# Topology and Data: A Tour

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- Also, the *nature* of the data we are obtaining is significantly different.
- In-class task: try to Chatgpt the following:
  - "What are the different types of data that could be produced in morden sience, engineering and everyday life?"

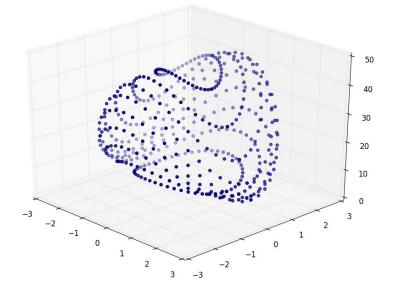
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  - Data is often very high-dimensional, restricting our ability to understand it (e.g., visualize) and process it
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  - Data is often very high-dimensional, restricting our ability to understand it (e.g., visualize) and process it
  - Data obtained is also very noisy and has more missing information
- Our ability to analyze this data, both in terms of quantity and the nature of the data, is clearly not keeping pace

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Based on: Gunnar Carlsson, Topology and Data

Img source: https://stackoverflow.com/questions/31294355/create-surface-grid-from-point-cloud-data-in-python

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- Geometry and topology are very natural tools for this:
  - E.g., geometry can be regarded as the study of distance between points
  - We typically work with large finite sets of data with distance defined on the objects (e.g., point cloud)
  - Tools from the various branches of geometry can be adapted to the study of point clouds

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- One example of data analytical methods based on geometry (and statistics) is the famous principal component analysis (PCA)
- It's a dimension-reduction technique
  - projecting high-dimensional data into lower-dimensional space
  - while keeping the spread of the data in the most significant directions

#### Principal Component Analysis (PCA) Transformation

Original data (high-dimensions)

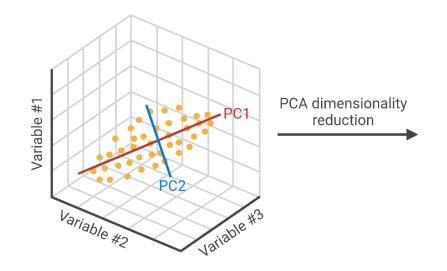


Image source: https://www.biorender.com/template/principal-component-analysis-pca-transformation

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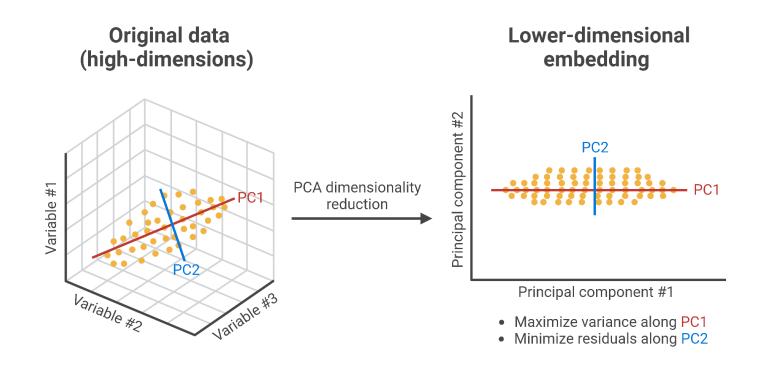


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Image source: https://www.sugarfit.com/blog/difference-between-type-1-and-type-2-diabetes

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- **Qualitative** information is important:
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  - E.g., when analyzing a data set for diabetes patients, it's important to understand that there are two types of the disease first, namely the juvenile and adult onset forms
  - We could also further develop quantitative methods for distinguishing them, but the first insight about the distinct forms of the disease is key

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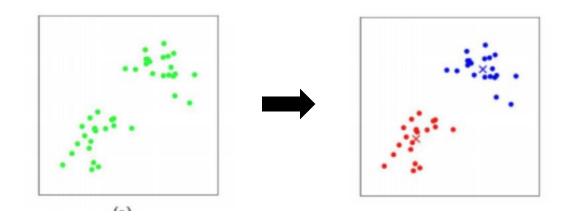
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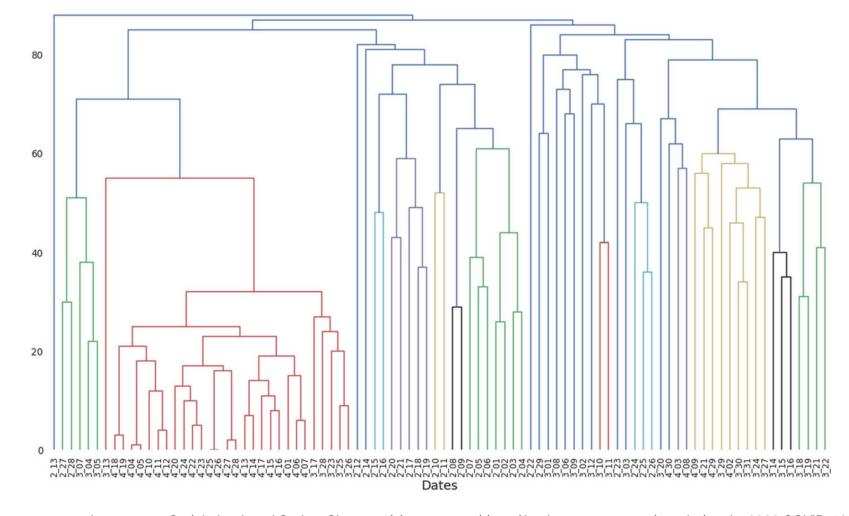
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  - But it is now well understood that maintaining the summary of the entire behavior of clustering under all possible parameter ε at once (called dendrogram) is more helpful

• Example of dendrogram (also called **Hierarchical Clustering**) hashtag usage on twitter during COVID-19:

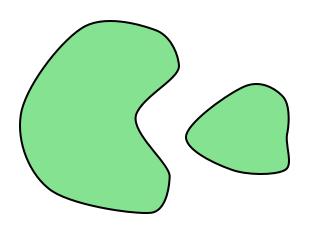
Hierarchical Clustering Dendrogram of Daily Clusterings

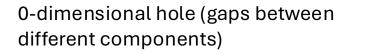


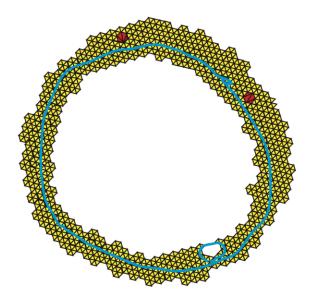
Img source: Cruickshank and Carley. Characterizing communities of hashtag usage on twitter during the 2020 COVID-19 pandemic by multi-view clustering

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  - We will learn topological methods that helps summarize **invariants** of data under a change of parameters

- Topology is exactly the branch of mathematics dealing with qualitative geometric information:
  - what the connected components of a space are
  - and more generally the connectivity information: the classification of loops and higher dimensional holes within the space



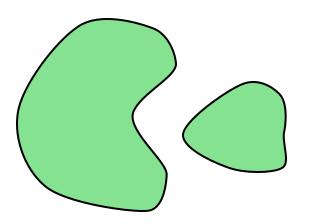




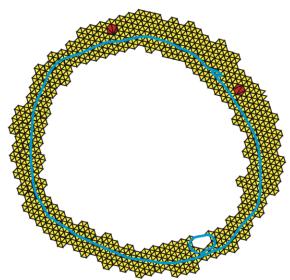
1-dimensional hole

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- Topology is exactly the branch of mathematics dealing with qualitative geometric information:
  - what the connected components of a space are
  - and more generally the connectivity information: the classification of loops and higher dimensional holes within the space
- This suggests that topological methodologies for data should be helpful in studying data qualitatively



0-dimensional hole (gaps between different components)



1-dimensional hole

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- We will look at more concrete examples of what topology can do later on

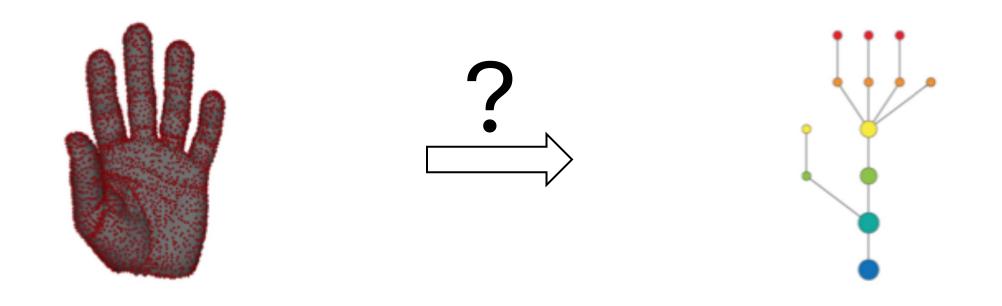
• In summary

"Data has Shape, Shape has Meaning"

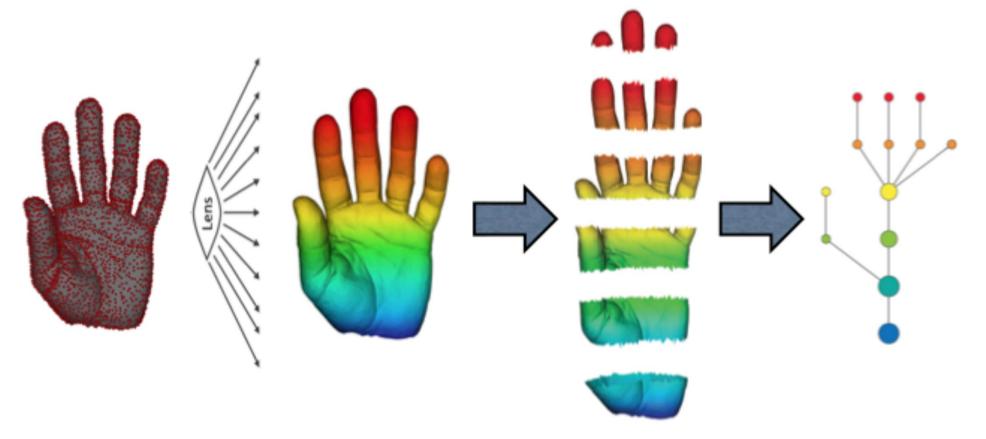
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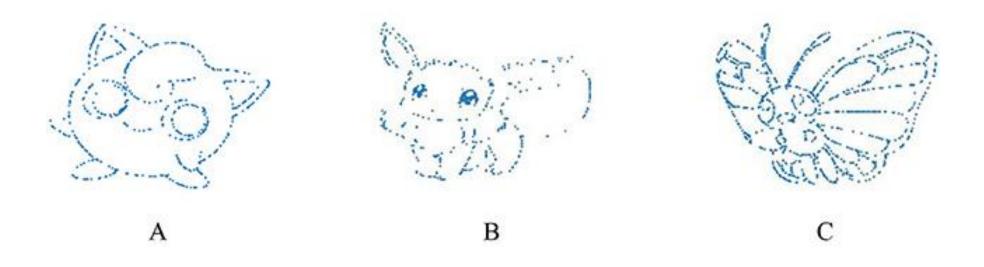
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- "Data has Shape, Shape has Meaning"
- Ex1: Using Mapper graph

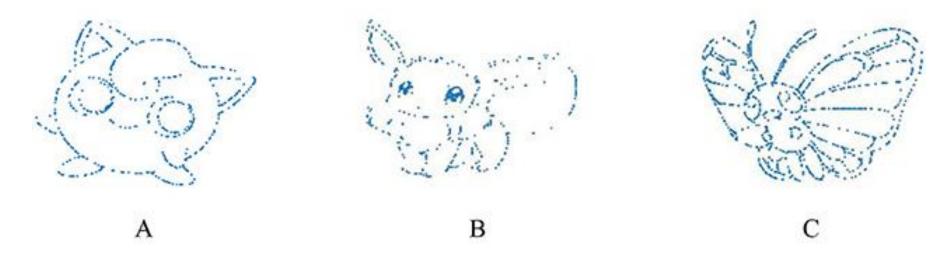


- "Data has Shape, Shape has Meaning"
- Ex2: How do we describe the shapes of the following three sets of points (for pokemons) and characterize their differences?



### Shape of Data

- "Data has Shape, Shape has Meaning"
- Ex2: How do we describe the shapes of the following three sets of points (for pokemons) and characterize their differences?
  - It's easy for human beings to 'delineate' the 'shapes' of these points form and to understand their difference (we naturally have intuitions about shapes)
  - But how to make computer understand shape?



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- In my own words: Topology studies *how points in a space connect to each other* within the space

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- Topology describes the properties of it that stay the same if we *stretch* it or *shrink* it or *bend* it, but without *gluing* things together or *breaking* it.

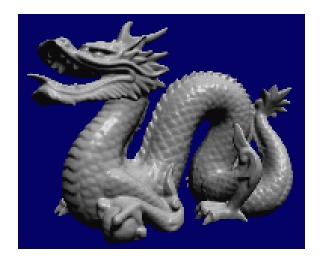


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- From a topological viewpoint, the following three are equivalent

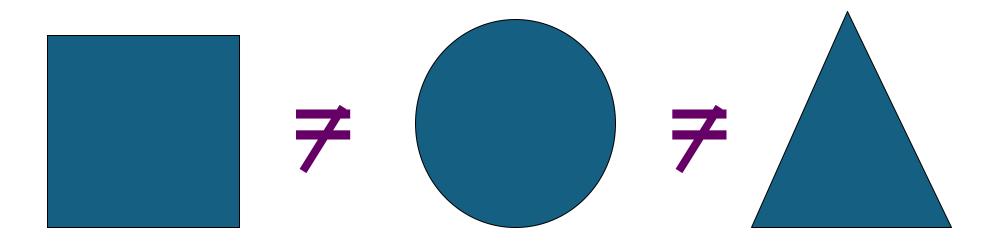


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- While the following are **not** equivalent



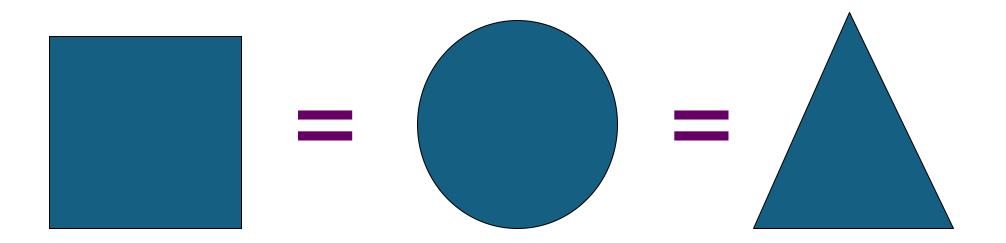






Courtesy of Isabel K. Darcy: http://www.math.uiowa.edu/~idarcy/AppliedTopology.html





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- Trick question: is the rubber band equivalent to a mobius strip (formed by inversely glueing the two ends of a paper tape; see: <u>https://www.youtube.com/watch?v=WJDalOreW88</u>)





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- Answer: no. Several reasons:
  - 1. To form a mobius band from the rubber band, you have to *break* the rubber band and *re-glue* the two ends (inversely), and these operations are not allowed (they are *not continuous*)

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  - 2. The rubber band is *orientable* (has two sides) while the mobius band is not (cannot differentiate the two sides): Orientability is an *invariant* that should be preserved if two spaces are equivalent. (TDA heavily draw upon other invariants such as *homology*, which we will look at later)

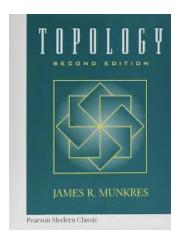
# Topology (more formally)

• From the book *Topology* by James Munkres

**Definition.** A *topology* on a set X is a collection  $\mathcal{T}$  of subsets of X having the following properties:

- (1)  $\emptyset$  and X are in  $\mathcal{T}$ .
- (2) The union of the elements of any subcollection of  $\mathcal{T}$  is in  $\mathcal{T}$ .
- (3) The intersection of the elements of any finite subcollection of  $\mathcal{T}$  is in  $\mathcal{T}$ .

A set X for which a topology  $\mathcal{T}$  has been specified is called a *topological space*.



# Topology (more formally)

- In the previous definition, each set  $S \in \mathcal{T}$  (notice that  $S \subseteq X$ ) is called an *open set*.
- The open sets are usually chosen to provide a notion of *"nearness"* without having a notion of distance defined.
- A topology allows defining properties such as
  - Continuity
  - Connectedness
  - Compactness

without defining a distance.

## Topology (more formally)

• An example of continuity:

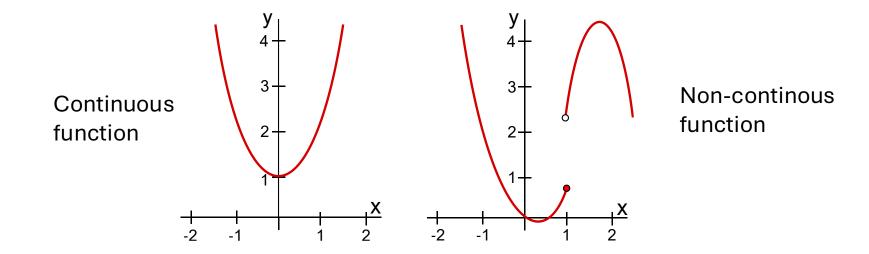


image src: https://www.sfu.ca/mathcoursenotes/Math%20157%20Course%20Notes/sec\_Continuit yIVT.html

# Continuity and topological equivalence (more formally)

Let X and Y be topological spaces. A function  $f : X \to Y$  is said to be *continuous* if for each open subset V of Y, the set  $f^{-1}(V)$  is an open subset of X.

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Let X and Y be topological spaces; let  $f : X \to Y$  be a bijection. If both the function f and the inverse function

 $f^{-1}:Y\to X$ 

are continuous, then f is called a *homeomorphism*.

"Homeomorphism" is the formal terminology for "topological equivalence" that we have been talking about

#### Examples of homeomorphic spaces

• On Wikipedia: <u>https://en.wikipedia.org/wiki/Topology</u>

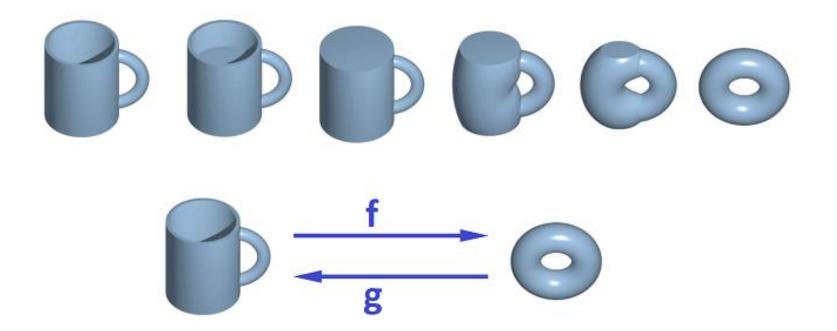


Image source: https://www.facebook.com/photo/?fbid=113751385892899&set=a.113751795892858

## Problem for our practical purpose

- Now we have (very roughly) defined a "topological structure" on our data, which can be used to describe the "shape" for the data
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- We need to "encode" the topology of your data into some format processible by the computer
- For this, we utilize some "numeric invariants" for the topological spaces
  - *Invariant*: something that does not change between spaces that are topologically equivalent (homeomorphic)

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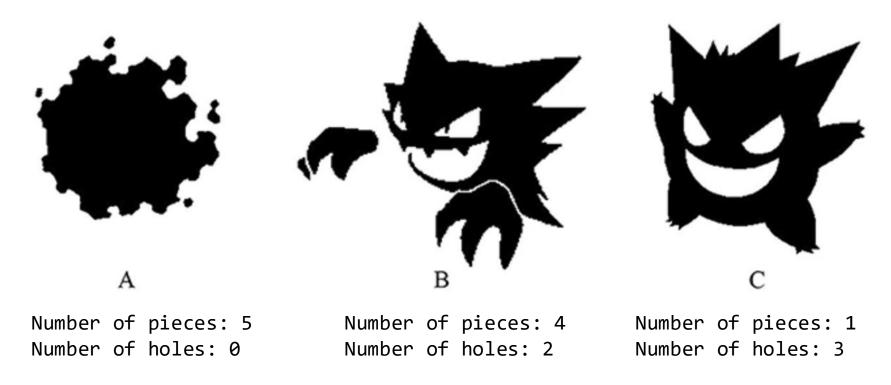
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- Remark: most formally, the numeric invariants are indeed called algebraic invariants

#### A toy version of algebraic invariant

• Counting the number of pieces and number of holes in an object

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• We will study this algebraic invariant more extensively later

# Discovering the shape of data by connecting the dots

• Let's try to discover the shape of the pokemon below



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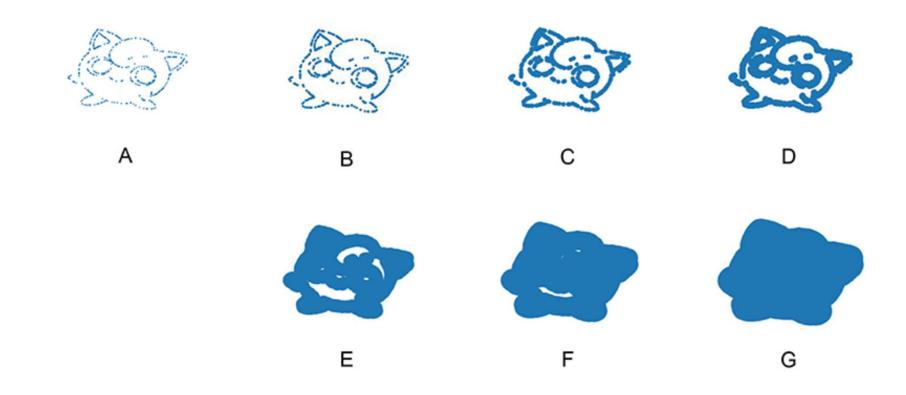
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- A problem with the data is that, it's just a discrete set of points — it doesn't form any "meaningful shapes" that we could count (and further make computer to process)
- In order to form some meaningful shape, we need to find a way to "connect the dots"

#### Connecting the dots

- We connect the dots by increasing their size.
  - As we make the dots larger, gaps between the dots become smaller, and eventually the dots overlap



#### Connecting the dots

• Remark: a more natural way of connecting the dots by drawing lines between the dots. We will do that more formally and extensively later

	Number of pieces	Number of holes
E.	224	0
Color B	101	0
Je of	17	2
	1	6
	1	6
	1	3
	1	0

Image source: https://kids.frontiersin.org/articles/10 .3389/frym.2021.551557

#### Connecting the dots

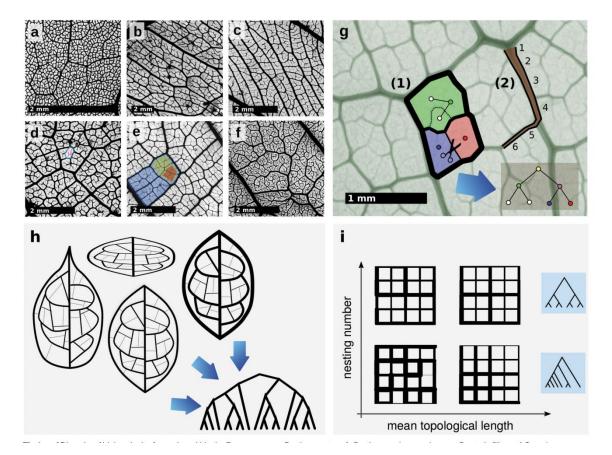
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- A person \*may\* be able to detect the "right size" for the previous example. But what about more involved shapes?
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#### Connecting the dots

- But another question arises: which size do we choose?
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- Furthermore, how do we let computer choose such a size?
- We will study a major tool in topological data analysis (TDA) called **Persistent Homology**, which provides a solution
- Hint: Persistent Homology does not try to find such a size, but rather it considers all the sizes and tracks the changes of the topological invariants, by tracking the how the pieces and holes persist, across all the sizes

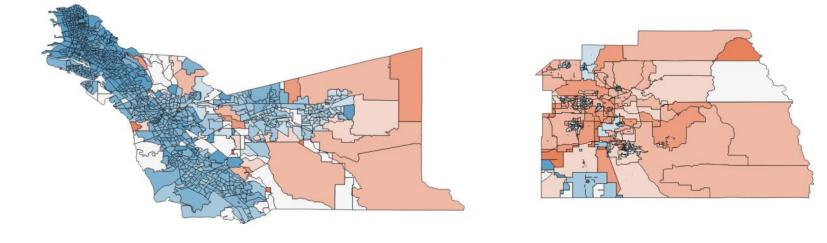
Examining *patterns of veins in leaves*: studied structure of >=100 leaves and found different patterns—like human fingerprints—in them

• These fingerprints help scientists identify leaves from small leaf fragments, and may also be helpful for improving our understanding of how leaves grow.



Ronellenfitsch, H., Lasser, J., Daly, D. C., and Katifori, E. 2015. Topological phenotypes constitute a new dimension in the phenotypic space of leaf venation networks.

Studying the voting patterns in California





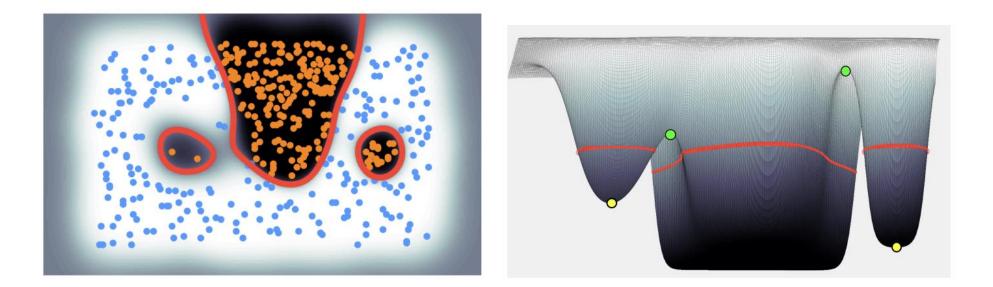
Feng, M., and Porter, M. A. 2021. Persistent homology of geospatial data: A case study with voting.

Utilizes *topological regularization* losses to rectify topological artifacts (broken legs, unrealistic thin structures, and small holes) in generating synthetic 3D models



Chen et al., Enhancing Implicit Shape Generators Using Topological Regularizations.

Utilizes *topological regularization* losses to reduce the topological complexity of the classification boundary of a binary classification task



Chen, Chao, Xiuyan Ni, Qinxun Bai, and Yusu Wang. A topological regularizer for classifiers via persistent homology.