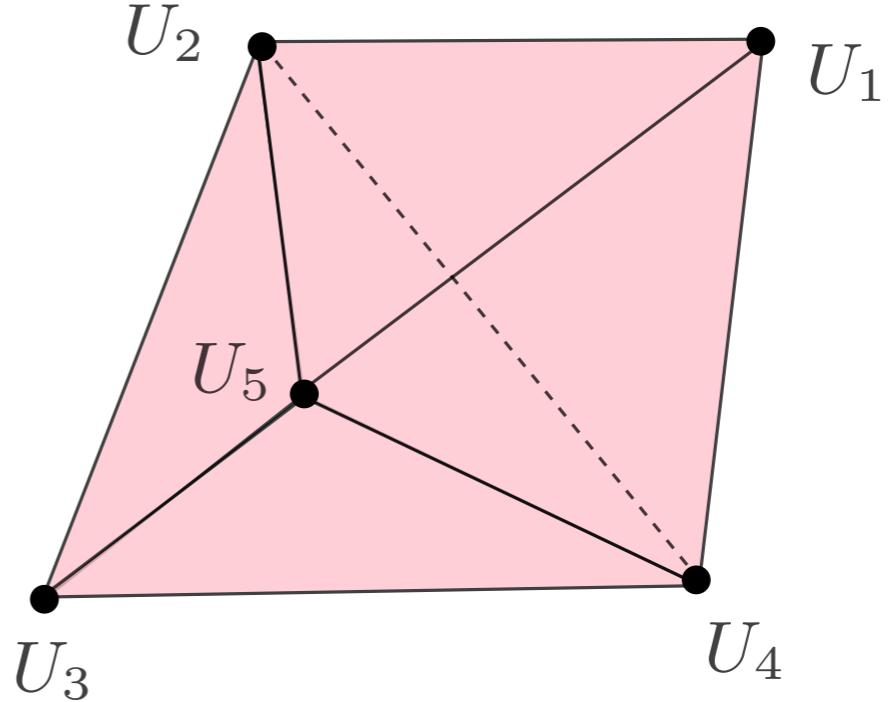
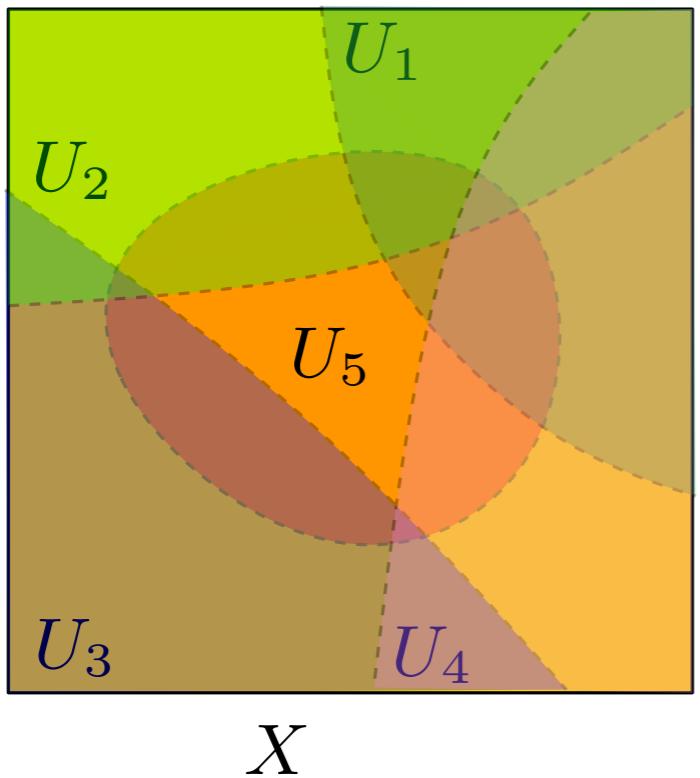


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Def: Given a cover of a topological space X , $\mathcal{U} = (U_i)_{i \in I}$, its **nerve** is the abstract simplicial complex $C(\mathcal{U})$ whose vertex set is \mathcal{U} and s.t.

$$\sigma = [U_{i_0}, U_{i_1}, \dots, U_{i_k}] \in C(\mathcal{U}) \text{ if and only if } \bigcap_{j=0}^k U_{i_j} \neq \emptyset.$$

Nerve complex



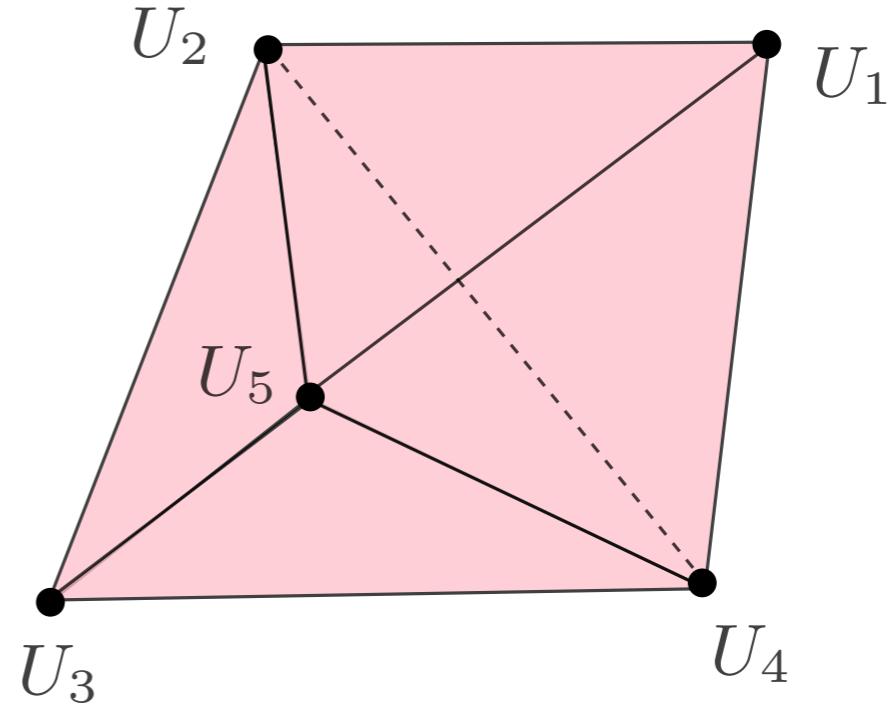
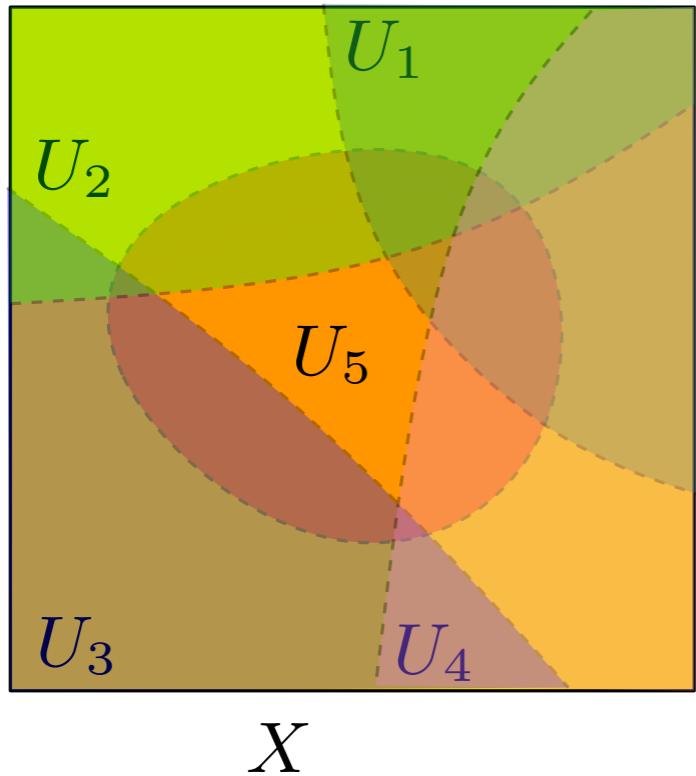
Def: An **open cover** of a topological space X is a collection $\mathcal{U} = (U_i)_{i \in I}$ of open subsets $U_i \subseteq X$, $i \in I$ where I is a set, such that $X \subseteq \bigcup_{i \in I} U_i$.

Def: Given a cover of a topological space X , $\mathcal{U} = (U_i)_{i \in I}$, its **nerve** is the abstract simplicial complex $C(\mathcal{U})$ whose vertex set is \mathcal{U} and s.t.

$$\sigma = [U_{i_0}, U_{i_1}, \dots, U_{i_k}] \in C(\mathcal{U}) \text{ if and only if } \bigcap_{j=0}^k U_{i_j} \neq \emptyset.$$

Nerve complex

[On the imbedding of systems of compacta in simplicial complexes, Borsuk, Fund. Math., 1948]



The Nerve Theorem: Let $\mathcal{U} = (U_i)_{i \in I}$ be a finite open cover of a subset X of \mathbb{R}^d such that any intersection of the U_i 's is either empty or contractible. Then X and $C(\mathcal{U})$ are homotopy equivalent.

In particular, every convex set is contractible.

$\check{\text{C}}$ ech and (Vietoris)-Rips complexes

Def: Given a point cloud $P = \{P_1, \dots, P_n\} \subset \mathbb{R}^d$, its $\check{\text{C}}$ ech complex of radius $r > 0$ is the abstract simplicial complex $C(P, r)$ s.t. $\text{vert}(C(P, r)) = P$ and

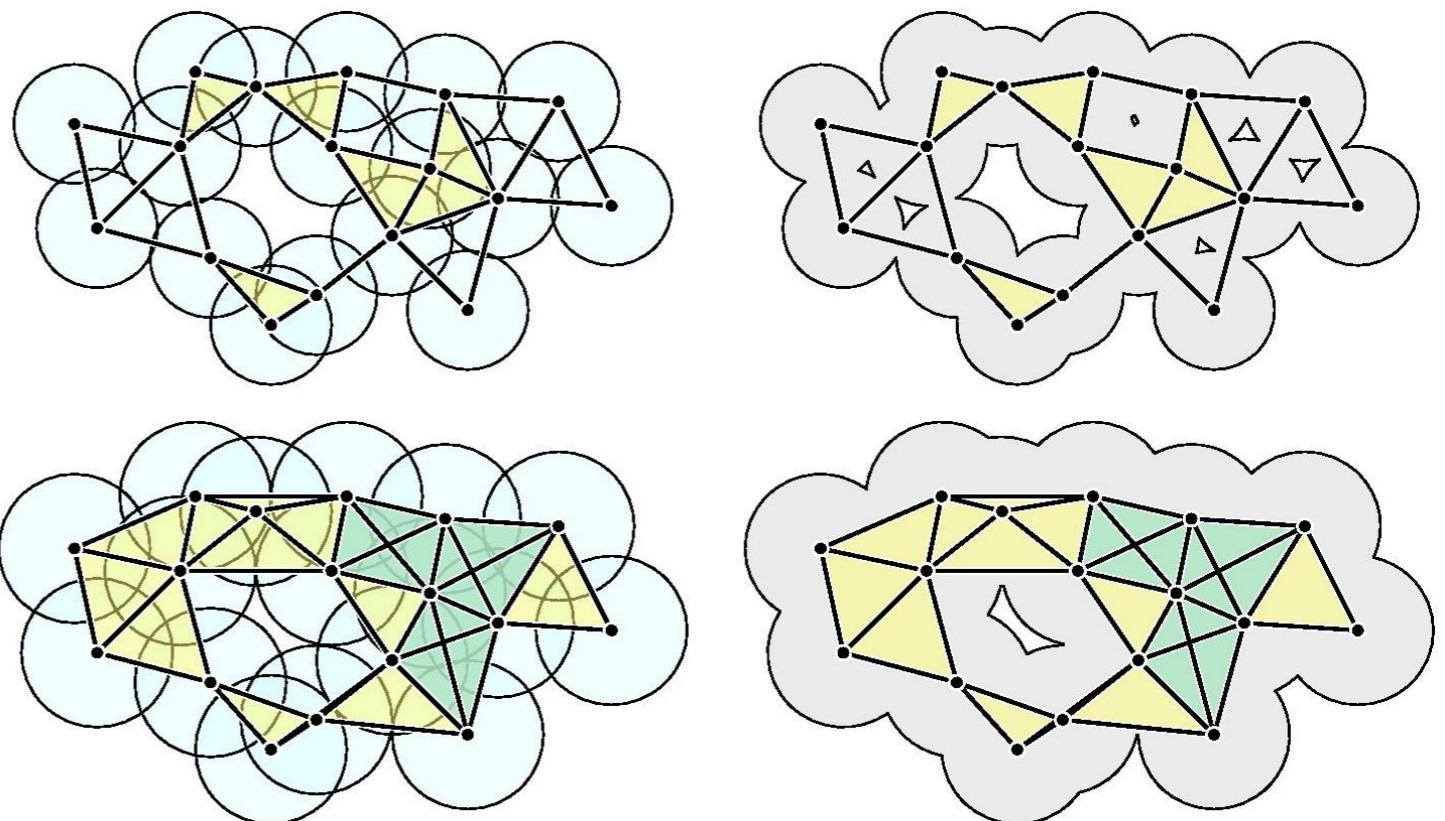
$$\sigma = [P_{i_0}, P_{i_1}, \dots, P_{i_k}] \in C(P, r) \quad \text{iif} \quad \cap_{j=0}^k B(P_{i_j}, r) \neq \emptyset.$$

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Q: Does the Nerve Theorem apply to $\check{\text{C}}$ ech complexes?



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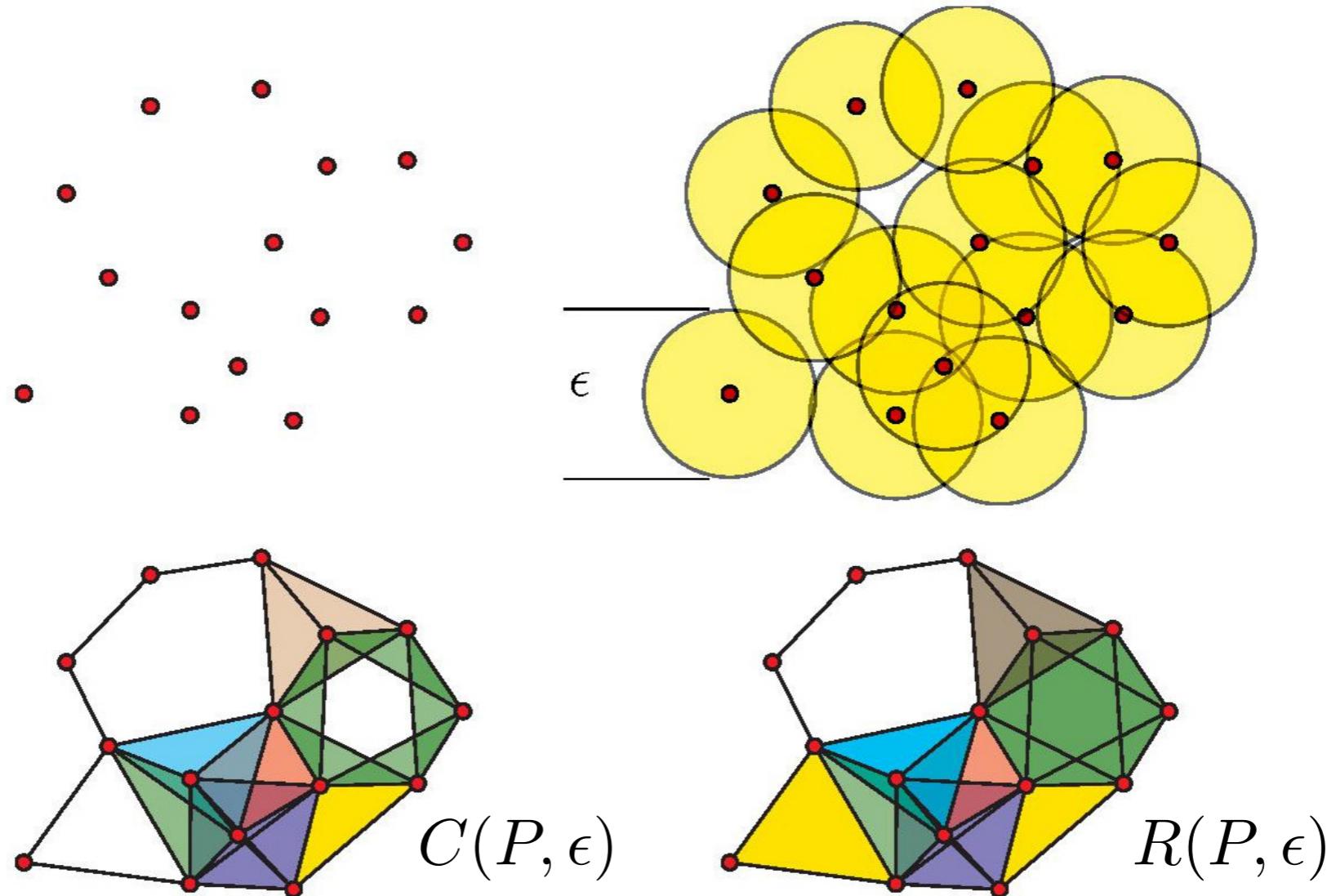
Def: Given a point cloud $P = \{P_1, \dots, P_n\} \subset \mathbb{R}^d$, its Rips complex of radius $r > 0$ is the abstract simplicial complex $R(P, r)$ s.t. $\text{vert}(R(P, r)) = P$ and

$$\sigma = [P_{i_0}, P_{i_1}, \dots, P_{i_k}] \in R(P, r) \quad \text{iif} \quad \|P_{i_j} - P_{i_{j'}}\| \leq 2r, \forall 1 \leq j, j' \leq k.$$

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Good news is that Rips and $\check{\text{C}}$ ech complexes are related:

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Prop: $R(P, r/2) \subseteq C(P, r) \subseteq R(P, r)$.

Q: Prove it.

Storing simplicial complexes

[*The Simplex Tree: An Efficient Data Structure for General Simplicial Complexes*, Boissonnat, Maria, Algorithmica, 2014]

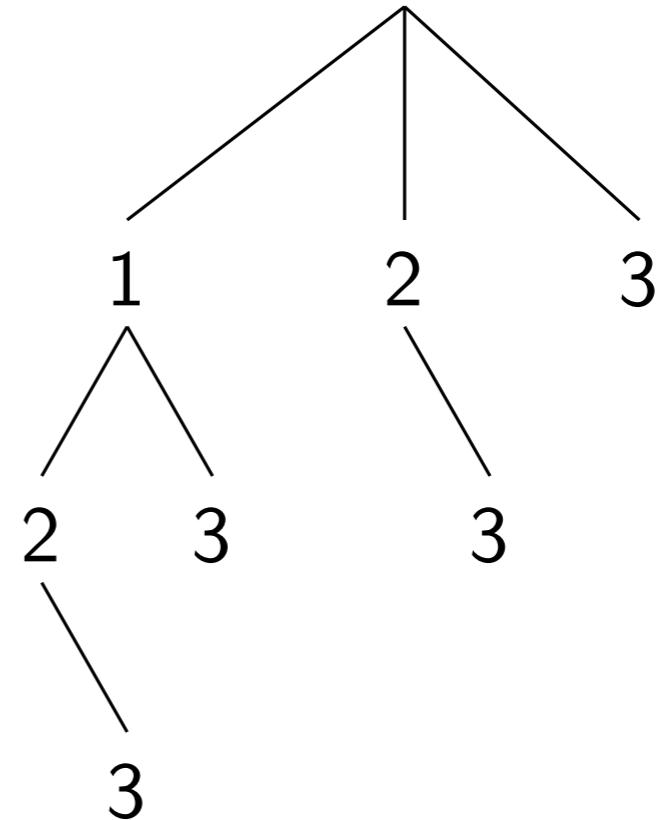
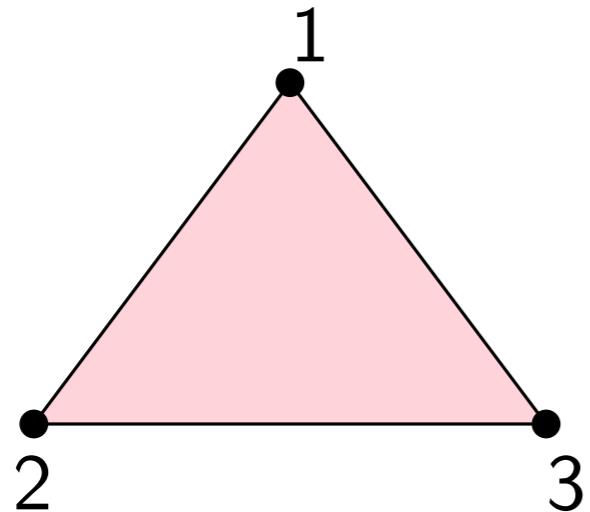
We want to store simplicial complexes with a data structure that allows to perform standard operations (insertion of a simplex, checking if a simplex is present, etc) in a fast and easy way.

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Idea: store sorted simplices in a prefix tree (also called *trie*).

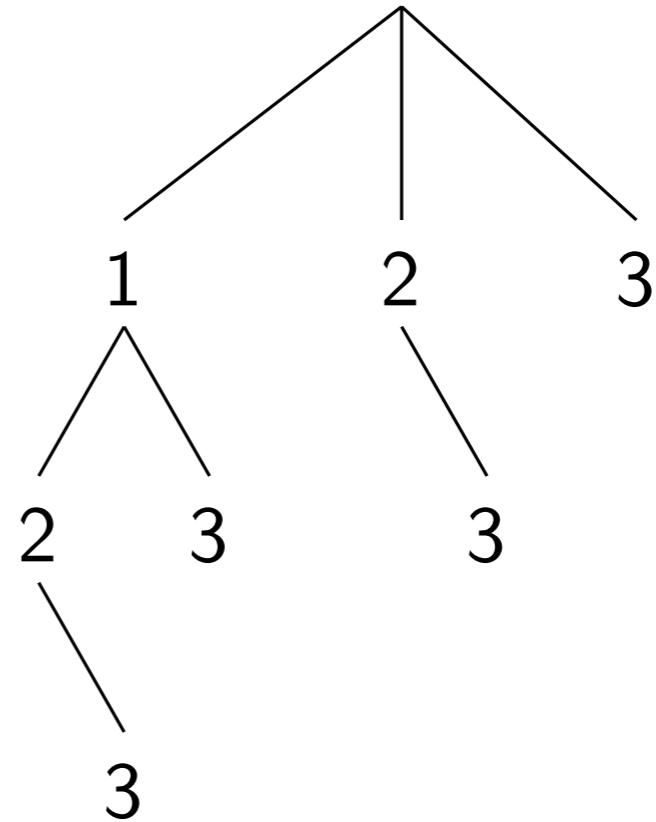
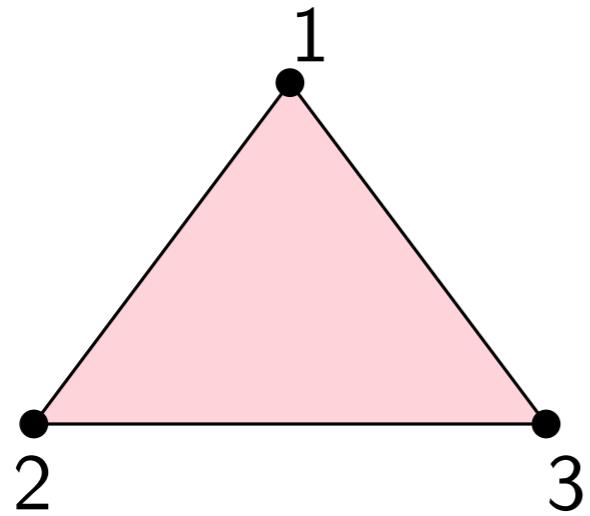


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This is called the *simplex tree*.

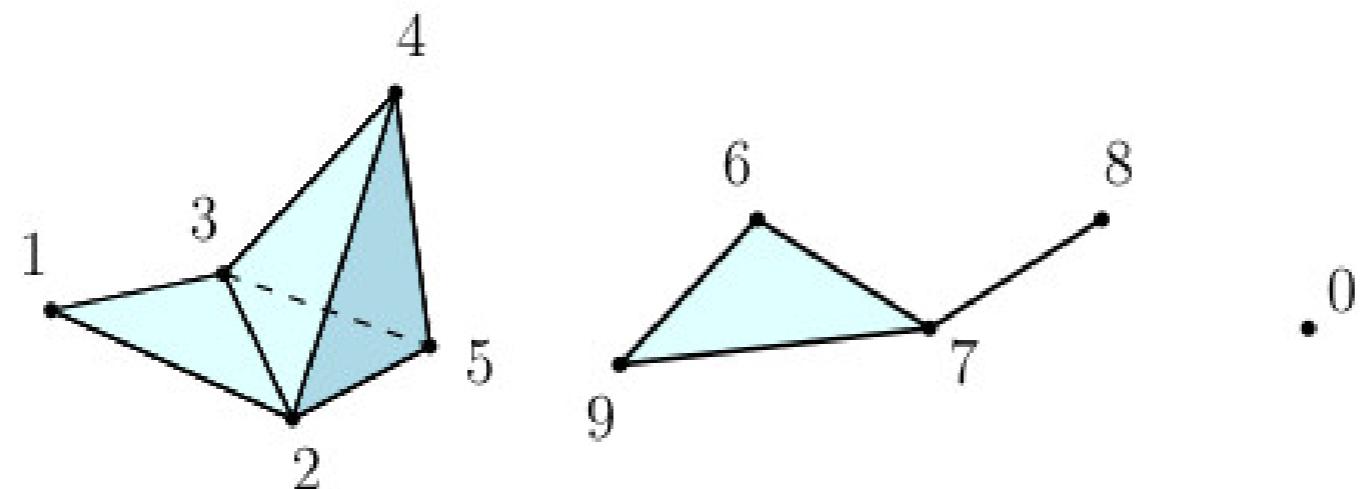
It allows to store all simplices explicitly without storing all adjacency relations, while maintaining low complexity for basic operations.

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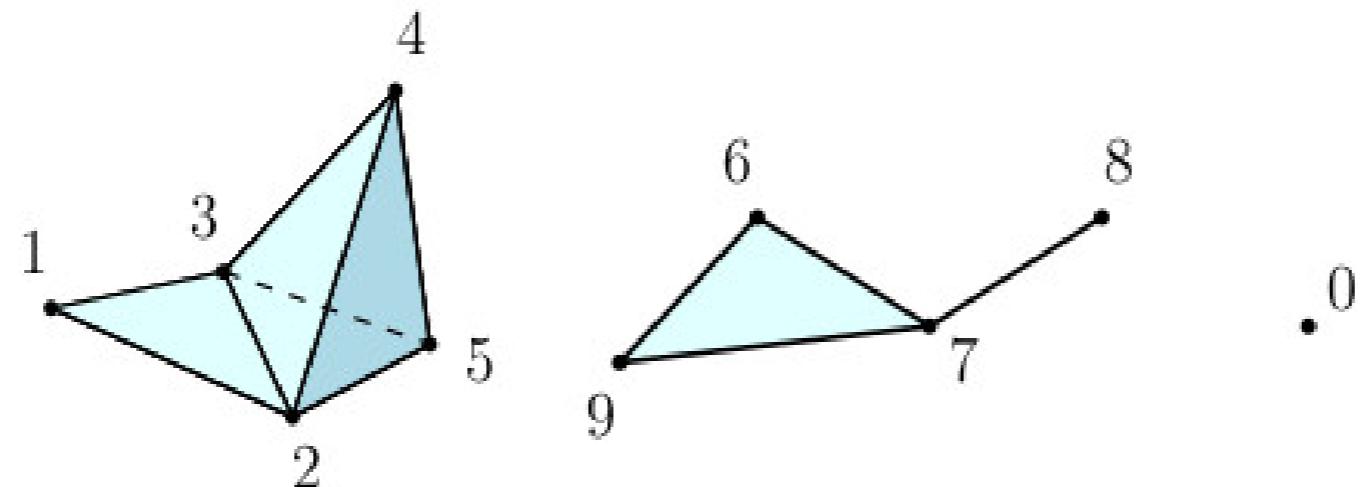


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Number of nodes in simplex tree = number of simplices

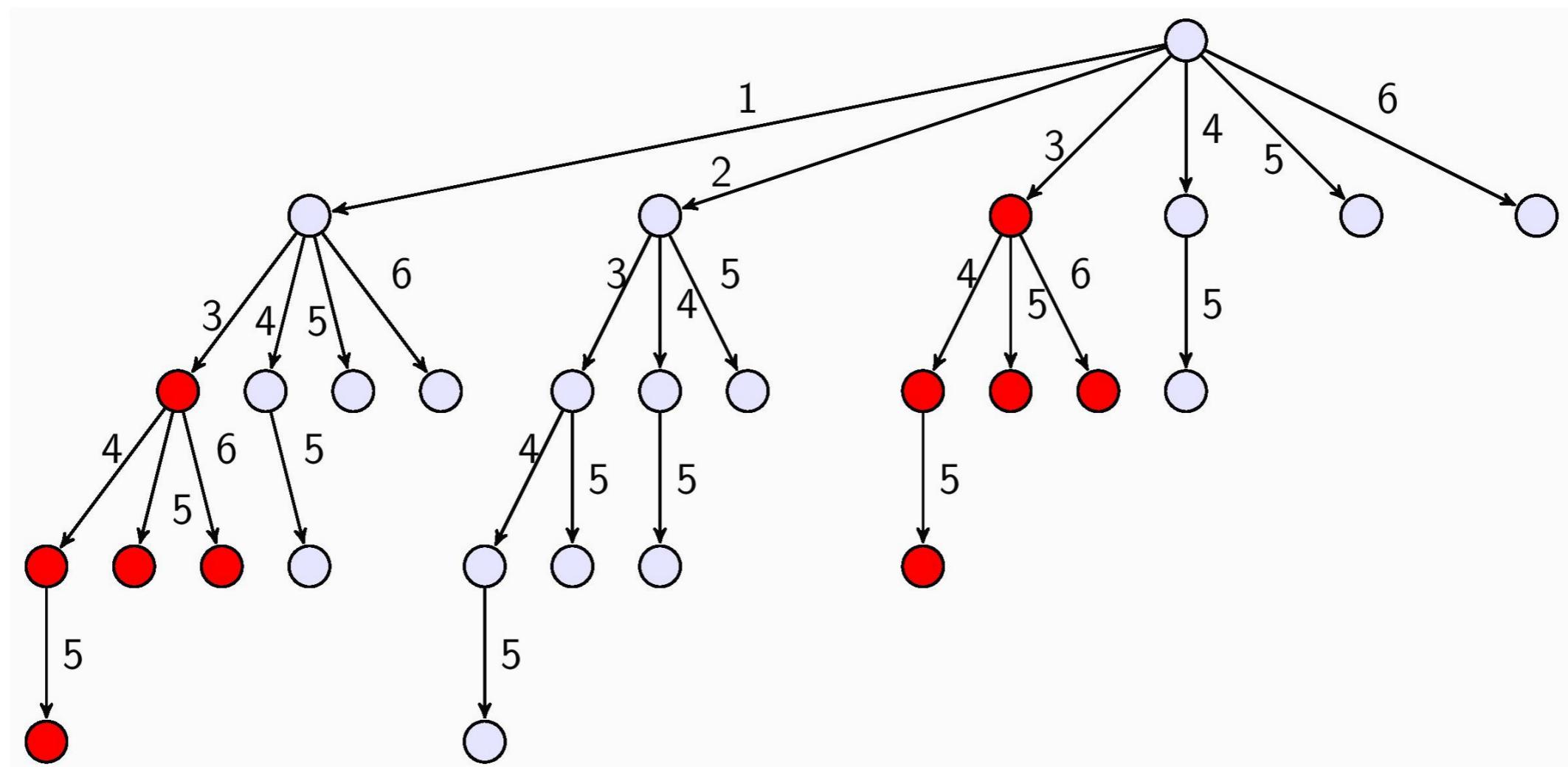
Depth of simplex tree = $1 + \text{dimension of complex}$

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