

# What is the shape of data?

A brief introduction to topological data analysis

Mauricio Montes

Auburn University

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- ▶ Data analysis is the process of studying and modeling data to extract useful information.
- ▶ Typically, we use statistics and machine learning to analyze data. We can improve these methods by using tools from topology.

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- ▶ We can think of manifolds as "things that locally look like"  $\mathbb{R}^n$ . Most objects in the real world are manifolds.
- ▶ If we take enough samples, we can approximate the shape of the manifold our data is living in. This allows us to answer topological questions about our data.

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- ▶ The challenge with the step above is **stability** under perturbations and noise.
- ▶ This topological data provides us with a new descriptor of the data. This can then be used in conjunction with other methods to improve our analysis.

# Topological Preliminaries

- ▶ A topological space is a pair  $(X, \tau)$  consisting of a set  $X$  with a collection of subsets  $\tau$ , called the topology on  $X$ , such that:
  1.  $\emptyset, X \in \tau$
  2. If  $U_i \in \tau$  for  $i \in I$ , then  $\bigcup_{i \in I} U_i \in \tau$
  3. If  $U_1, U_2 \in \tau$ , then  $U_1 \cap U_2 \in \tau$
- ▶ The Euclidean topology on  $\mathbb{R}^n$  is the collection of all open subsets of  $\mathbb{R}^n$ . We define an open set as a set  $B_r(x) = \{y \in \mathbb{R}^n : d(x, y) < r\}$ . Where  $d$  is the Euclidean metric.



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- ▶ We say that a topological space is Hausdorff if for any two points  $x, y \in X$  there exist disjoint open sets  $U, V \in \tau$  such that  $x \in U$  and  $y \in V$ .

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- ▶ A basis for a topological space  $(X, \tau)$  is a collection  $\mathcal{B}$  of open sets such that every open set in  $\tau$  can be written as a union of sets in  $\mathcal{B}$ .
- ▶ A topological space is second countable if it has a countable basis.

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- ▶ In the Euclidean topology, the topology is actually generated by the collection of open balls. These are a basis for the Euclidean topology.

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- ▶ A homeomorphism is a continuous map  $f : X \rightarrow Y$  such that there exists a continuous map  $g : Y \rightarrow X$  such that  $g \circ f = id_X$  and  $f \circ g = id_Y$ . (Bicontinuous, Bijective)

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- ▶ If there exists a homeomorphism between two topological spaces, we say that they are homeomorphic.

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- ▶ A topological manifold is a second countable Hausdorff space that is locally homeomorphic to  $\mathbb{R}^n$ .
- ▶ The last condition means that for every point  $x \in X$ , there exists an open set  $U \in \tau$  such that  $x \in U$  and  $U$  is homeomorphic to an open set  $\hat{U} \subseteq \mathbb{R}^n$ .
- ▶ To make this a smooth manifold, we require that the homeomorphism is actually a diffeomorphism. (Smooth map with smooth inverse)

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- ▶ A  $k$ -simplex is the convex hull of  $k + 1$  linearly independent points in  $\mathbb{R}^n$ .
- ▶ A simplicial complex  $K$  is a collection of simplices that satisfies two conditions:
  1. If  $\sigma \in K$ , then every face of  $\sigma$  is also in  $K$ .
  2. If  $\sigma_1, \sigma_2 \in K$ , then  $\sigma_1 \cap \sigma_2$  is either empty or a face of both  $\sigma_1$  and  $\sigma_2$ .