

# Topological Data Analysis

Fall 2025

Syllabus  
Top-Review



# Course intro

- Syllabus & HW → main webpage
- HW submission → Canvas  
(+ gradebook)
- Prereqs: some linear algebra  
some programming background
- Slack?
- HW: Mix of pen/paper & Coding  
↳ flexible, so see me if you  
have any issues!

# Textbook(s)

Main reference:

free pdf

1:51PM Fri Aug 15

rcs.purdue.edu

CSE 40113: Algorithms, Spring 2025

Topological Data Analysis Book

**Book : Computational Topology for Data Analysis**  
(published by Cambridge University Press)

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[MAA review](#) and [zbMath review](#)

[Errata](#) detected in the print are corrected in electronic version marked with red text

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• Contents

Chapter 1: Basic Topology

- a. Topological spaces, metric space topology
- b. Maps: homeomorphisms, homotopy equivalence, isotopy
- c. Manifolds
- d. Functions on smooth manifolds
- e. Notes and Exercises

Chapter 2 (i) . Complexes

- a. Simplicial complexes
- b. Nerves, Čech and Vietoris-Rips complexes
- c. Sparse complexes (Delaunay, Alpha, Witness)
- d. Graph induced complexes

Others:

Listed on Syllabus

Most are available in the library, or  
Come visit my office!

## Project:

Ideally, this will connect to your research.

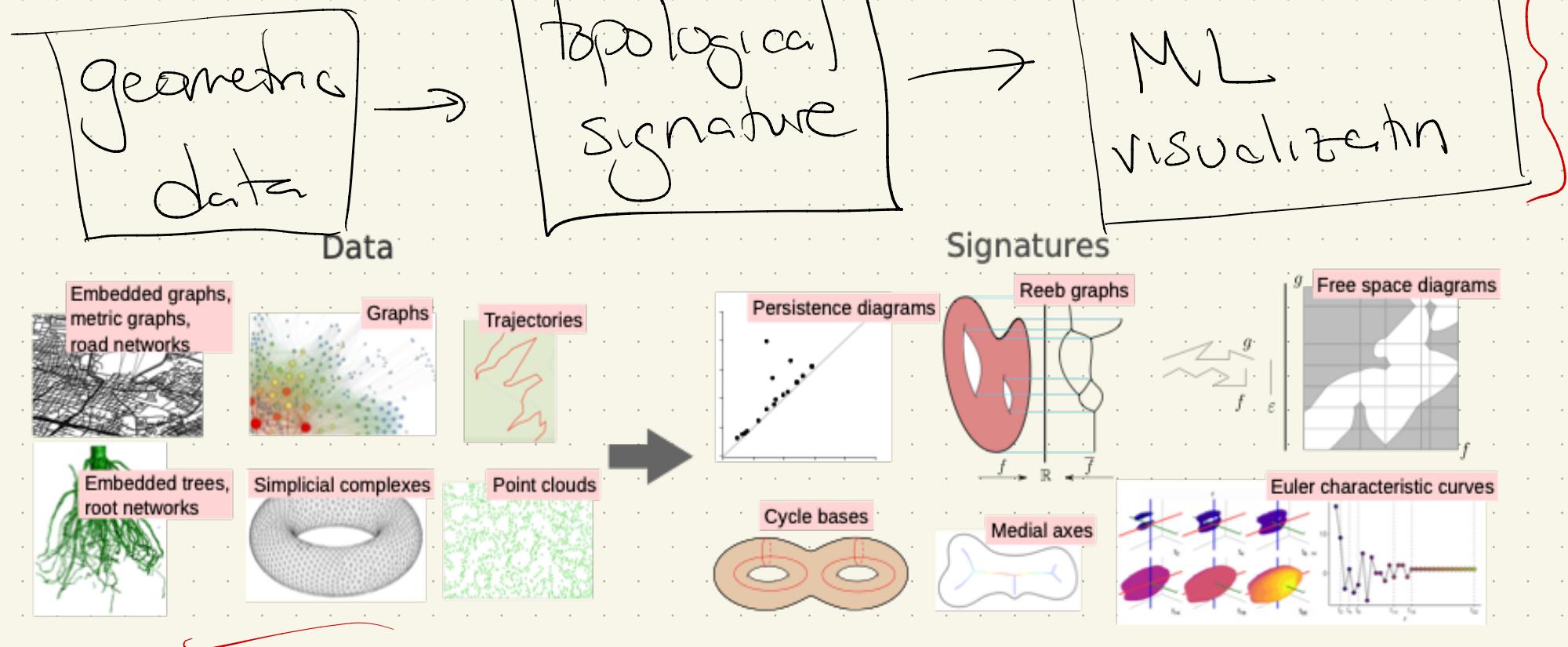
But - we'll have several assignments to help explore topics, if you're unsure of what direction to take.

### Outline:

- talk summary → in Sept
- paper "chase" { in October
- project proposal
- final presentation + submission

# What is topological data analysis?

Traditional pipeline:



Goals:

Concise  
stable & robust } computable!  
Metrics

## Some history:

Recognizing exact topology can be hard.  
How so?

- Deciding if 2 4-manifolds are homeomorphic is undecidable  
[Markov 1960, van Meter 2005]
- Deciding if you can "unknot" a curve using a fixed number of moves is NP-Hard  
[de Mesmay, Sedgwick, & Tancer 2021]



## An approach

Since we can't solve the problem exactly, focus on invariants and ways to simplify the data.

This is not new:

Examples:

- Knot invariants
- Curve skeletons
- Manifold approximations
  - ↳ ie meshes

A first example: Euler characteristic

Introduced first by Maurolico in 1537.

(This is known as

Then published by Euler in 1758:

Name	Image	Vertices	Edges	Faces	Euler characteristic:
		$V$	$E$	$F$	$\chi = V - E + F$
Tetrahedron		4	6	4	2
Hexahedron or cube		8	12	6	2
Octahedron		6	12	8	2
Dodecahedron		20	30	12	2
Icosahedron		12	30	20	2



For any embedded planar graph

$G = (V, E)$  with  $F$  faces

$$V - E + F = 2$$

Note: planar  $\Rightarrow$  embedded on sphere

More generally:

Euler characteristic:  $V - E + F = \chi$

Name	Image	$\chi$
Interval		1
Circle		0
Disk		1
Sphere		2
Torus (Product of two circles)		0
Double torus		-2
Triple torus		-4
Real projective plane		1
Möbius strip		0

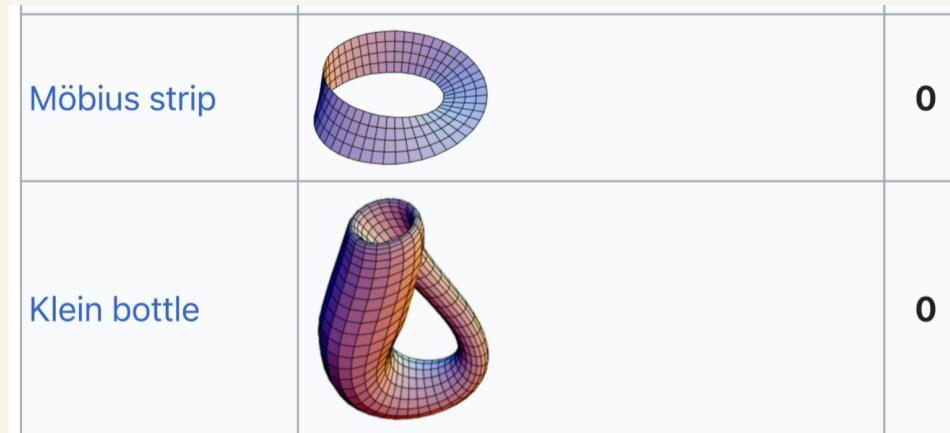
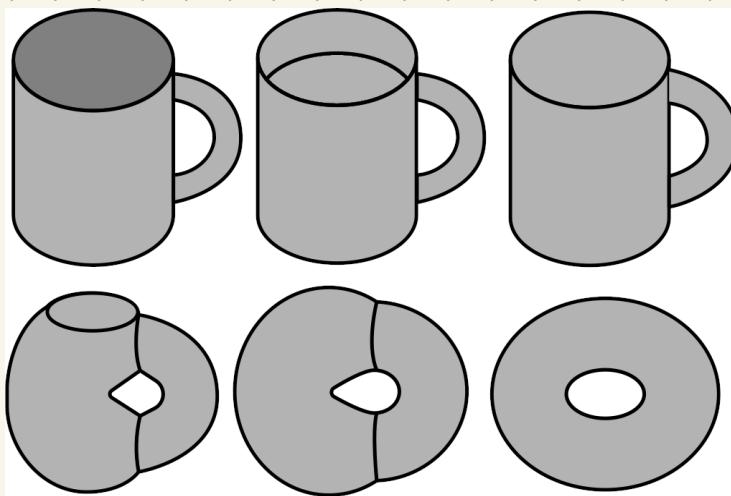
Ideal for computers:

- requires a discrete representation
- Given a data structure encoding  $V, E, F$  easy to calculate

$$V - E + F$$

Topological signatures: invariants

Different Euler characteristic  
⇒ different space



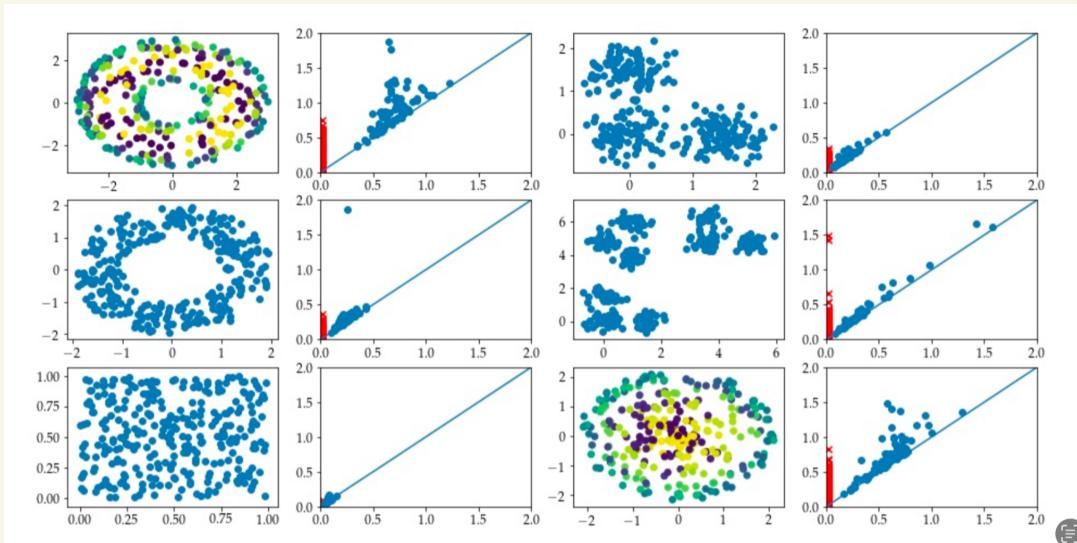
But different spaces might have the same Euler characteristic  
(as well as very different geometry)

# Back to signatures

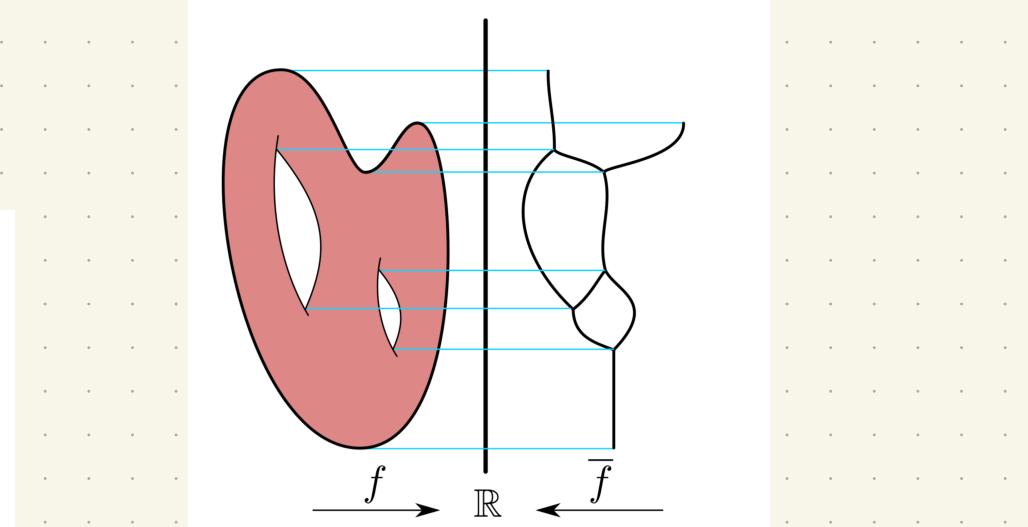
We'll cover a range of possible choices, on a sliding scale of complexity + discriminativity.

graph-based

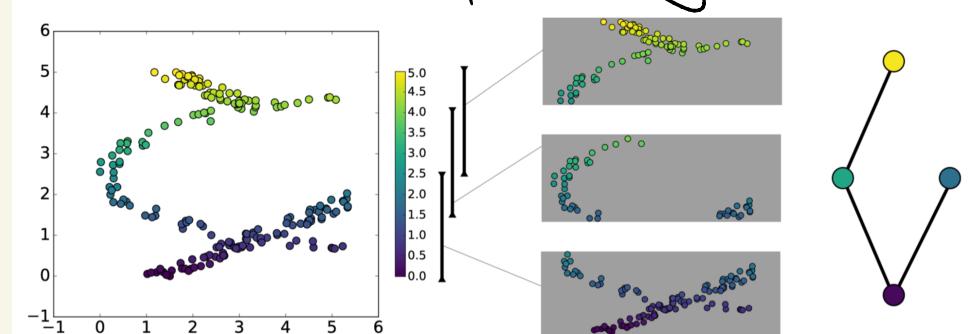
Examples:



Persistent homology  
Algebraic topology)



Reeb & Mapper graphs



## Active research directions

This is a fairly young & vibrant area.

### Emerging directions:

- Machine Learning with TDA
- Time series & dynamic systems
- Parallelization
- Visualization
- Algebraic methods & multi-dimensional persistence
- Many applications: atmospheric data, image processing, biomedical, neuroscience, vision, etc.

# Our goals

Understand the computation + interpretation  
of several commonly used tools in TDA:

- Euler curves
- Persistent homology
- Mapper & Reeb graphs
- Morse-Smale Complexes

In the end, understand what types of  
signatures are likely to be both  
useful & practical on data sets  
(as well as which open source tools exist).

But first - topology!

Chapter 1 covers an intro to topology.

Depending on math, might seem  
obvious, or might seem very hard!

Either is ok.

Worth reading textbook to be sure  
you get main definitions in context,  
& come see me if you have questions.

# Topology

**Topological space**: a set  $T$  with elements (called points) + a set of subsets  $\tau$ , such that

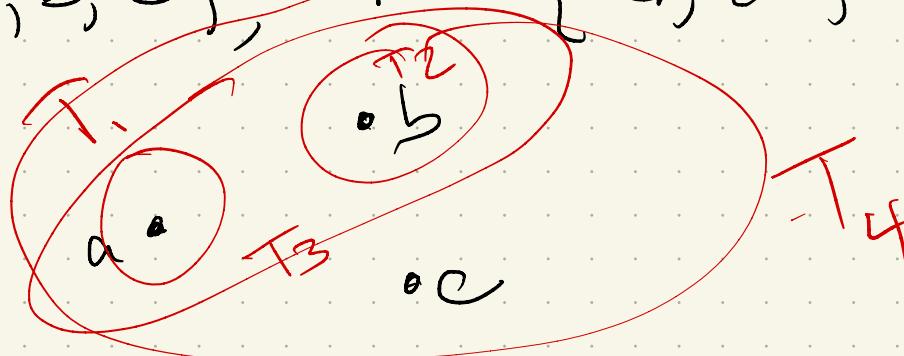
- $\emptyset, T \in \tau$

- $\forall U \subseteq T$ , union of sets in  $U$  is in  $\tau$

- $\forall$  finite  $U \subseteq T$ , intersection of sets in  $U$  is also in  $\tau$

Ex:  $T = \{a, b, c\}$ ,  $\tau = \{\emptyset, \{a\}, \{b\}, \{a, b\}, \{a, b, c\}\}$

Check:



## Metric Space:

a pair  $(\Pi, d)$ , where  $\Pi$  is a set and  
 $d: \Pi \times \Pi \rightarrow \mathbb{R}$  satisfies other:  $d(p, q) \geq 0$

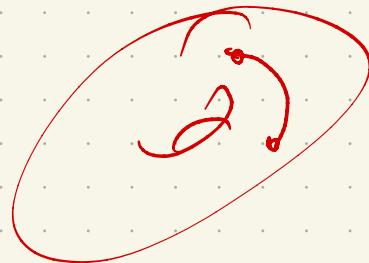
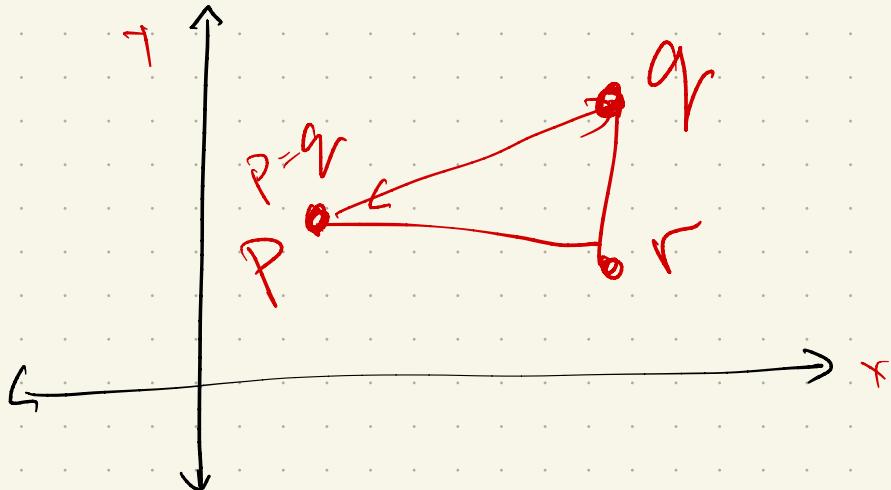
$$\bullet d(p, q) = 0 \Leftrightarrow p = q$$

$$\bullet d(p, q) = d(q, p) \quad \forall p, q \in \Pi$$

~~triangle inequality~~  $d(p, q) \leq d(p, r) + d(r, q) \quad \forall p, q, r \in \Pi$

Example:  $\Pi = \mathbb{R}^2$ ,  $d((u_1, u_2), (v_1, v_2)) =$

$$\sqrt{(u_1 - v_1)^2 + (u_2 - v_2)^2}$$



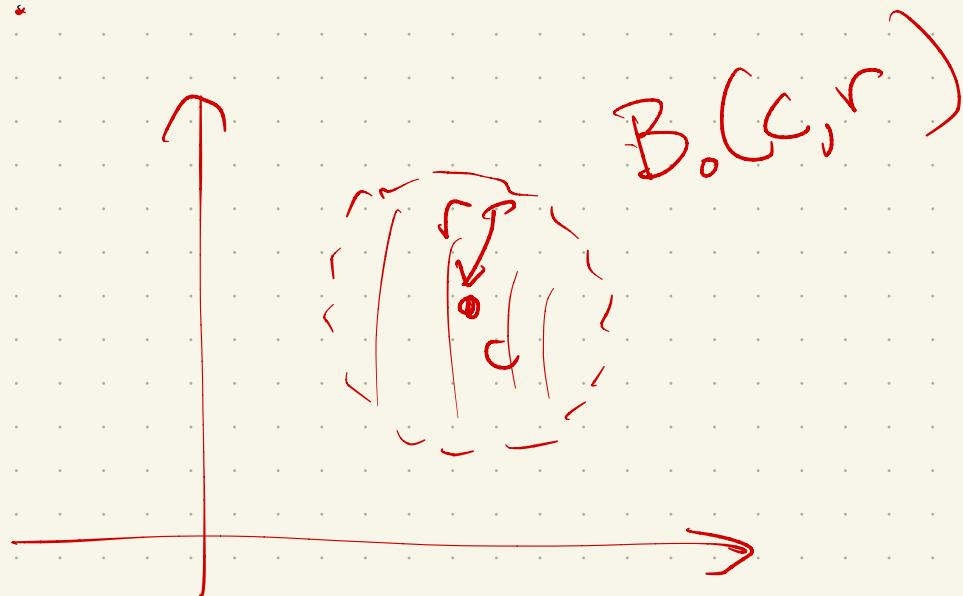
# Metric topology

Given a metric space  $(\mathbb{H}, d)$ , an open metric ball is

$$B_o(c, r) = \{ p \in \mathbb{H} \mid d(p, c) < r \}$$

The metric topology is the set of all metric balls.

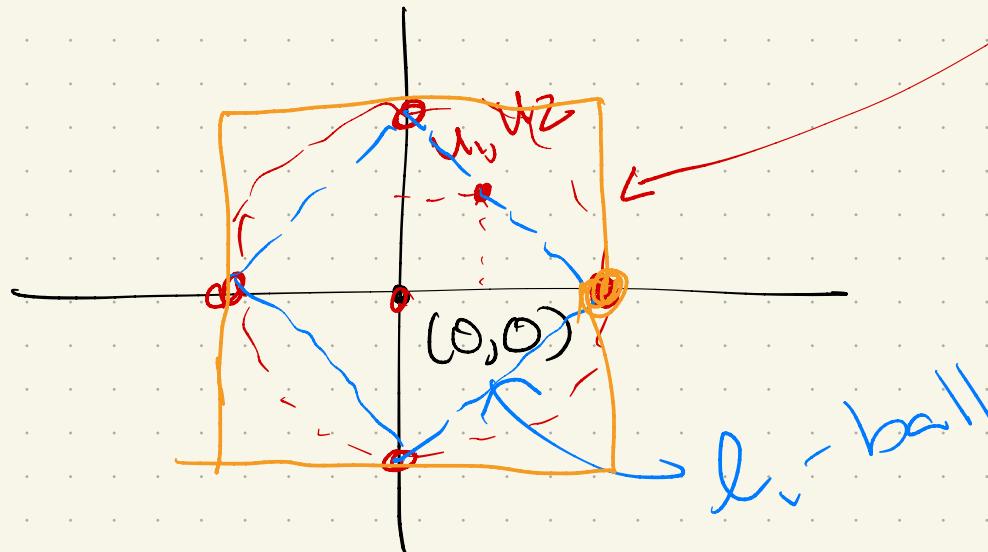
Ex:  $\mathbb{R}^2$  again:



Many different metric topologies!

Fix  $\mathbb{R}^2$ , & let's try  $B_{\underline{0}}(0, 1)$  for:

- $\|u-v\|_1 = \underbrace{|u_1-v_1|}_{\text{l}_1\text{-metric}} + \underbrace{|u_2-v_2|}_{\text{l}_1\text{-metric}} < 1?$
- $\|u-v\|_2 = \sqrt{(u_1-v_1)^2 + (u_2-v_2)^2}$   $\text{l}_2\text{-metric}$
- $\|u-v\|_\infty = \max \{ |u_1-v_1|, |u_2-v_2| \}$   $\text{l}_\infty\text{-metric}$



$\text{l}_p$ -metrics

## Open & closed sets

Fixing a topology  $T$ ,  $U$  is open if  $U \in T$ .  
We say  $U$  is closed if  $T \setminus U$  is open.

Set theory  
complement

Back to first example:

$$T = \{a, b, c\}, \quad \mathcal{T} = \{\emptyset, \{a\}, \{b\}, \{a, b\}, \{a, b, c\}\}$$

Closed sets:

$\mathcal{T}$   
open

$$T - \emptyset = T$$

$$T - \{a, b\} = \{c\}$$

$$T - \{a\} = \{b, c\}$$

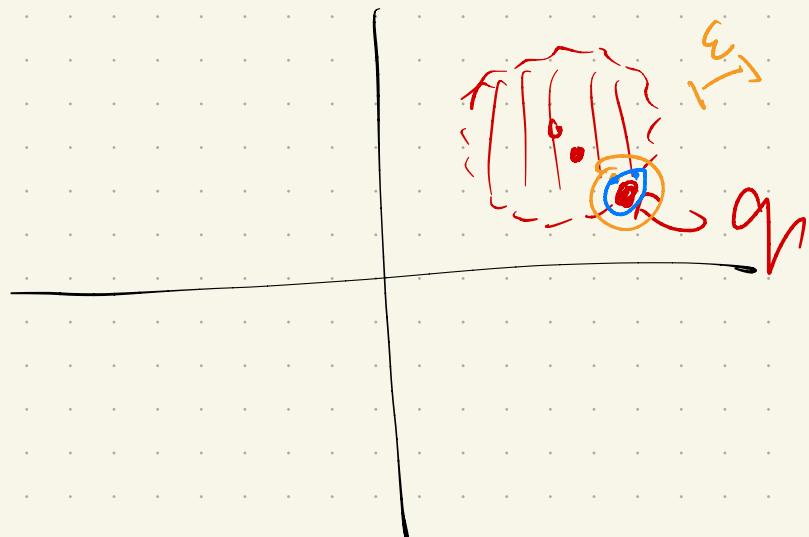
$$T - \{a, b, c\} = \emptyset$$

$$T - \{b\} = \{a, c\}$$

In a metric space, can get some alternate definitions:

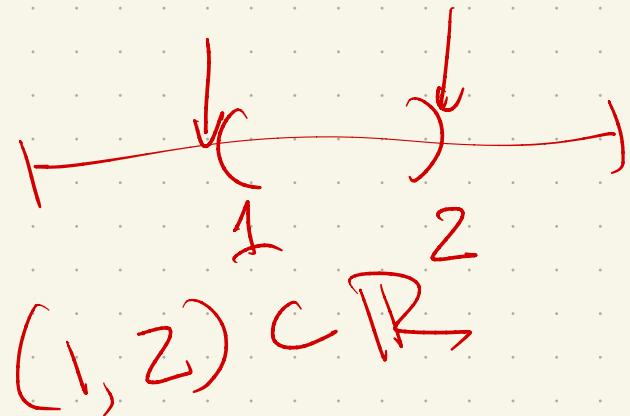
Consider  $Q \subseteq \mathbb{T}$ . A point  $p \in \mathbb{T}$  is a limit point of  $Q$  if  $\forall \varepsilon > 0$ ,  $Q$  contains some point  $q \neq p$  with  $d(pq) < \varepsilon$ .

Example:  $\mathbb{R}^2$  &  $B_0(p, 1)$ :



distance  $< 1$

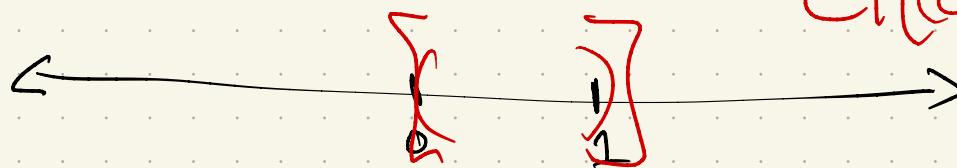
$[1, 2] \subset (1, 2)$



The **closure** of a point set  $Q \subseteq T$  is the set containing  $Q$  and all of its limit points.

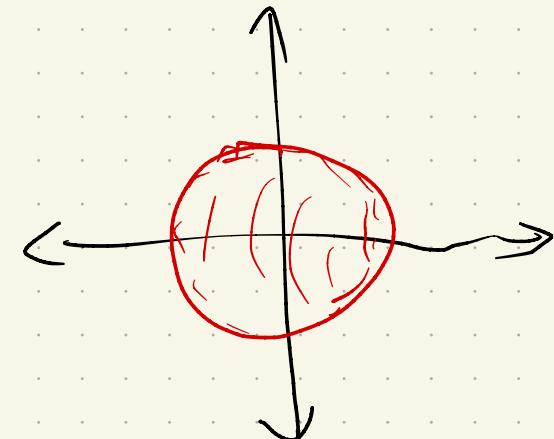
↳ written  $C_1(Q)$ , or  $\overline{Q}$   
 We say  $Q$  is **closed** if  $Q = C_1(Q)$ .

Example:  $(0, 1) \subseteq \mathbb{R}$  is an open ball



$$C_1((0, 1)) = [0, 1]$$

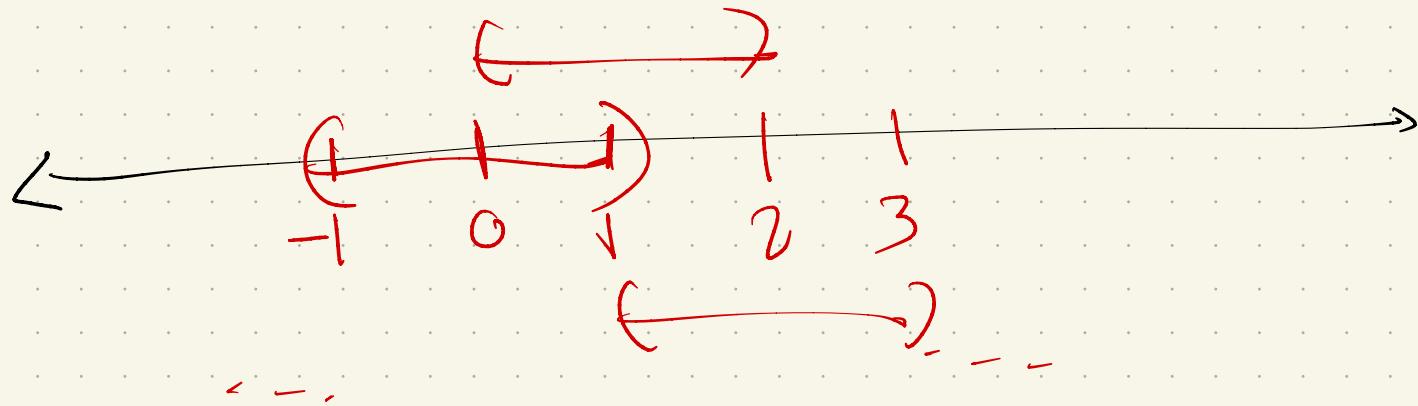
Example:  $\mathbb{R}^2 + B_0(0, 1)$



A **open** (resp, closed) **cover** of a topological space  $(T, \tau)$  is a collection  $C$  of open (resp. closed) sets st.

$$\underline{T} = \bigcup_{c \in C} \underline{c}$$

Example:  $\mathbb{R}$ ,  $C = \{(n-1, n+1) | n \in \mathbb{Z}\}$



A topological space is **disconnected**  
if  $\exists$  2 disjoint nonempty open sets  
 $U, V \in T$  s.t.  $T = U \cup V$ .

(The space is **connected** if it is  
not disconnected.)

Ex:  $A = (1, 2) \cup (3, 4) \subset \mathbb{R}$

Note: **Subspace topology**: Given  $U \subseteq T$ ,  
 $U$  can inherit topology from  $T$  via  
 $\{x \cap U \mid x \in T\}$

Next time:

today 1.1 + 1.2

Maps, homeomorphisms, & homotopies!

(See remainder of Chapter 1)

Overall goal: understand enough about  
maps to get to "nice" functions,  
& start Morse theory

Homework 0: Send me an email!  
(HW1 also posted, but you have  
2 weeks.)