

Persistent connections

Ulrich Bauer

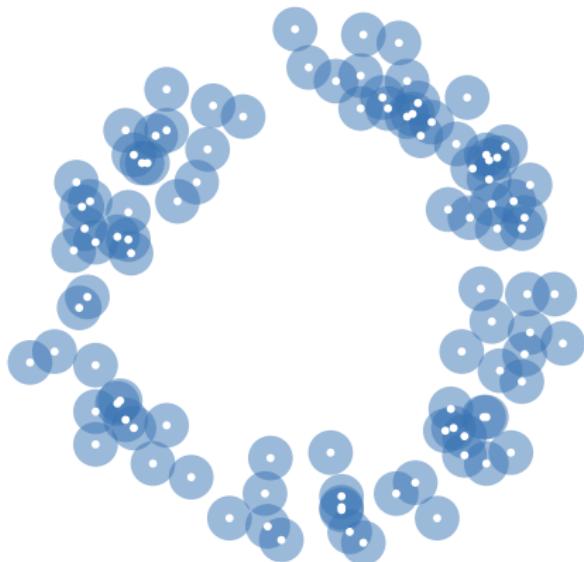
TUM

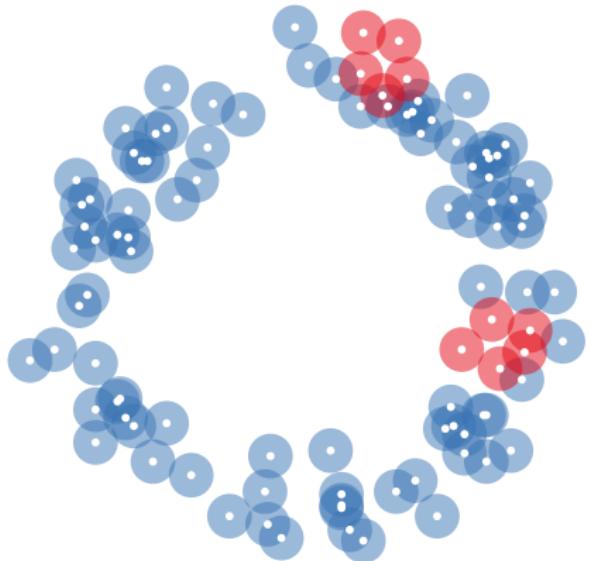
October 13, 2016

SFB days · TU Berlin

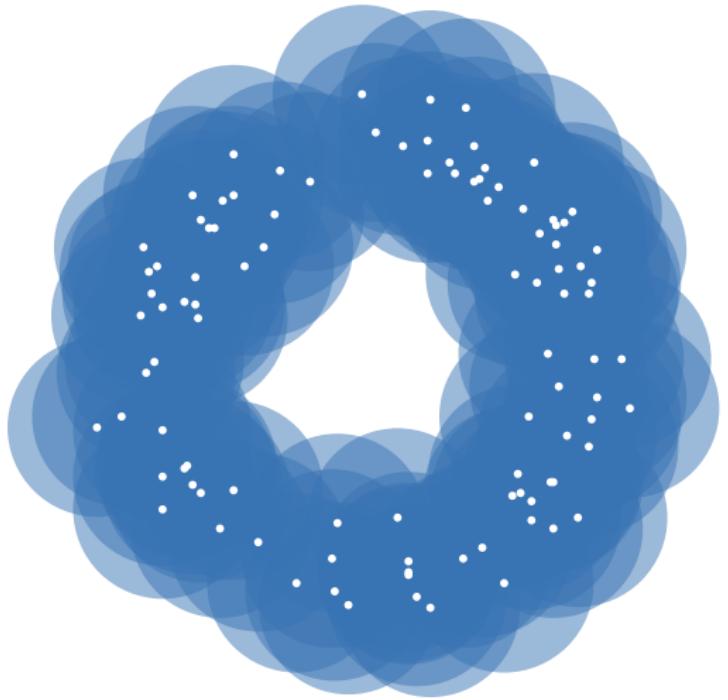
Persistent homology

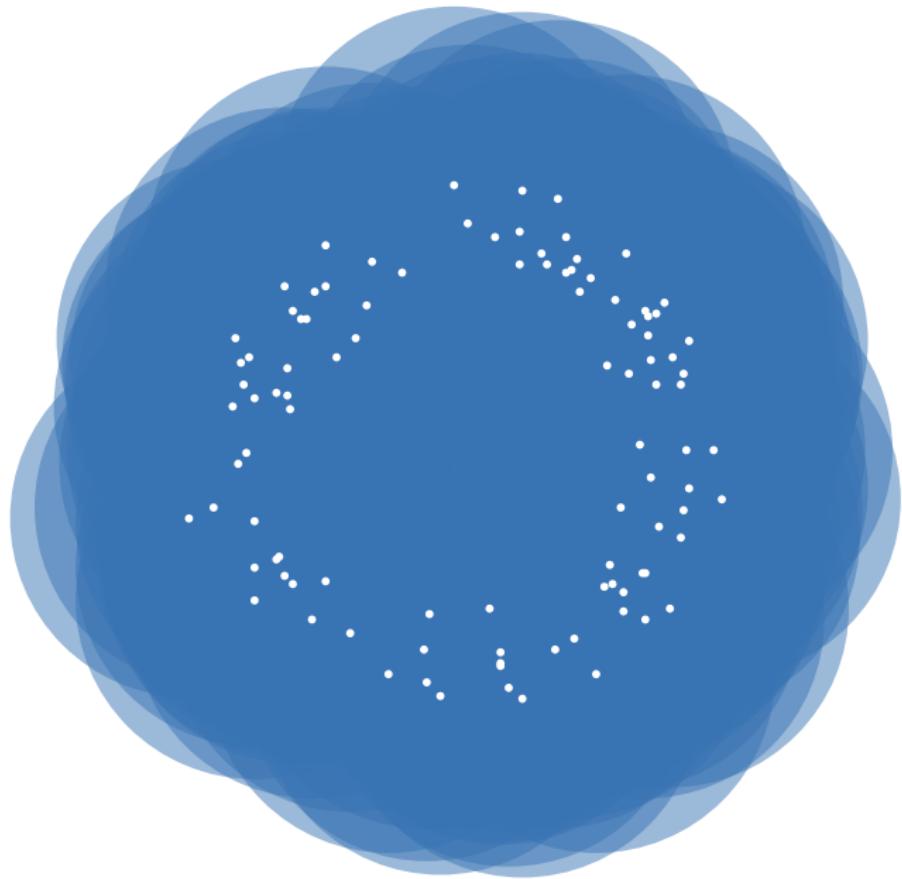


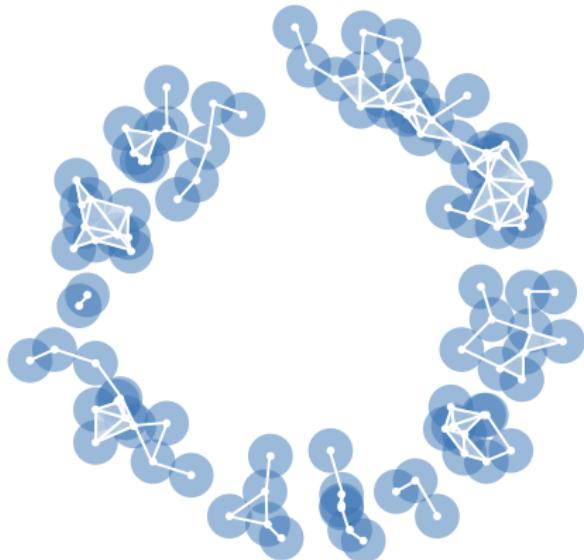


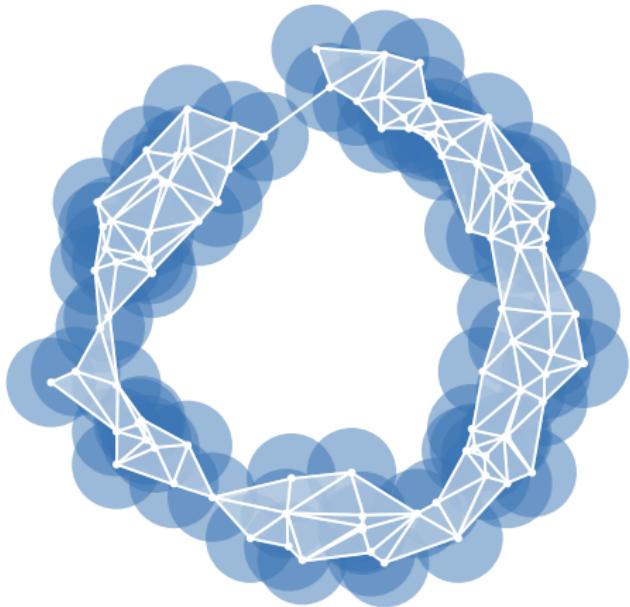


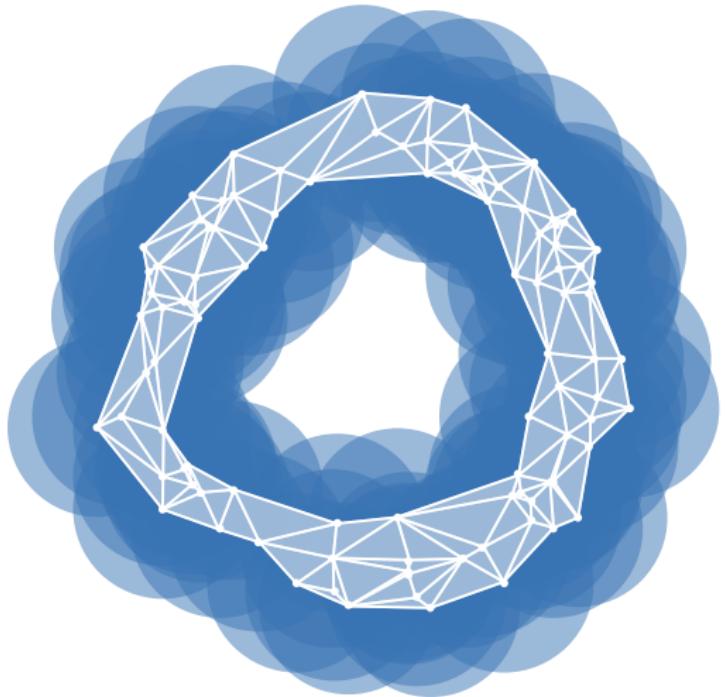


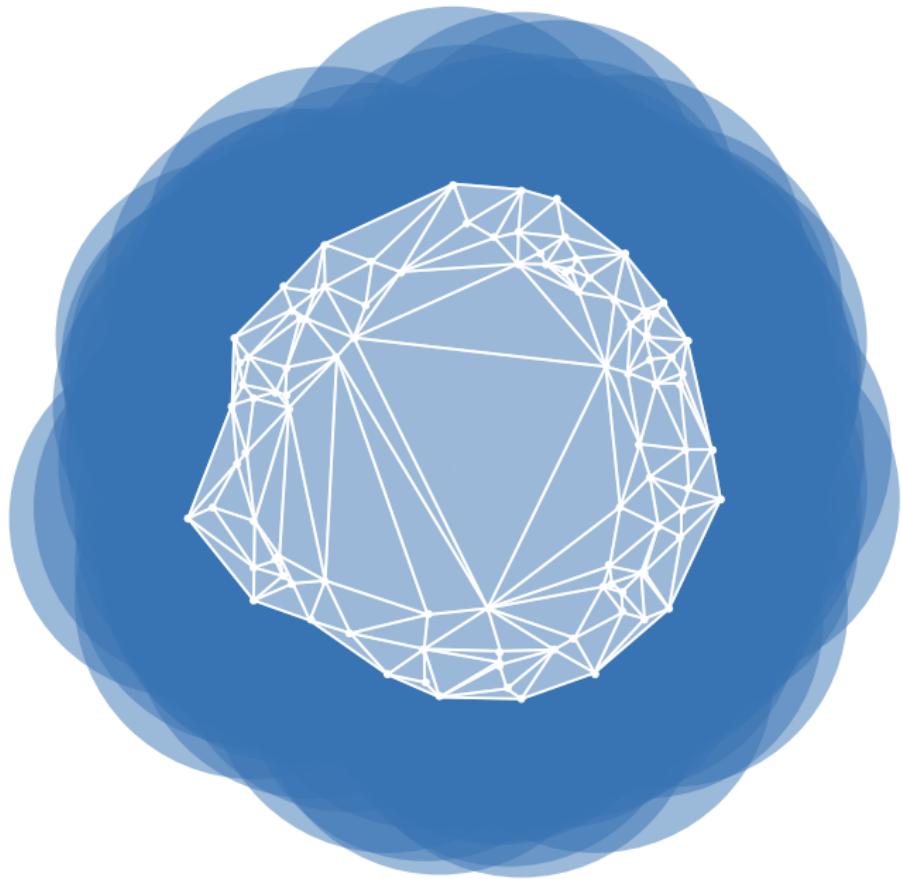




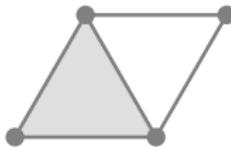






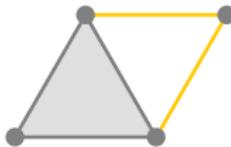


What is homology?



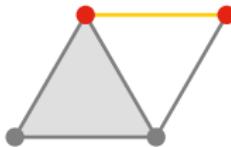
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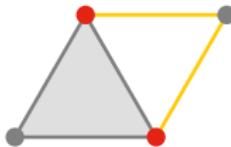
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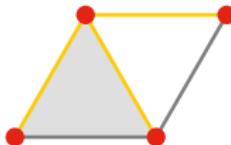
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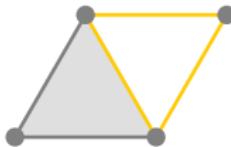
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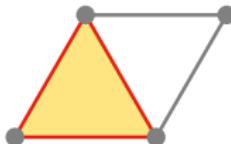
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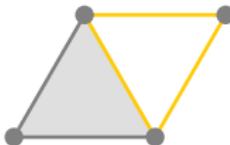
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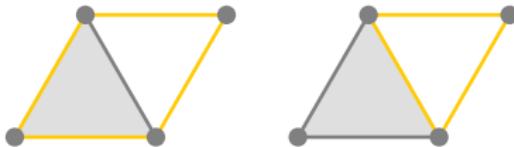
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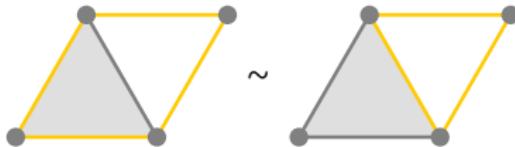
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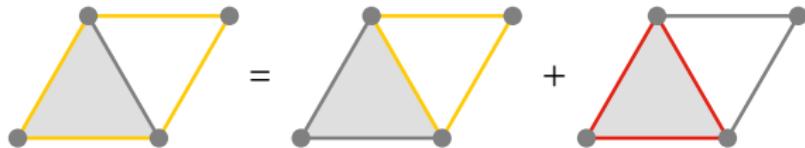
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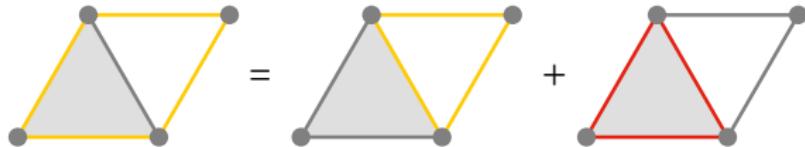
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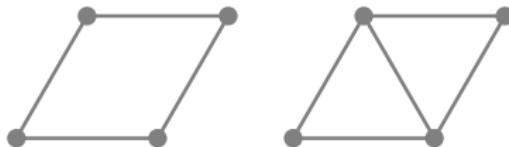
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- The equivalence classes form the *homology group*
$$H_*(K) = Z_*(K)/B_*(K)$$

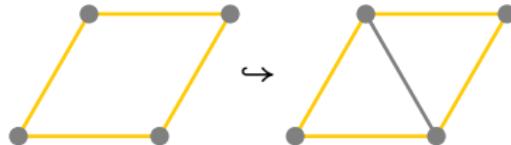
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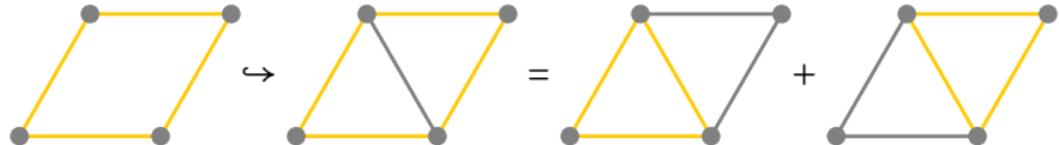


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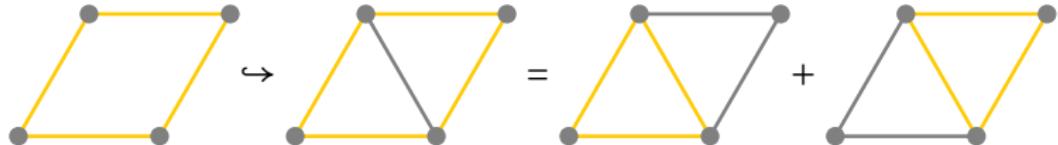


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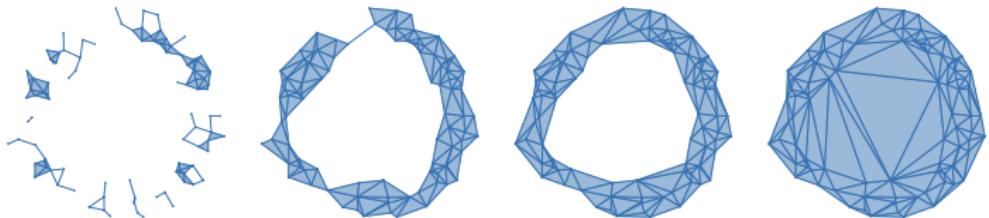


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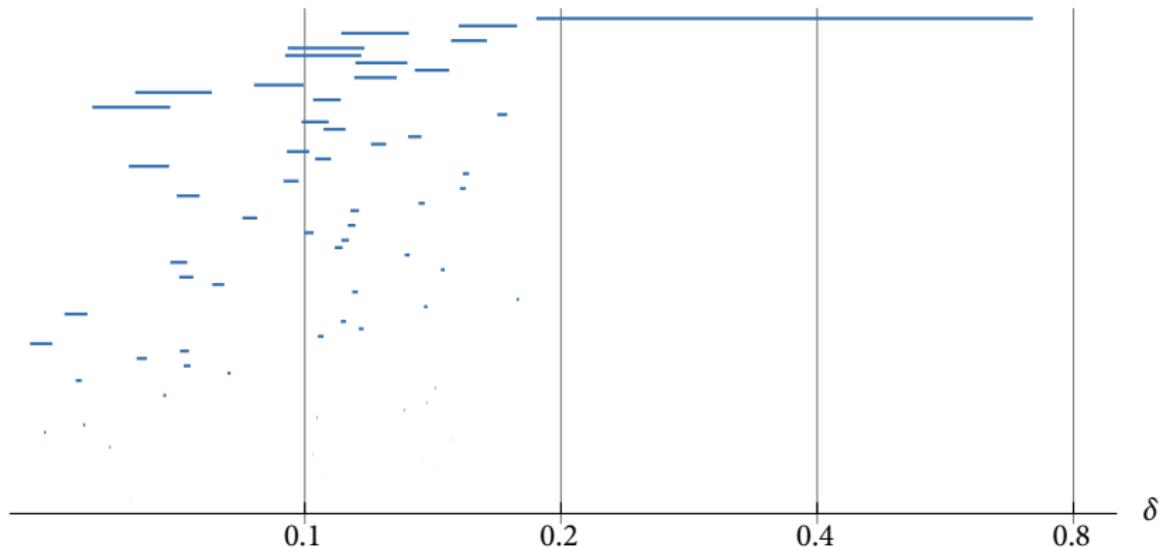
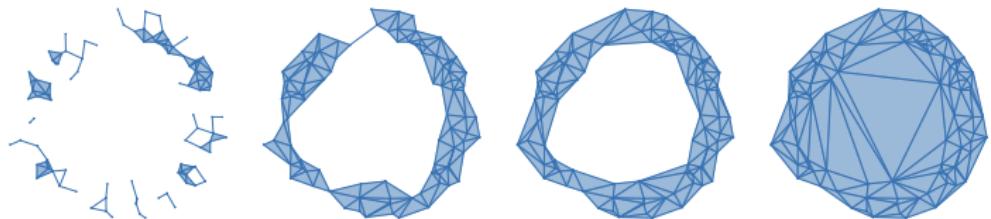
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Homology does not have a canonical basis

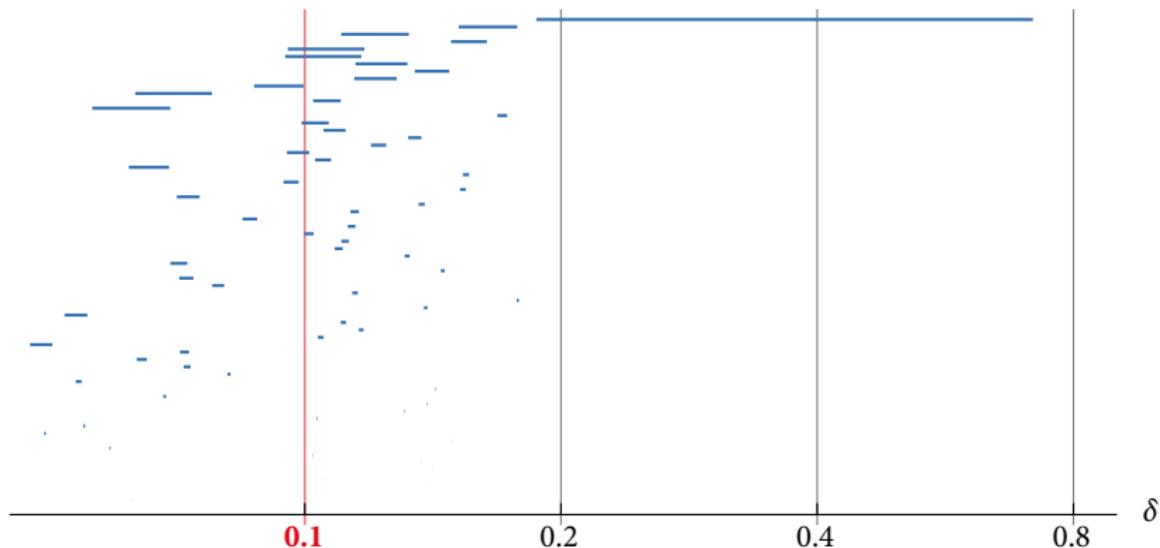
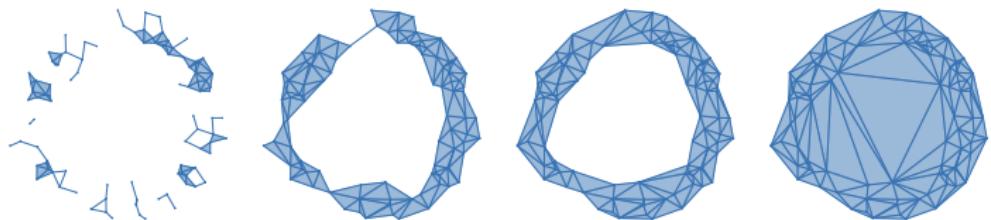
What is persistent homology?



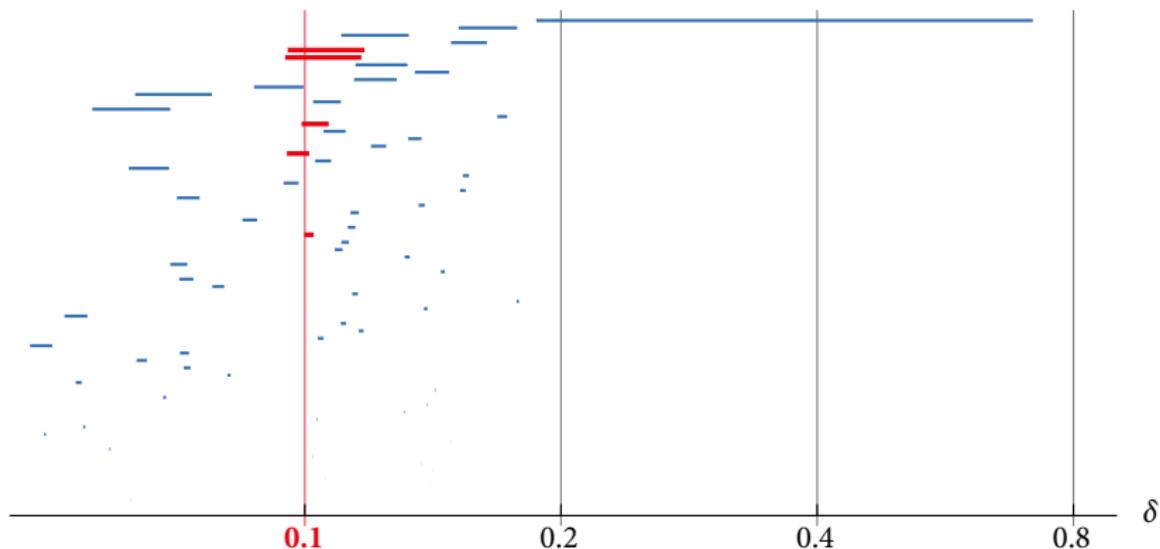
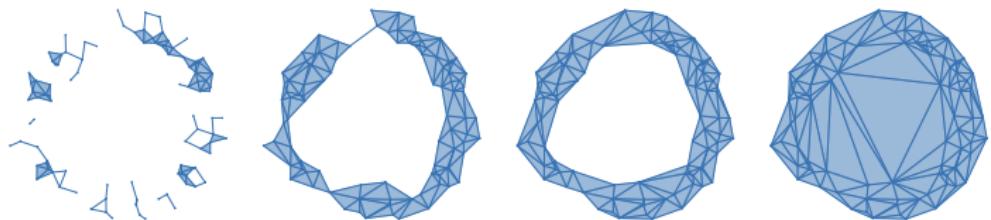
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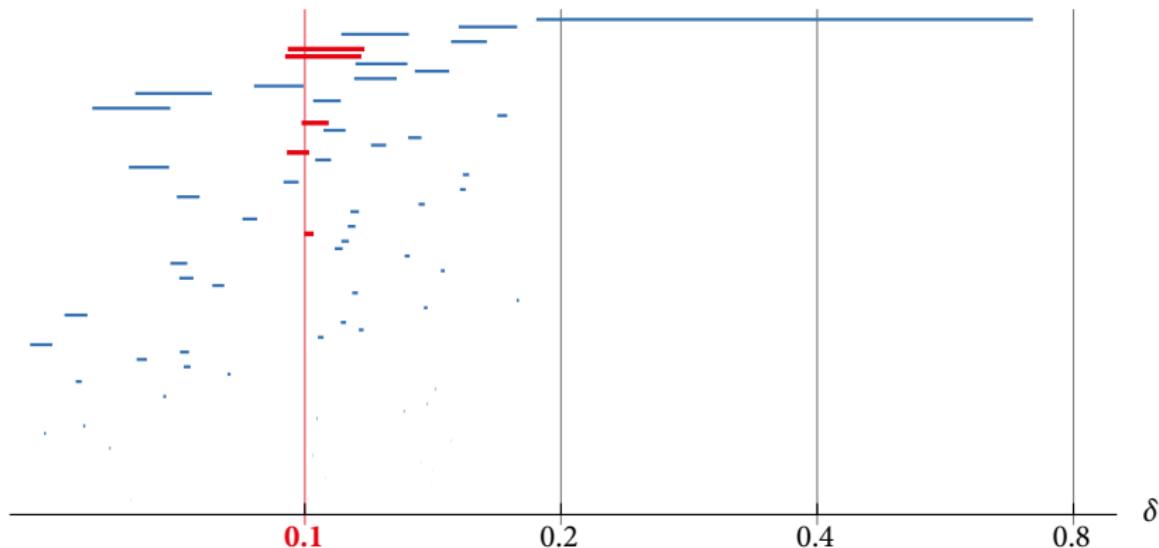
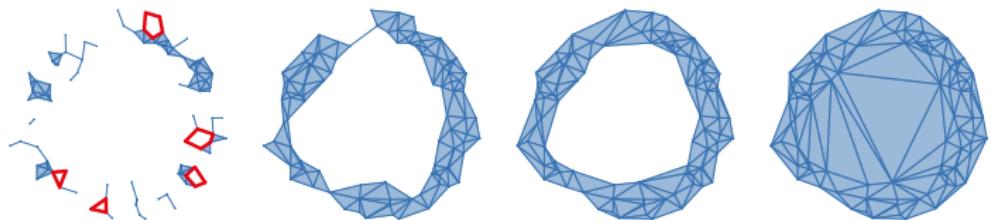
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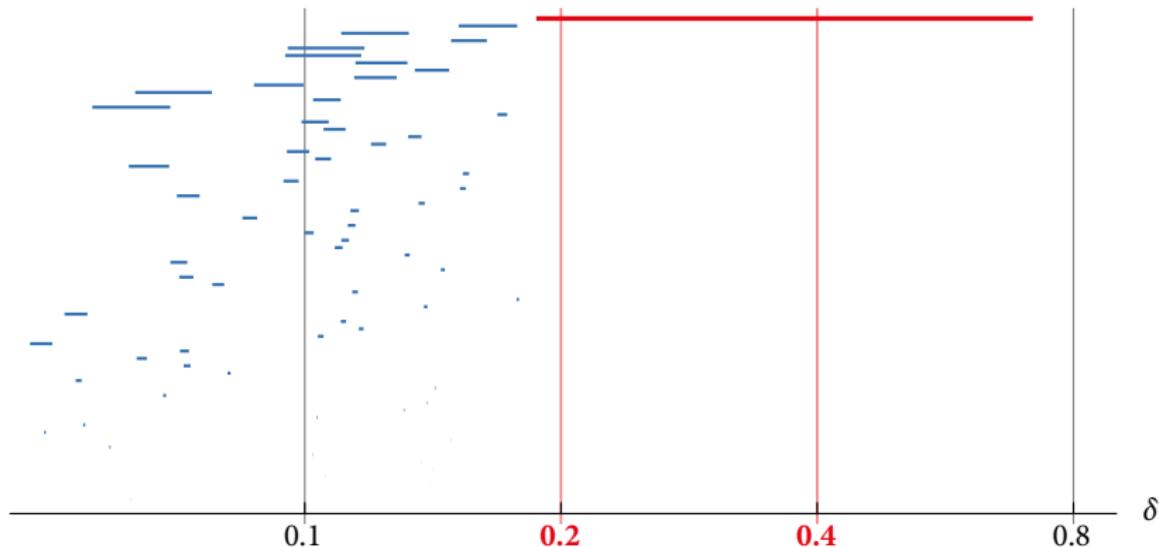
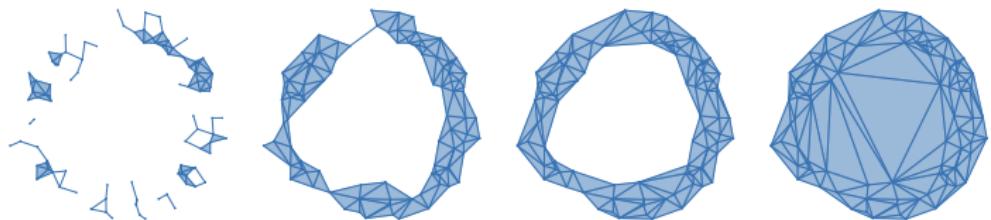
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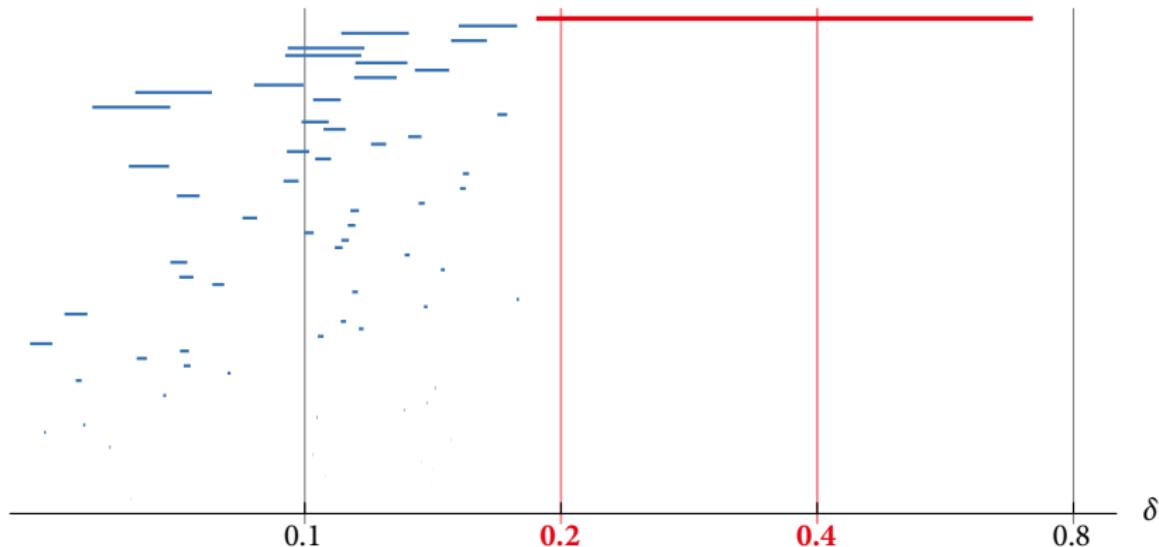
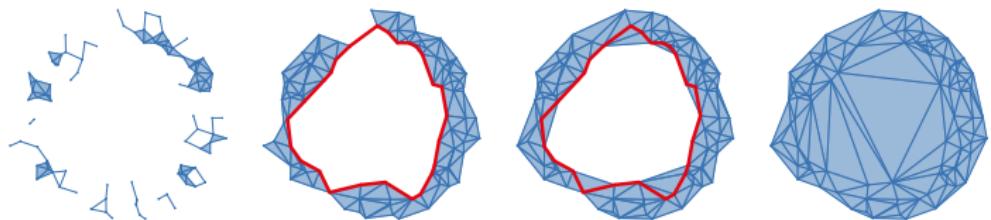
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Numerics

Cauchy's differentiation formula

Differentiation by integration:

- Let $f : U \subseteq \mathbb{C} \rightarrow \mathbb{C}$ be holomorphic,
- $D \subseteq U$ a closed disk, and
- $a \in \text{int } D$.

Then

$$f^{(n)}(a) = \frac{n!}{2\pi i} \oint_{\gamma} \frac{f(z)}{(z-a)^{n+1}} dz$$

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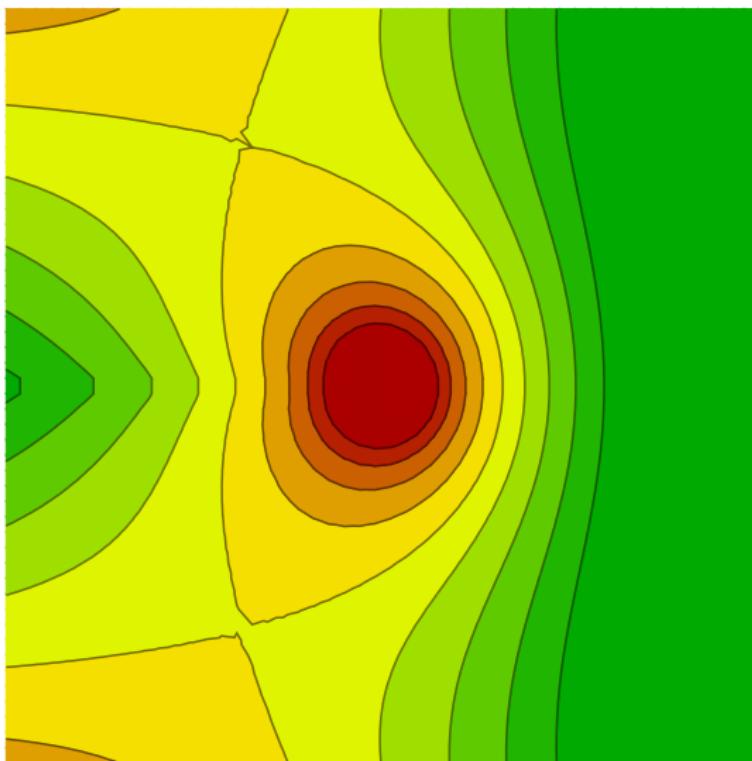
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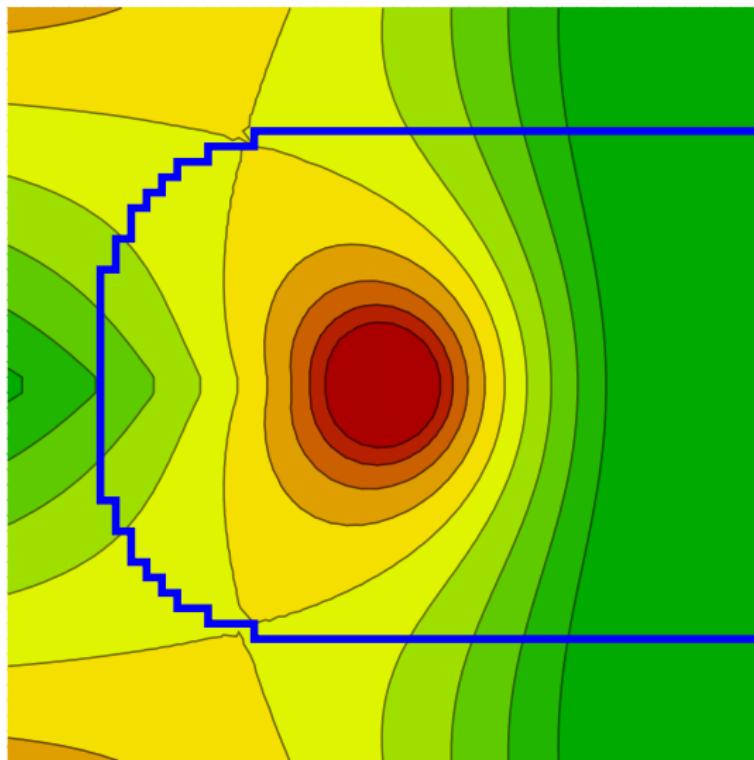
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- But not all cycles are created equal
(for numerical integration)!

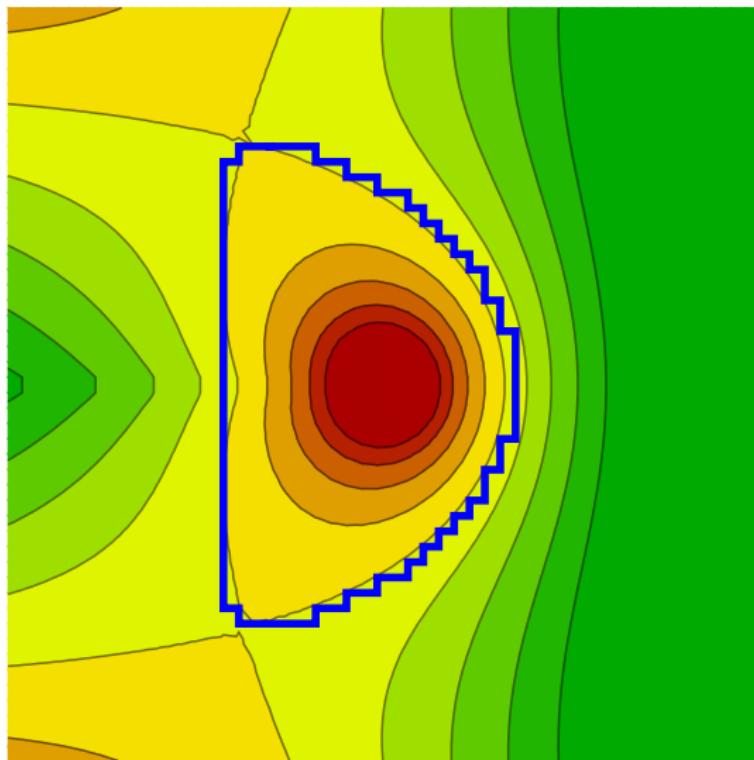
Optimizing condition of contour integration



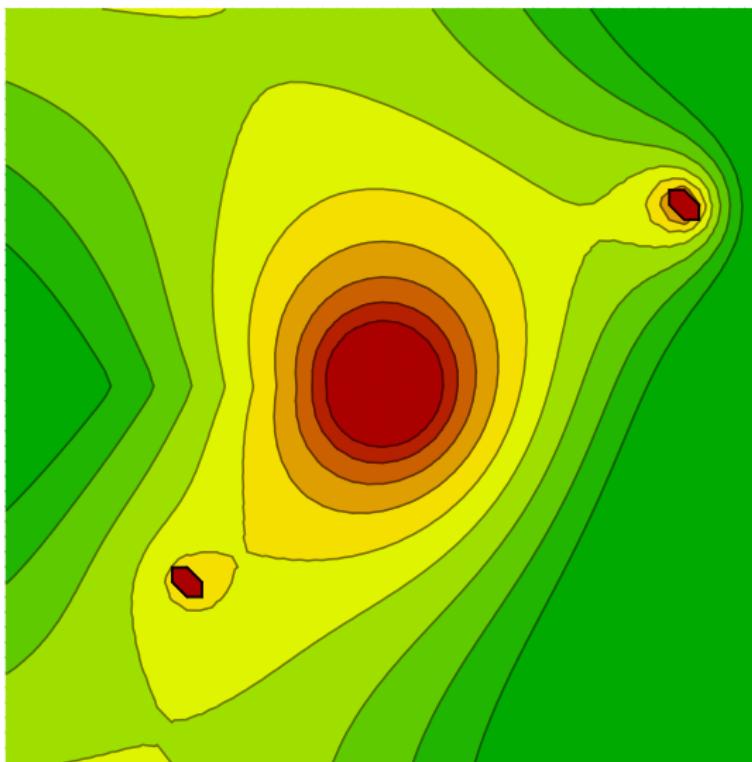
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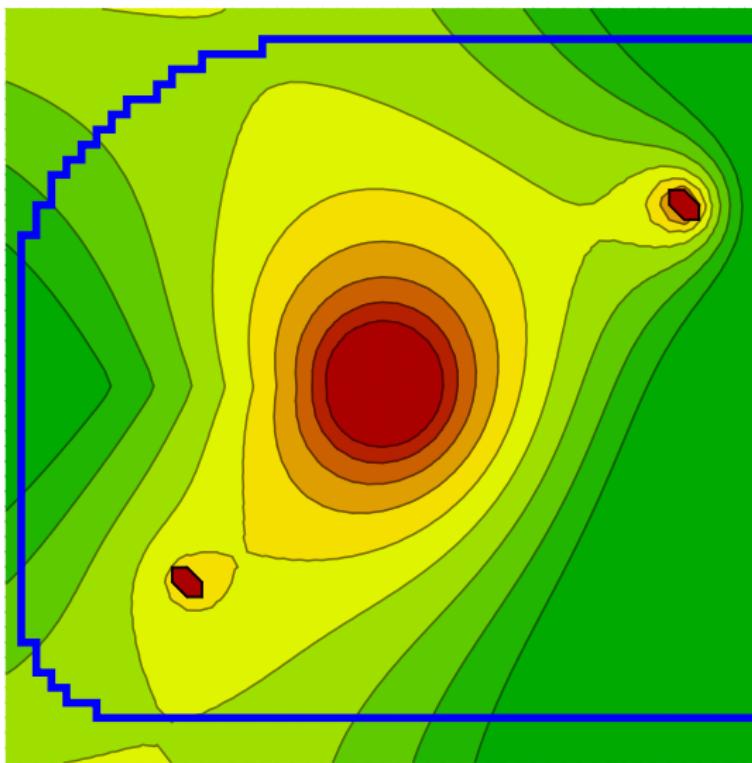
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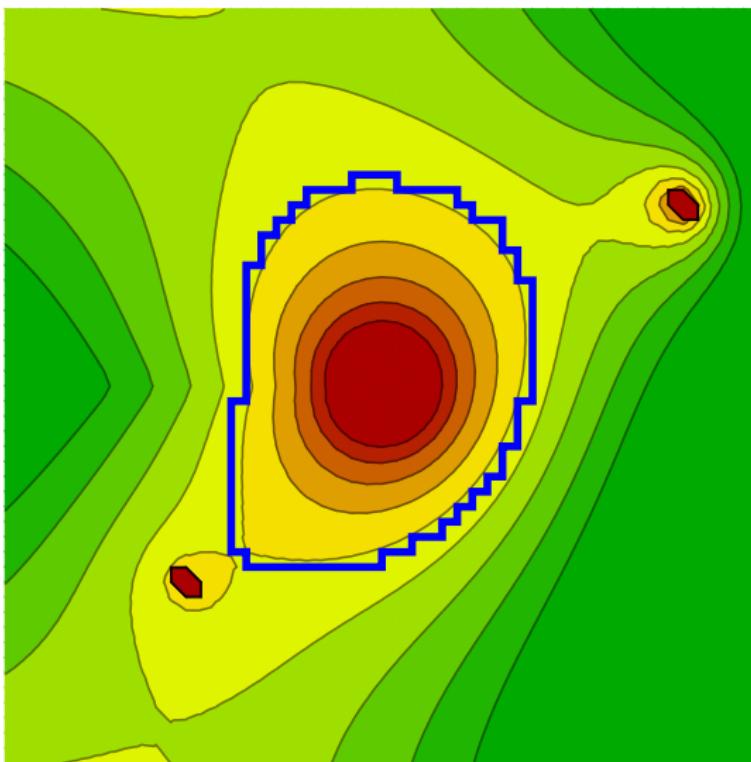
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Computation

Vietoris–Rips filtrations

Consider a finite metric space (X, d) .

The *Vietoris–Rips complex* is the simplicial complex

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Goal:

- compute homology $H_d(\text{Rips}_t(X))$
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- together with induced maps $H_d(\text{Rips}_s(X) \hookrightarrow \text{Rips}_t(X))$
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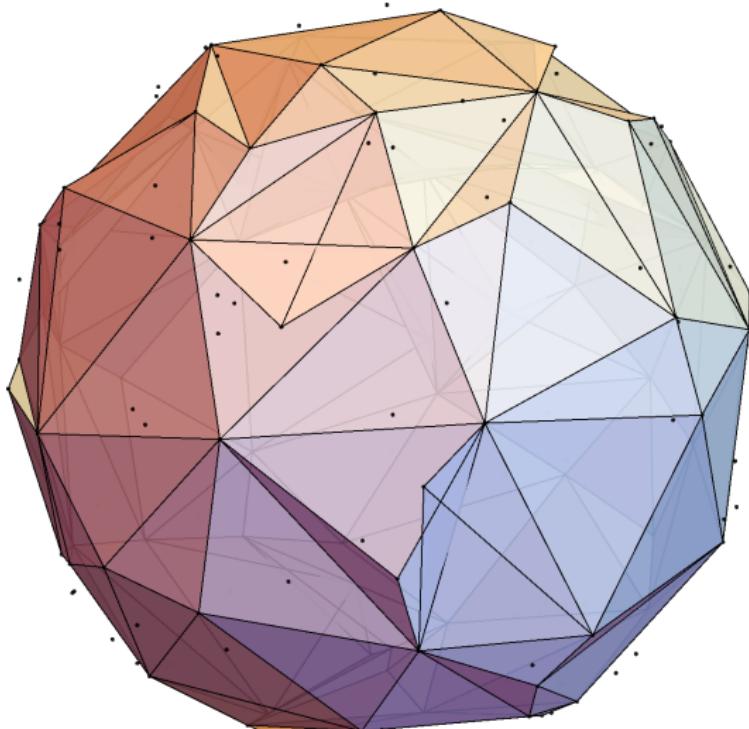
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Note: $\text{Rips}_t(X)$ is the full simplex with vertices X

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Previous software:

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Ripser: 2 seconds, 160 MB

- less than 1000 lines of code in a single C++ file
- no external dependencies

The four special ingredients

The improved performance is based on 4 insights:

- Skip inessential columns
- Compute cohomology
- Implicit boundary matrix
- Apparent pairs

Homology by matrix reduction

Computing homology $H_* = Z_*/B_*$ (recall: $B_* \subseteq Z_* \subseteq C_*$):

- compute basis for boundaries $B_* = \text{im } \partial_*$
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Notation:

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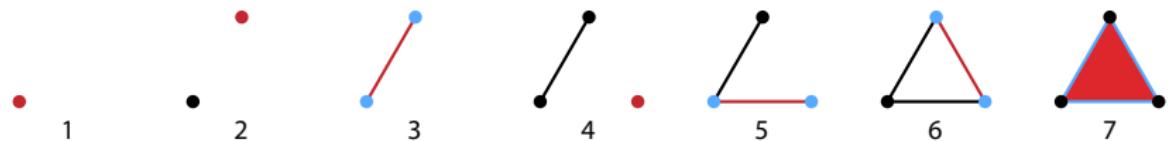
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Result:

- $R = D \cdot V$ is reduced (each column has a unique pivot)
- V is full rank upper triangular

Matrix reduction

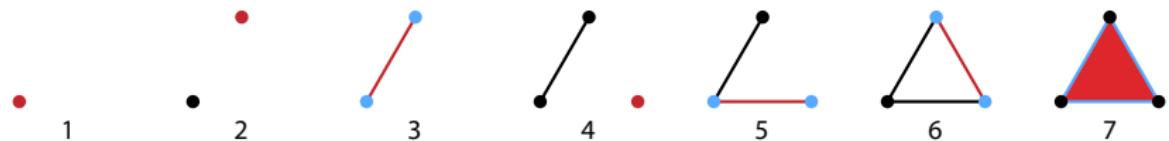


	1	2	3	4	5	6	7
1			1		1		
2			1			1	
3							1
4				1	1		
5						1	
6							1
7							

$= D \cdot$

	1	2	3	4	5	6	7
1	1						
2		1					
3			1				
4				1			
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6						1	
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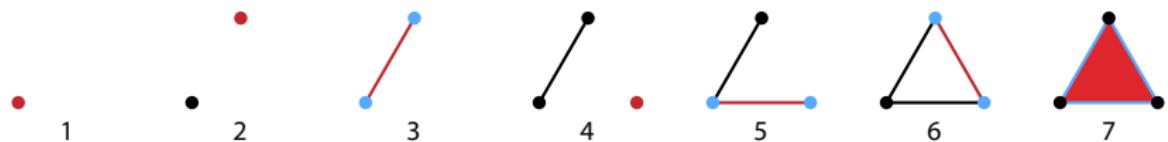


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$= D \cdot$

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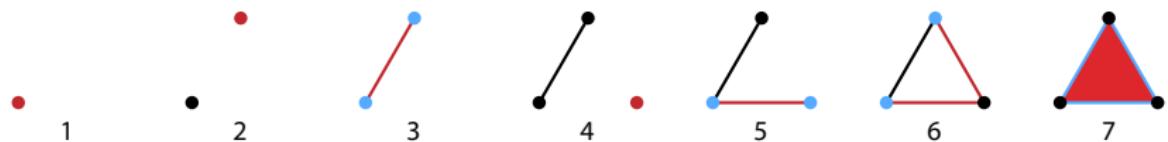


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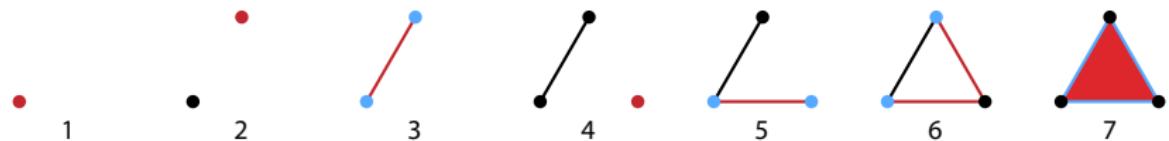


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1			1		1	1	
2			1			1	
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4				1	0		
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6							1
7							

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	1	2	3	4	5	6	7
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2		1					
3			1				
4				1			
5					1	0	
6						1	
7							1

Matrix reduction

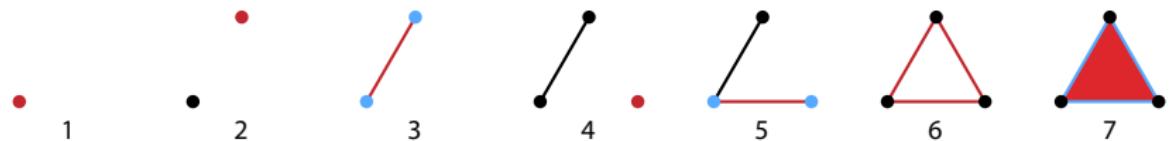


	1	2	3	4	5	6	7
1			1		1	1	
2			1			1	
3							1
4				1			
5						1	
6							1
7							

$= D \cdot$

	1	2	3	4	5	6	7
1	1						
2		1					
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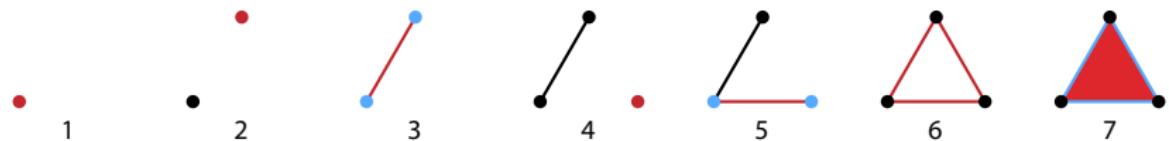


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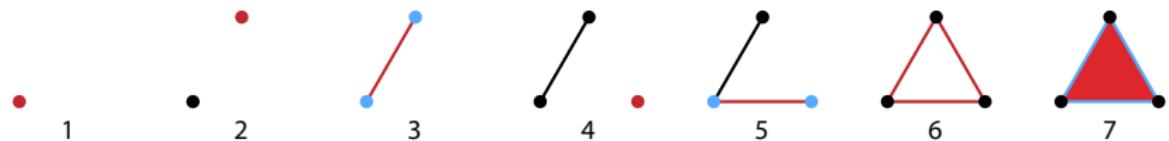


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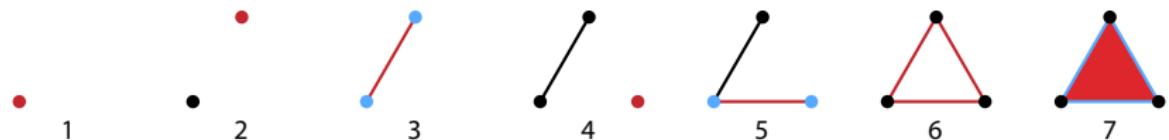


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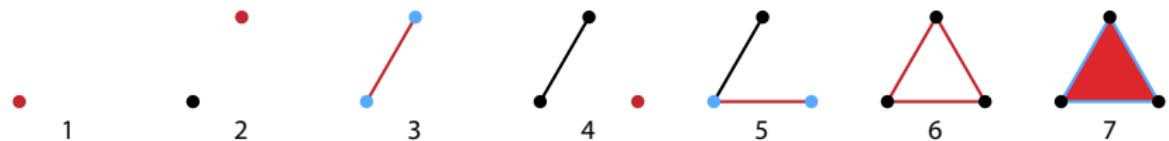


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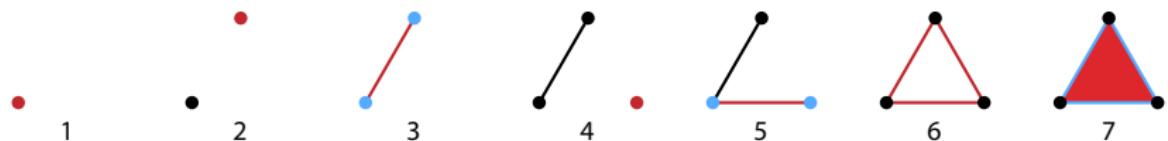


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Skipping inessential columns

Assume that

- $R = D \cdot V$ is reduced (unique column pivots)
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Those columns can be skipped entirely!

Persistent cohomology

We have seen: many columns of $R = D \cdot V$ are not needed

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For persistence barcodes in low dimensions $d < k$:

- Number of skipped indices for reducing D^T (cohomology) is much larger than for D (homology)
 - reducing boundary matrix produces basis for $H_k(K_k)$, which is not needed
- The resulting persistence barcode is the same

Implicit boundary matrix

Typical approach:

- Boundary matrix with respect to filtration-ordered basis
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Approach for Ripser: *don't store what you can compute*

- Boundary matrix with respect to fixed basis order (lexicographic)
- Implicitly defined by number of vertices, dimension
- Row/columns recomputed instead of stored
- Store only reduction coefficient matrix V
- recompute columns of $R = D \cdot V$ from D, V
 - Typically, V is much sparser and smaller than R

Apparent pairs

Apparent pairs

- provide a close connection between persistence and discrete Morse theory
- are both persistence pairs and Morse pairs
- typically cover almost all simplices in the Rips filtration
- provide a shortcut for computation

Natural filtration settings

Typical assumptions on the filtration:

- general filtration persistence (in theory)
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Conclusion:

- Discrete Morse theory sits in the middle between persistence and persistence (hah!)

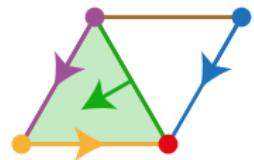
Combinatorics

Discrete Morse theory

Definition (Forman 1998)

A *discrete vector field* on a cell complex is a partition of the set of simplices into

- singleton sets $\{\phi\}$ (*critical cells*), and
- pairs $\{\sigma, \tau\}$, where σ is a facet of τ .

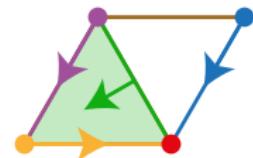


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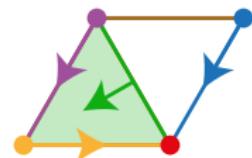
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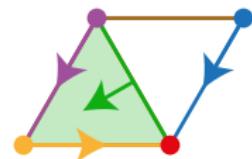


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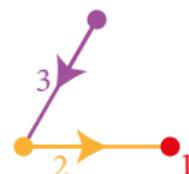
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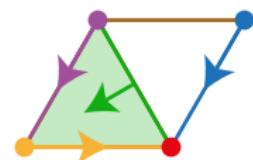


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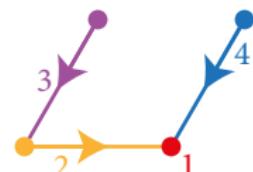
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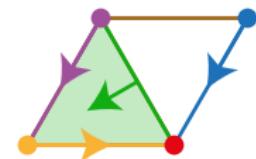


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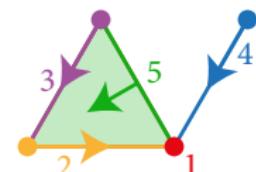
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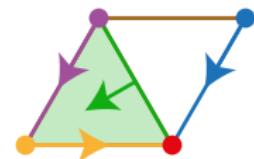


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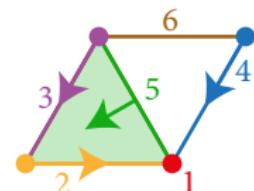
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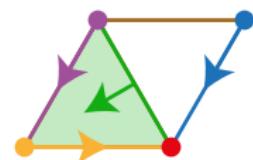


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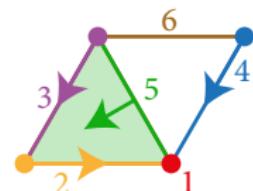
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Fundamental theorem of discrete Morse theory

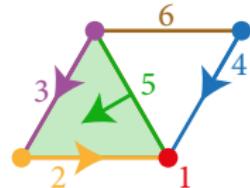
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If $(s, t]$ contains no critical value of f ,
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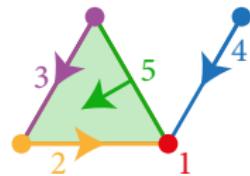


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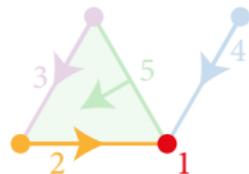


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This homotopy equivalence is compatible with the filtration.

Corollary

K and M have isomorphic persistent homology
(with regard to the sublevel set filtration of f).

Morse pairs and persistence pairs

Consider a *Morse filtration* (one or two simplices at a time).

Morse pair (σ, τ) :

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Morse pair (σ, τ) :

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Consider a *simplexwise filtration* (one simplex at a time).

Persistence pair (σ, τ) :

- inserting simplex σ creates a new *homological* feature
- inserting τ destroys that feature again

From Morse theory to persistence and back

Proposition (from Morse to persistence)

The pairs of a Morse filtration are apparent 0-persistence pairs for the simplexwise refinement of the filtration.

Apparent persistence pair (σ, τ) :

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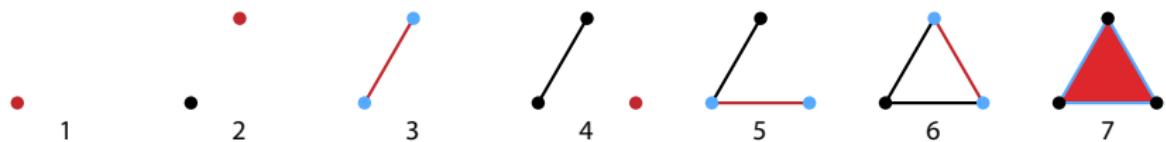
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Consider an arbitrary filtration with a simplexwise refinement.

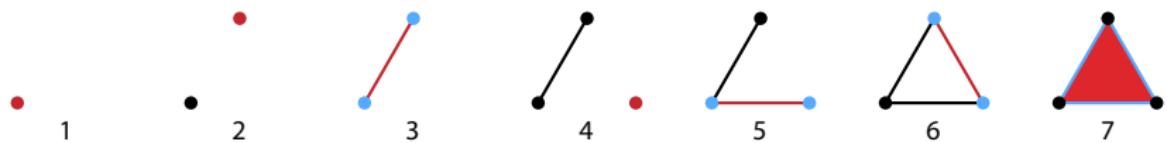
The apparent 0-persistence pairs yield a Morse filtration

- *refining the original one, and*
- *coarsening the simplexwise one.*

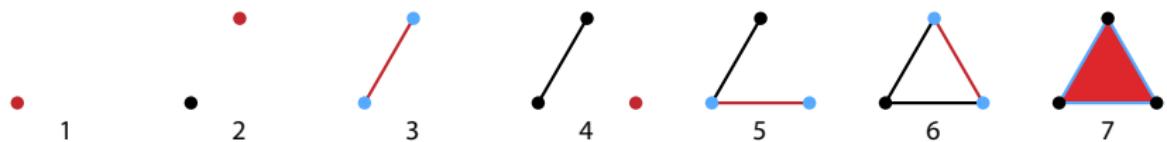
Example: apparent pairs



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$$D = \begin{array}{|c|c|c|c|c|c|c|c|} \hline & 1 & 2 & 3 & 4 & 5 & 6 & 7 \\ \hline 1 & & & 1 & & 1 & & \\ \hline 2 & & & 1 & & & 1 & \\ \hline 3 & & & & & & & 1 \\ \hline 4 & & & & & 1 & 1 & \\ \hline 5 & & & & & & & 1 \\ \hline 6 & & & & & & & 1 \\ \hline 7 & & & & & & & \\ \hline \end{array}$$

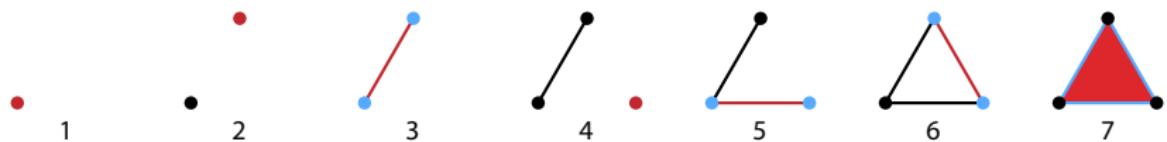
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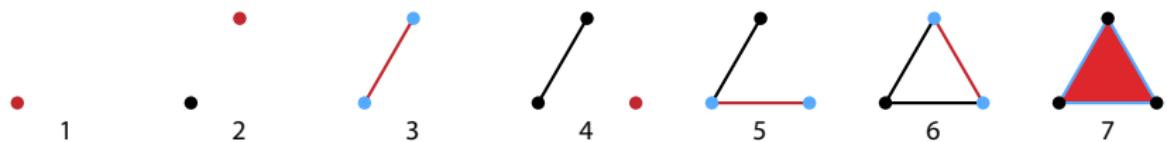
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- Special cases have appeared in literature before
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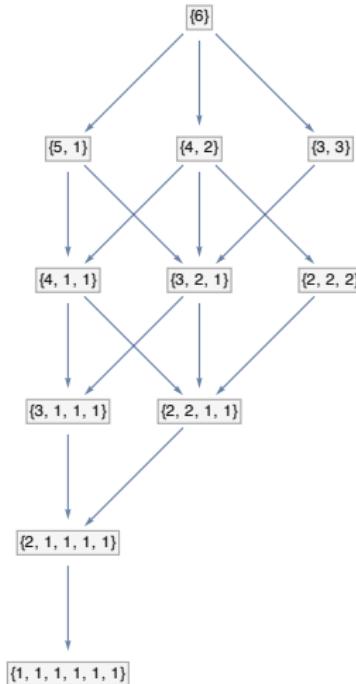
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- Simple and important example: lexicographic order, given by total order on vertices
- Special cases have appeared in literature before
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- Often, this discrete Morse function has good properties (few critical points)

Applications to topological combinatorics

Example: order complex of integer partitions



History

When was persistent homology invented?

- [Edelsbrunner/Letscher/Zomorodian 2000]

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When was persistent homology invented first?

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ANNALS OF MATHEMATICS
Vol. 41, No. 2, April, 1940

RANK AND SPAN IN FUNCTIONAL TOPOLOGY

BY MARSTON MORSE

(Received August 9, 1939)

1. Introduction.

The analysis of functions F on metric spaces M of the type which appear in variational theories is made difficult by the fact that the critical limits, such as absolute minima, relative minima, minimax values etc., are in general infinite in number. These limits are associated with relative k -cycles of various dimensions and are classified as 0-limits, 1-limits etc. The number of k -limits suitably counted is called the k^{th} type number m_k of F . The theory seeks to establish relations between the numbers m_k and the connectivities p_k of M . The numbers p_k are finite in the most important applications. It is otherwise with the numbers m_k .

The theory has been able to proceed provided one of the following hypotheses is satisfied. The critical limits cluster at most at $+\infty$; the critical points are isolated;¹ the problem is defined by analytic functions; the critical limits taken in their natural order are well-ordered. These conditions are not generally fulfilled. The generality of the theory rested upon the fact that the cases treated approximate in a certain sense the most general problems which it is

When was persistent homology invented first?

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Any time Exact homomorphism sequences in homology theory ed.ac.uk [PDF]

JL Kelley, E Pitcher - Annals of Mathematics, 1947 - JSTOR

The developments of this paper stem from the attempts of one of the authors to deduce relations between homology groups of a complex and homology groups of a complex which is its image under a simplicial map. Certain relations were deduced (see [EP 1] and [EP 2] ...)

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Sort by relevance Marston Morse and his mathematical works ams.org [PDF]

R Bott - Bulletin of the American Mathematical Society, 1980 - ams.org

American Mathematical Society. Thus Morse grew to maturity just at the time when the subject of Analysis Situs was being shaped by such masters as Poincaré, Veblen, LEJ Brouwer, GD Birkhoff, Lefschetz and Alexander, and it was Morse's genius and destiny to ...

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Unstable minimal surfaces of higher topological structure

M Morse, CB Tompkins - Duke Math. J., 1941 - projecteuclid.org

1. Introduction. We are concerned with extending the calculus of variations in the large to multiple integrals. The problem of the existence of minimal surfaces of unstable type contains many of the typical difficulties, especially those of a topological nature. Having studied this ...

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[PDF] Persistence in discrete Morse theory psu.edu [PDF]

U Bauer - 2011 - Citeseer

The goal of this thesis is to bring together two different theories about critical points of a scalar function and their relation to topology: Discrete Morse theory and Persistent homology. While the goals and fundamental techniques are different, there are certain

When was persistent homology invented first?

BULLETIN (New Series) OF THE
AMERICAN MATHEMATICAL SOCIETY
Volume 3, Number 3, November 1980

MARSTON MORSE AND HIS MATHEMATICAL WORKS

BY RAOUL BOTT¹

1. Introduction. Marston Morse was born in 1892, so that he was 33 years old when in 1925 his paper *Relations between the critical points of a real-valued function of n independent variables* appeared in the Transactions of the American Mathematical Society. Thus Morse grew to maturity just at the time when the subject of Analysis Situs was being shaped by such masters² as Poincaré, Veblen, L. E. J. Brouwer, G. D. Birkhoff, Lefschetz and Alexander, and it was Morse's genius and destiny to discover one of the most beautiful and far-reaching relations between this fledgling and Analysis; a relation which is now known as *Morse Theory*.

In retrospect all great ideas take on a certain simplicity and inevitability, partly because they shape the whole subsequent development of the subject. And so to us, today, Morse Theory seems natural and inevitable. However one only has to glance at these early papers to see what a tour de force it was in the 1920's to go from the mini-max principle of Birkhoff to the Morse inequalities, let alone extend these inequalities to function spaces, so that by the early 30's Morse could establish the theorem that for any Riemann

When was persistent homology invented first?

inequalities between the dimensions of the π_i and those of $H(\pi_i)$. Thus the Morse inequalities already reflect a certain part of the “Spectral Sequence magic”, and a modern and tremendously general account of Morse’s work on rank and span in the framework of Leray’s theory was developed by Deheuvels [D] in the 50’s.

Unfortunately both Morse’s and Deheuvel’s papers are not easy reading. On the other hand there is no question in my mind that the papers [36] and [44] constitute another tour de force by Morse. Let me therefore illustrate rather than explain some of the ideas of the rank and span theory in a very simple and tame example.

In the figure which follows I have drawn a homeomorph of $M = S^1$ in the plane, and I will be studying the height function $F = y$ on M .

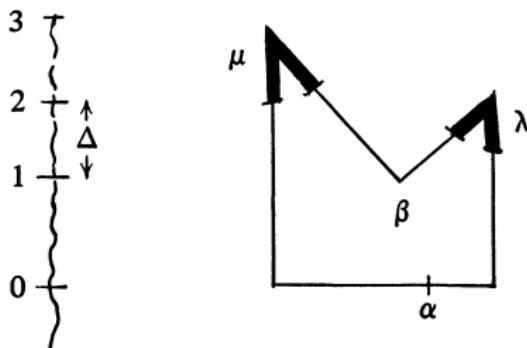


FIGURE 8

The values a where $H(a, a^-) \neq 0$ are indicated on the left, and correspond-

Morse's functional topology

Early precursors of persistence and spectral sequences:

- *F-homology classes*: similar to persistent homology
 - *inferior/superior cycle limits*: birth/death
- *k-caps*: related to elements of spectral sequence
 - *cap span*: persistence

Morse's functional topology

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- applies to a broad class of functions on metric spaces
(not necessarily continuous)

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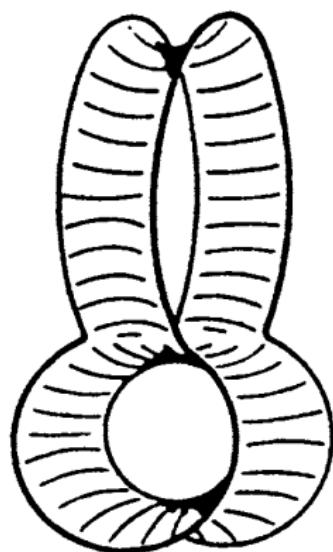
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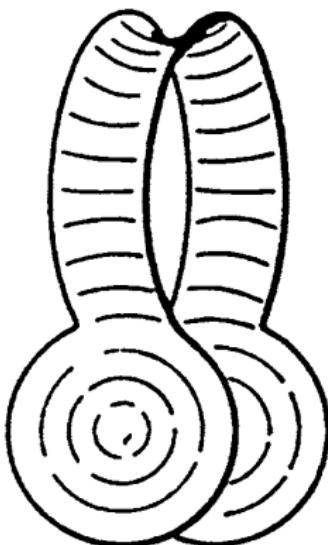
Key aspects:

- applies to a broad class of functions on metric spaces (not necessarily continuous)
- uses Vietoris homology with field coefficients
- inclusions of sublevel sets have homology of finite rank (*q-tame* persistent homology)
- focus on controlled behavior in pathological cases

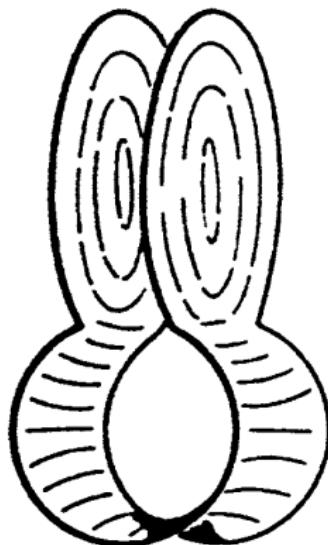
Motivation and application: minimal surfaces



(a)



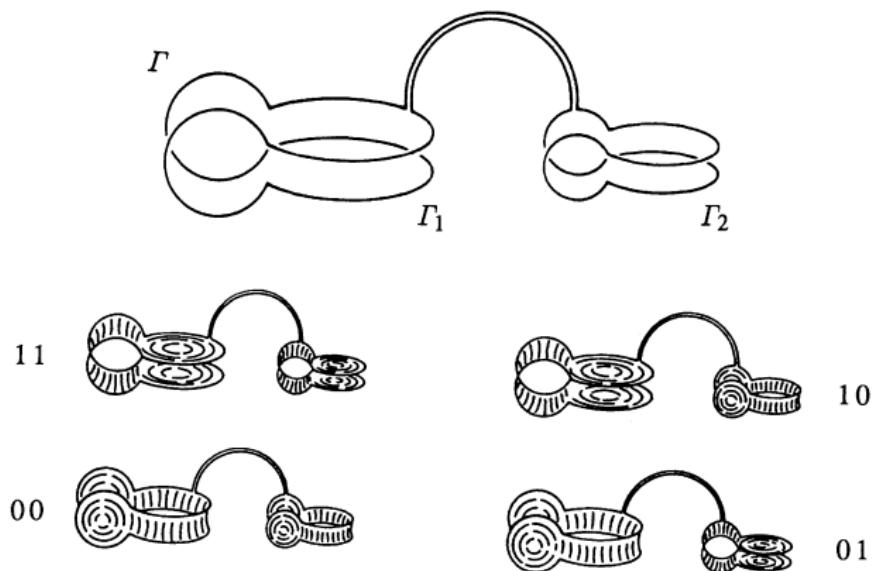
(b)



(c)

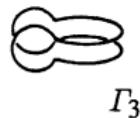
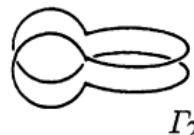
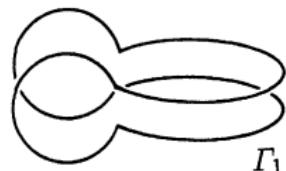
(from Dierkes et al.: Minimal Surfaces, Springer 2010)

Motivation and application: minimal surfaces

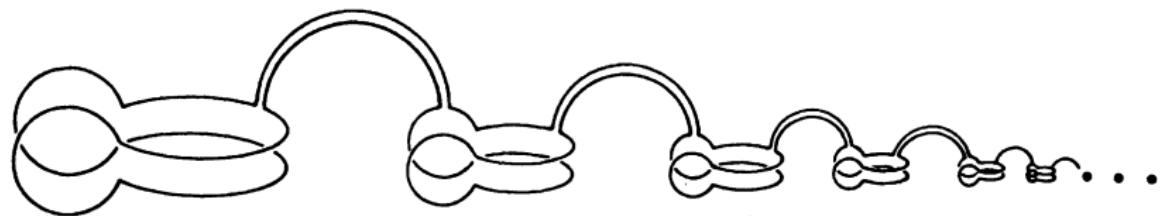


(from Dierkes et al.: Minimal Surfaces, Springer 2010)

Motivation and application: minimal surfaces



...



(from Dierkes et al.: Minimal Surfaces, Springer 2010)

Existence of unstable minimal surfaces

Using persistent homology:

- Number of ϵ -persistent critical points (minimal surfaces) is finite for any $\epsilon > 0$
- Morse inequalities for ϵ -persistent critical points

Theorem (Morse, Tompkins 1939)

There is a C_1 curve bounding an unstable minimal surface (of Morse index 1).

Outlook

Geometry from homology?

