

The logo for Inria, featuring the word "Inria" in a red, cursive script font, set against a white rectangular background.

# A Space and Time Efficient Implementation for Computing Persistent Homology

Clément Maria

# 1

## Introduction

## Background

### Abstract Simplicial Complex.

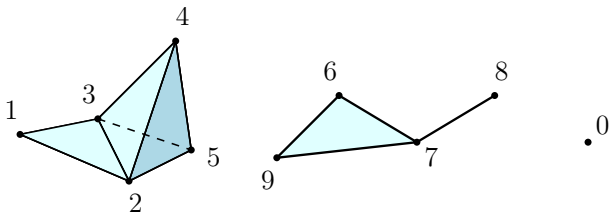
Given a set  $V = \{1 \cdots n\}$  of vertices, an **abstract simplicial complex**  $\mathcal{K}$  on  $V$  is a family of subsets of vertices s.t.:

$$\forall \sigma \in \mathcal{K} : \tau \subseteq \sigma \Rightarrow \tau \in \mathcal{K}$$

Such  $\sigma$  is called a simplex

Dimension of a simplex  $\sigma$ :  $\dim(\sigma) = |\sigma| - 1$

Dimension of the simplicial complex  $\mathcal{K}$ :  $\dim(\mathcal{K}) = \max_{\sigma \in \mathcal{K}} \dim(\sigma)$ .



# Really Short Introduction to Persistent Homology with $\mathbb{Z}_2$ Coefficients

Given a simplicial complex  $\mathcal{K}$ ,  $\mathcal{K}^p$  the  $p$ -dimensional simplices:

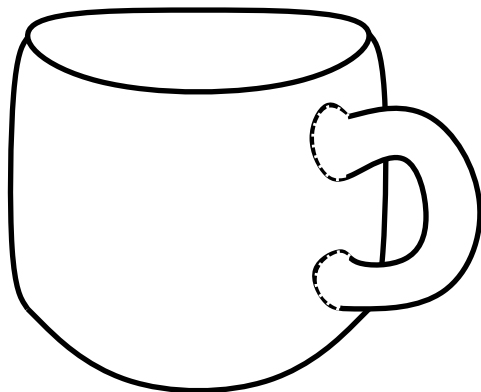
$$C_p = 2^{\mathcal{K}^p}$$

$$\begin{aligned} \partial_p : \quad C_p &\rightarrow C_{p-1} \\ \sigma = \{v_0, \dots, v_p\} &\rightarrow \partial\sigma = \sum_{i=0 \dots p} \{v_0, \dots, \widehat{v}_i, \dots, v_p\} \\ \partial_p \partial_{p+1} &= 0 \end{aligned}$$

$$\begin{cases} \text{Ker}(\partial_p) = Z_p & \partial_p \partial_{p+1} = 0 \Rightarrow B_p \subseteq Z_p \\ \text{Im}(\partial_{p+1}) = B_p & H_p = Z_p / B_p \end{cases}$$

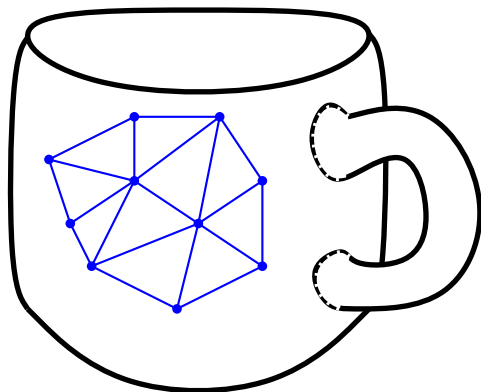
# Really Short Introduction to Persistent Homology with $\mathbb{Z}_2$ Coefficients

$$\begin{cases} \text{Ker}(\partial_p) &= Z_p \\ \text{Im}(\partial_{p+1}) &= B_p \end{cases} \quad \begin{cases} \partial_p \partial_{p+1} = 0 &\Rightarrow B_p \subseteq Z_p \\ H_p = Z_p / B_p \end{cases}$$



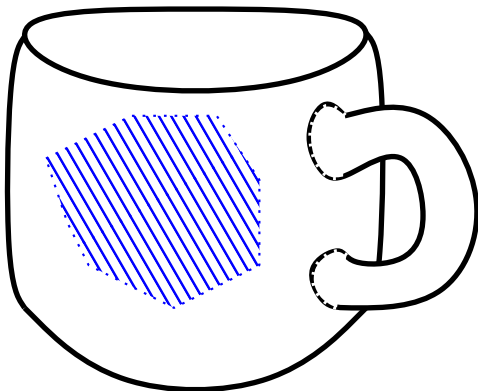
# Really Short Introduction to Persistent Homology with $\mathbb{Z}_2$ Coefficients

$$\begin{cases} \text{Ker}(\partial_p) &= Z_p \\ \text{Im}(\partial_{p+1}) &= B_p \end{cases} \quad \begin{cases} \partial_p \partial_{p+1} = 0 &\Rightarrow B_p \subseteq Z_p \\ H_p = Z_p / B_p \end{cases}$$



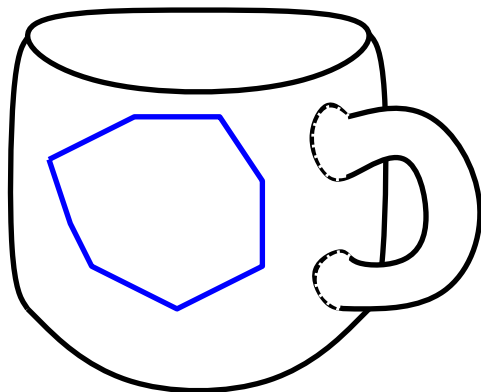
# Really Short Introduction to Persistent Homology with $\mathbb{Z}_2$ Coefficients

$$\begin{cases} \text{Ker}(\partial_p) &= Z_p \\ \text{Im}(\partial_{p+1}) &= B_p \end{cases} \quad \begin{cases} \partial_p \partial_{p+1} = 0 &\Rightarrow B_p \subseteq Z_p \\ H_p = Z_p / B_p \end{cases}$$



# Really Short Introduction to Persistent Homology with $\mathbb{Z}_2$ Coefficients

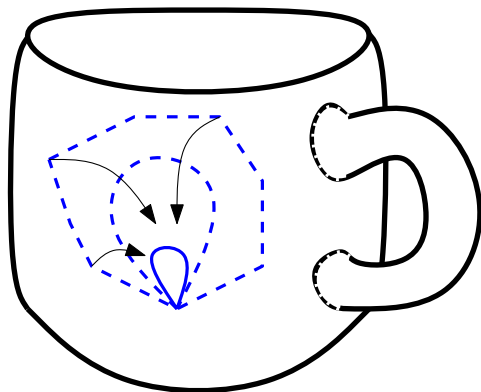
$$\begin{cases} \text{Ker}(\partial_p) &= Z_p \\ \text{Im}(\partial_{p+1}) &= B_p \end{cases} \quad \begin{cases} \partial_p \partial_{p+1} = 0 &\Rightarrow B_p \subseteq Z_p \\ H_p = Z_p / B_p \end{cases}$$





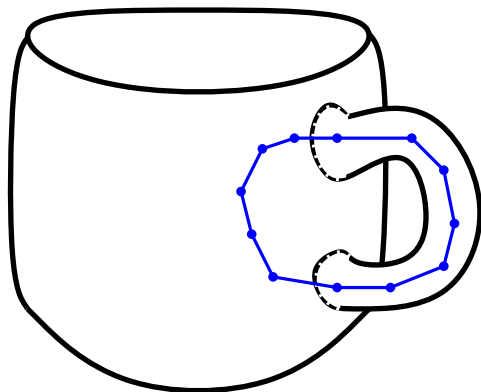
# Really Short Introduction to Persistent Homology with $\mathbb{Z}_2$ Coefficients

$$\begin{cases} \text{Ker}(\partial_p) & = & Z_p & & \partial_p \partial_{p+1} = 0 & \Rightarrow & B_p \subseteq Z_p \\ \text{Im}(\partial_{p+1}) & = & B_p & & H_p = Z_p / B_p & & \end{cases}$$



# Really Short Introduction to Persistent Homology with $\mathbb{Z}_2$ Coefficients

$$\begin{cases} \text{Ker}(\partial_p) &= Z_p \\ \text{Im}(\partial_{p+1}) &= B_p \end{cases} \quad \begin{cases} \partial_p \partial_{p+1} = 0 &\Rightarrow B_p \subseteq Z_p \\ H_p = Z_p / B_p \end{cases}$$

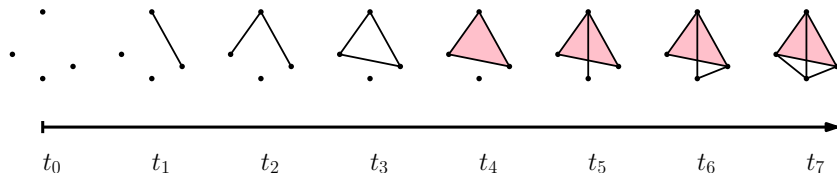


# Filtration of a Simplicial Complex

A *filtration* of a simplicial complex  $\mathcal{K}$  is defined by an indexed family of simplicial complexes  $(\mathcal{K}_{t_i})_{i=0\dots n}$  such that:

$$\emptyset = \mathcal{K}_{t_0} \subset \mathcal{K}_{t_1} \subset \dots \subset \mathcal{K}_{t_{n-1}} \subset \mathcal{K}_{t_n} = \mathcal{K}$$

We can attach to each face of  $\mathcal{K}$  a *time of appearance*  $t_i$ .



**Incremental Algorithm:**  $\sigma = \mathcal{K}_i \setminus \mathcal{K}_{i-1}$ ,  $\dim(\sigma) = p$

**If**  $\sigma$  belongs to a  $p$ -cycle **then**  $\beta_p ++$

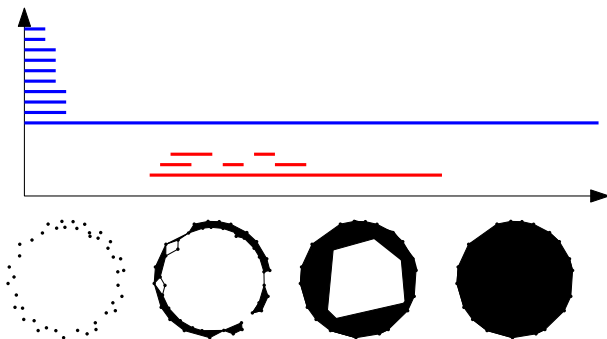
**else**  $\beta_{p-1} --$

# Persistence Diagram

**Incremental Algorithm:**  $\sigma = \mathcal{K}_i \setminus \mathcal{K}_{i-1}$ ,  $\dim(\sigma) = p$

**If**  $\sigma$  belongs to a  $p$ -cycle **then**  $\beta_p ++$

**else**  $\beta_{p-1} --$



# Implementation for Computing Persistent Homology

Focus on simplices:

1. Represent the filtered simplicial complex:
  - ▶ Represent the simplices and their time of appearance
  - ▶ Retrieve incidence relation (compute  $\partial\sigma$ )
2. Attach topological information to simplices
  - ▶ Check if a new simplex is in a cycle or not
  - ▶ Update the topological information under insertion of a simplex
3. Interface the two structures

# 2

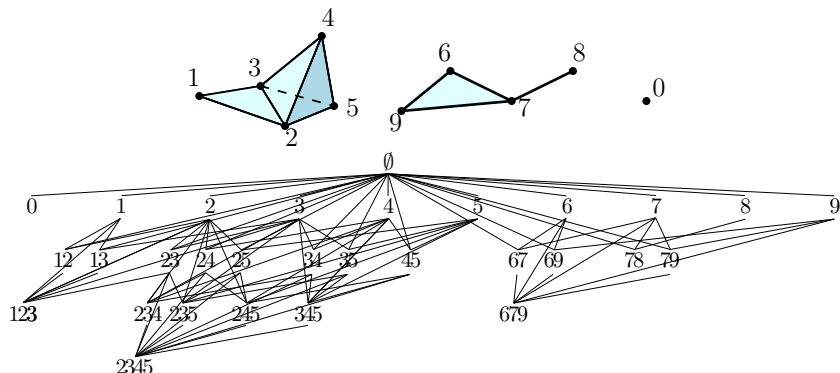
## The Simplex Tree

[Boissonnat, M. '12]

an Efficient Data Structure for  
General Simplicial Complexes

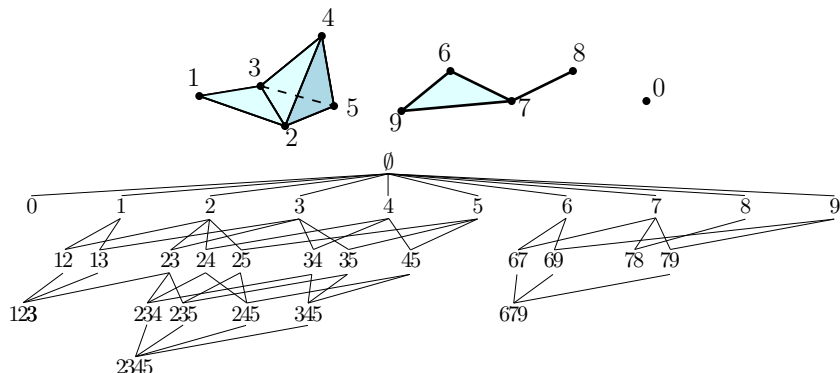
# The Simplex Tree

A simplicial complex is a partially ordered set with regard to inclusion. All the incidences are represented in the following:



# The Simplex Tree

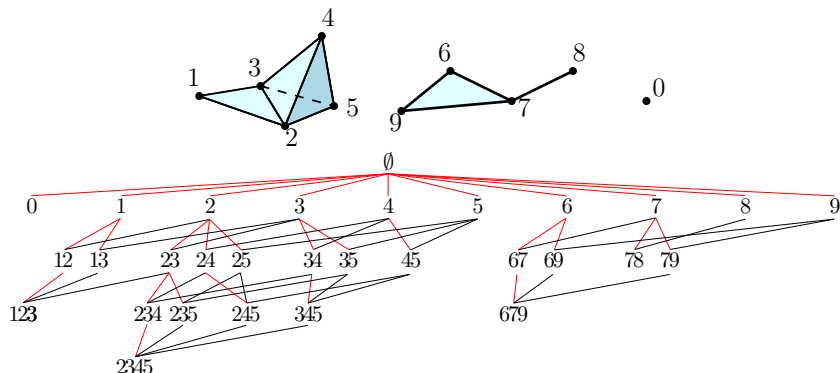
Only incidences from an element to the smallest bigger elements are represented in the Hasse diagram:





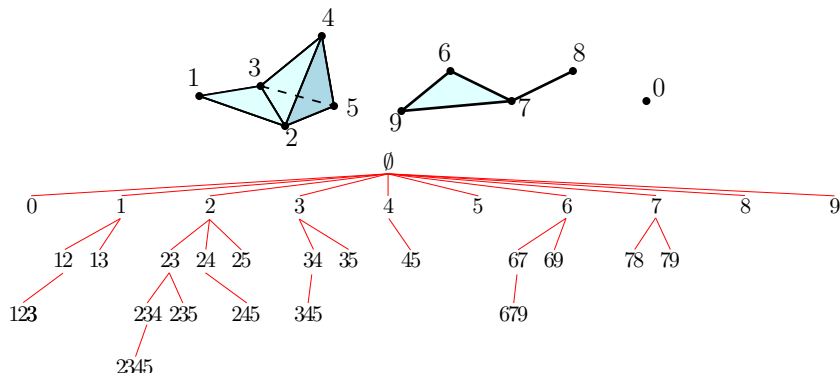
# The Simplex Tree

We select a specific spanning tree of the Hasse diagram. The chosen incidences respect also the lexicographic order:



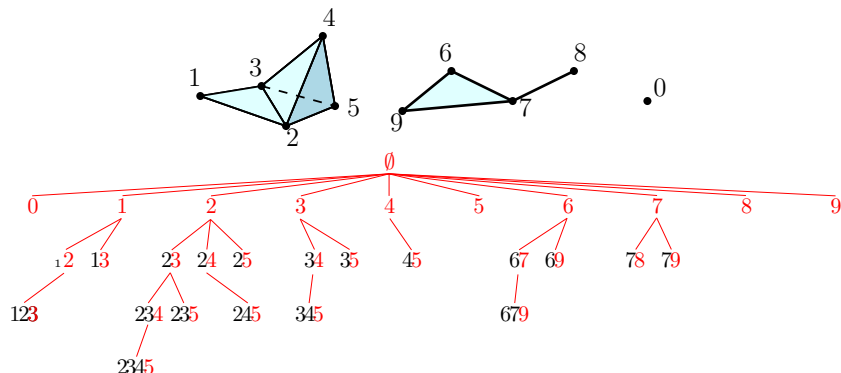
# The Simplex Tree

We select a specific spanning tree of the Hasse diagram. The chosen incidences respect also the lexicographic order:



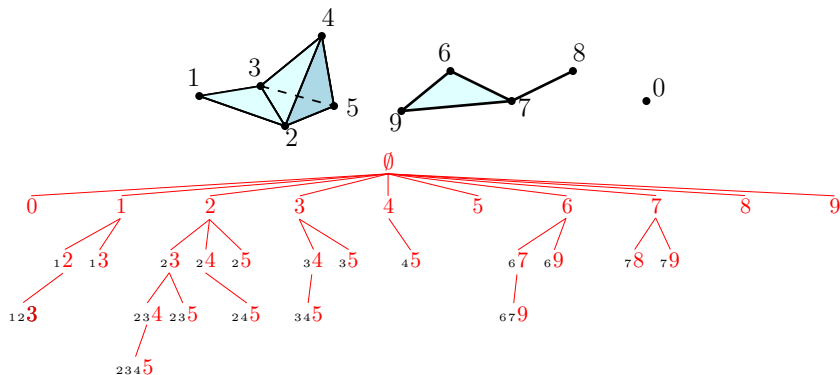
# The Simplex Tree

We keep only the biggest vertex in each simplex. The vertices of a simplex are encountered in the path from the root to its node:



# The Simplex Tree

We keep only the biggest vertex in each simplex. The vertices of a simplex are encountered in the path from the root to its node:

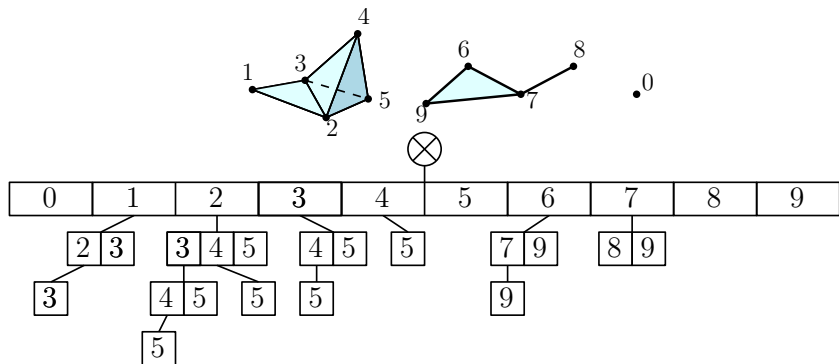


# The Simplex Tree

[Boissonnat, M. '12]

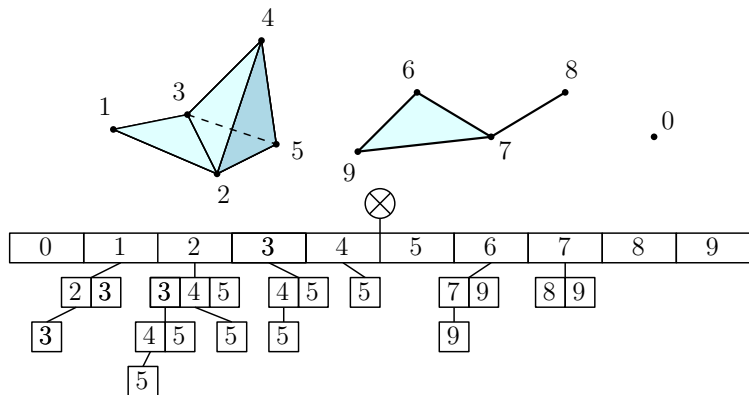
We call this structure a *Simplex Tree*.

**Memory complexity:**  $O(1)$  per simplex.



# The Simplex Tree

EQUIVALENTLY: to each simplex  $\sigma \in \mathcal{K}$  we associate the ordered list of its vertices, and store it in a *trie*.

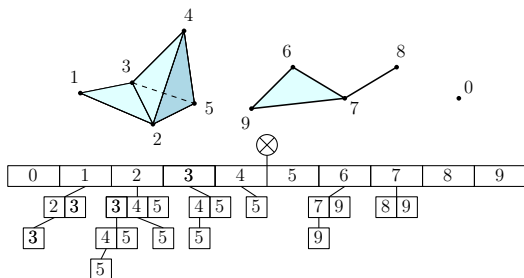


# The Simplex Tree

**Memory complexity:**  $O(1)$  per simplex.

**Some properties:**

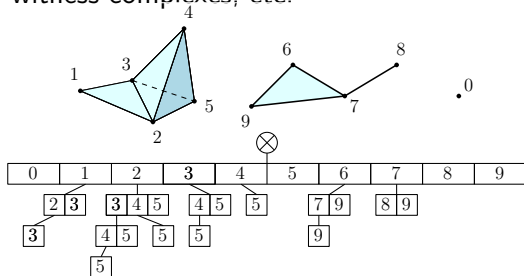
Simplex Tree		Simplicial Complex
#nodes	=	$\#\mathcal{K}$
depth	=	$\dim(\mathcal{K}) + 1$
#children( $\sigma$ )	$\leq$	$\deg(\sigma)$



# Simplex tree: Atomic operations

The Simplex Tree allows to compute efficiently:

- ▶ **Look-up/Insertion/Deletion** of a simplex
- ▶ The **facets** and **subfaces** of a simplex
- ▶ The **cofaces** of a simplex
- ▶ **Edge contractions**
- ▶ **Elementary collapses**
- ▶ **Fast construction** of Flag complexes, witness complexes, relaxed witness complexes, etc.

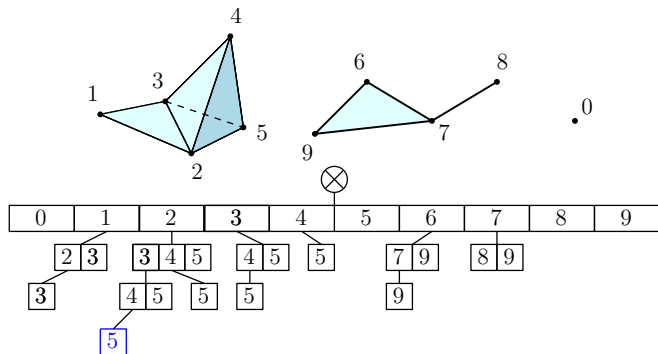




# Facets/Boundary Computation

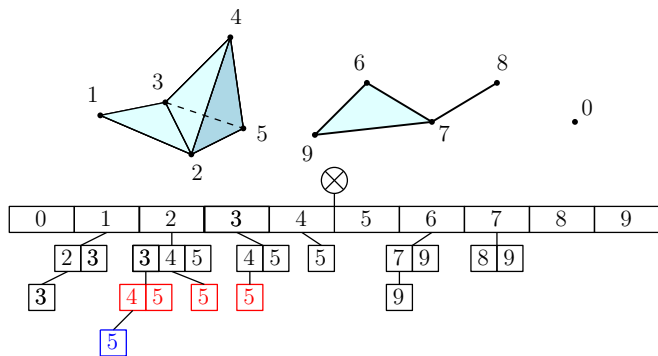
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



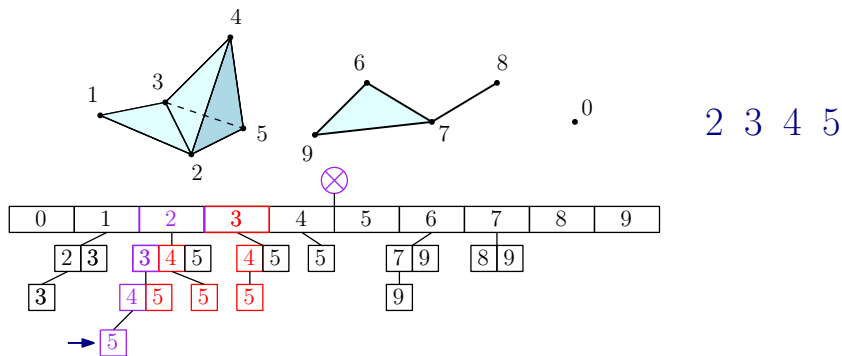
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



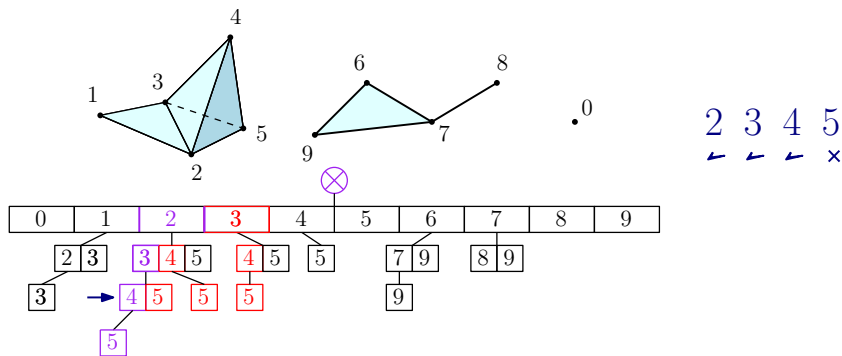
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



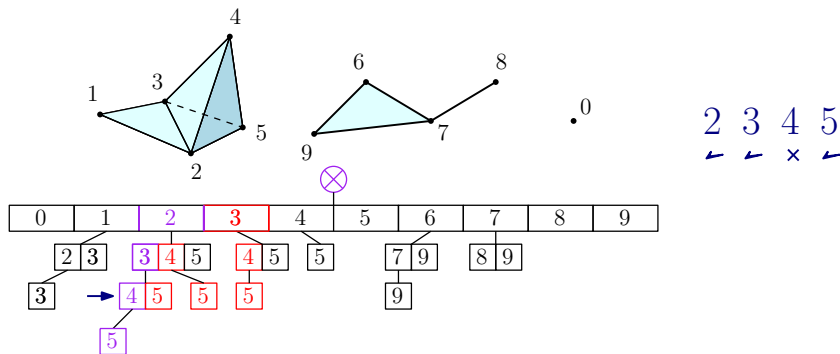
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



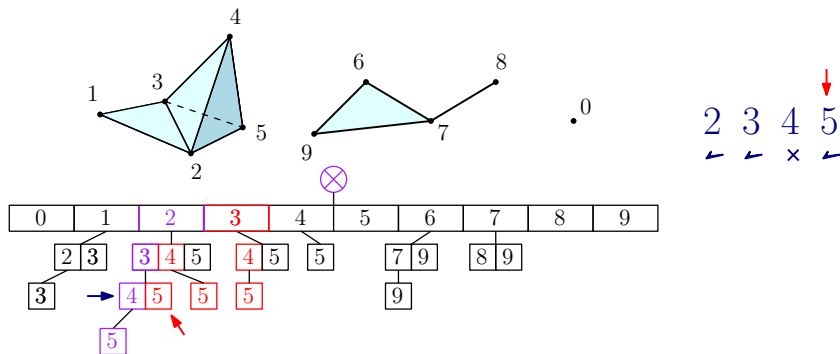
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



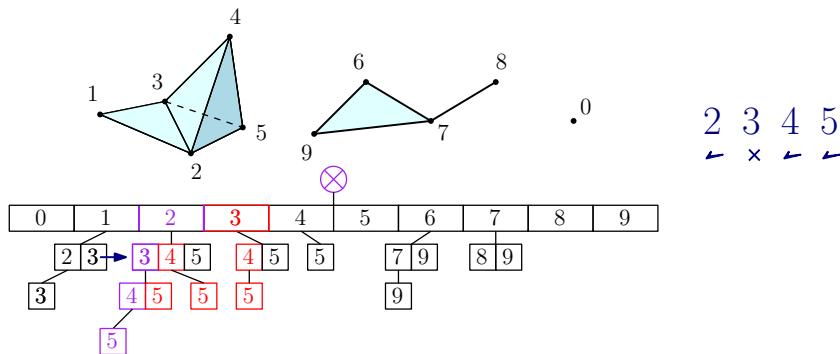
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



# Facets/Boundary Computation

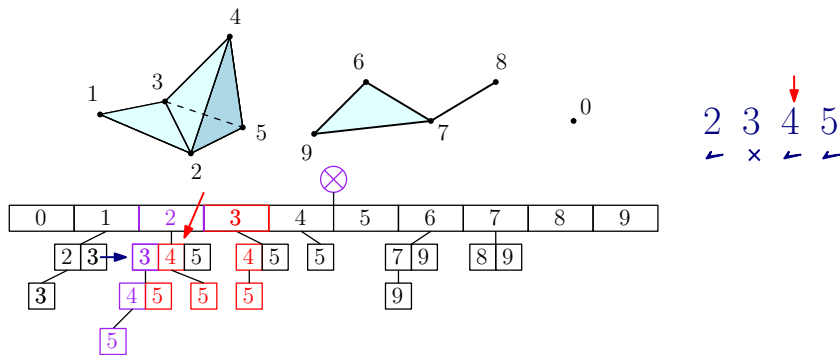
Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.





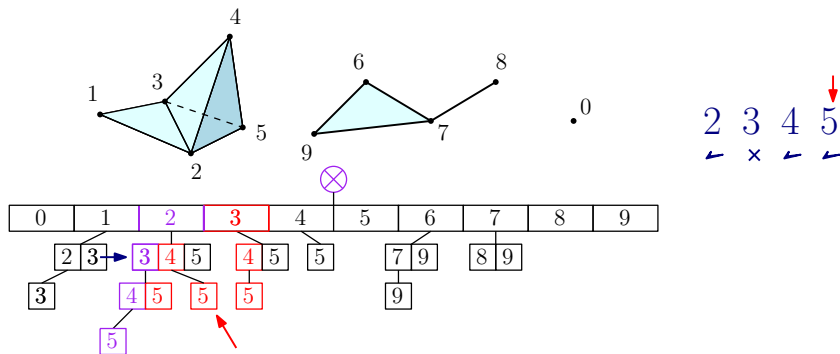
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



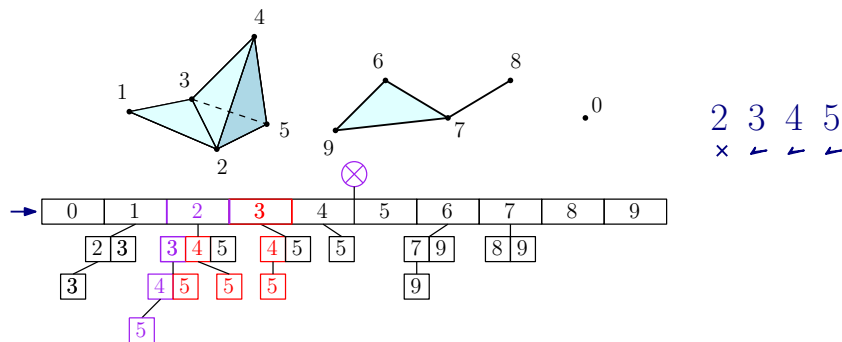
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



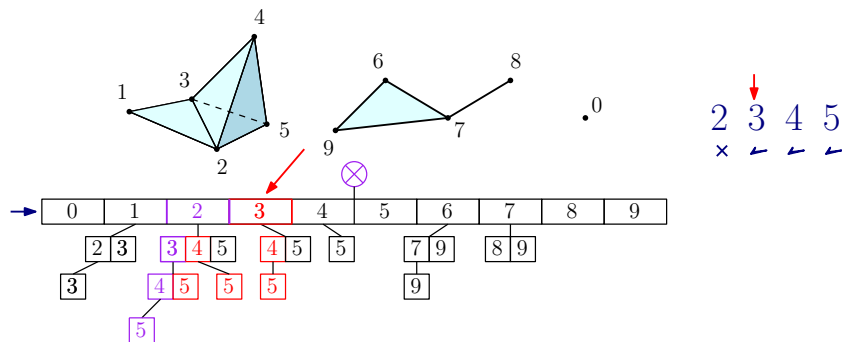
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



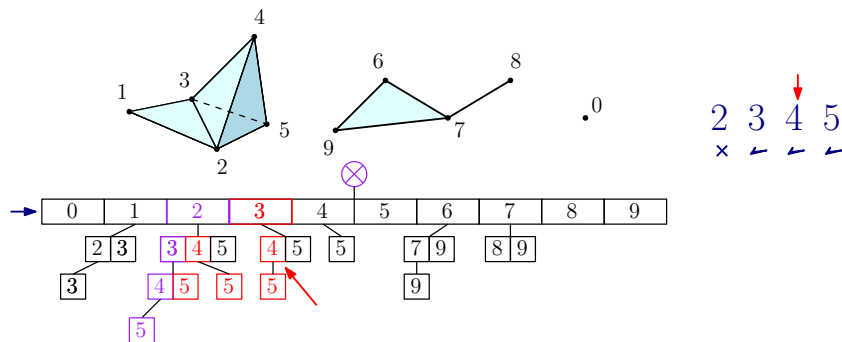
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



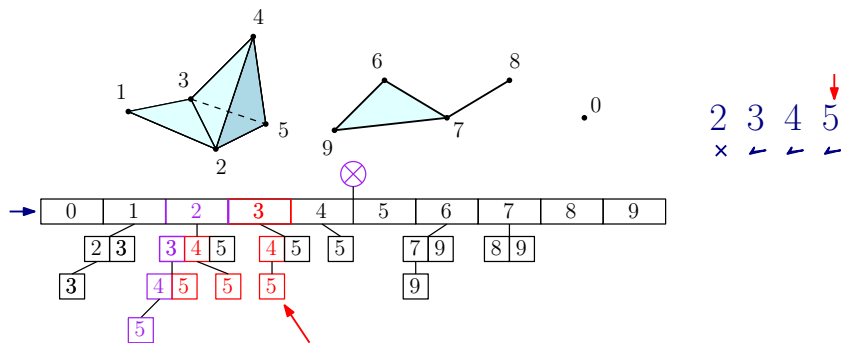
# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



# Facets/Boundary Computation

Given a simplex  $\sigma$  we find its  $|\sigma|$  codimension 1 subfaces.



# 3

## Persistent Cohomology and Annotations

Given a simplicial complex  $\mathcal{K}$ , let  $\mathcal{K}^p$  denote the set of  $p$ -simplices in  $\mathcal{K}$ .

### Definition

An *annotation* for  $\mathcal{K}^p$  is an assignment  $a^p : \mathcal{K}^p \rightarrow \mathbb{Z}_2^g$  of a binary vector  $a_\sigma = a^p(\sigma)$  of length  $g$ . Linear extension to  $C_p$ .

### Definition

An annotation  $a : \mathcal{K}^p \rightarrow \mathbb{Z}_2^g$  is *valid* if:

1.  $g = \text{rank } H_p(K)$ , and
2. two  $p$ -cycles  $z_1$  and  $z_2$  have  $a_{z_1} = a_{z_2}$  iff their homology classes  $[z_1]$  and  $[z_2]$  are identical.



## Theorem

The following two statements are equivalent

1. An annotation  $a : \mathcal{K}^p \rightarrow \mathbb{Z}_2^g$  is valid
2. The cochains  $\{\phi_j\}_{j=1\dots g}$  given by  $\phi_j(\sigma) = a_\sigma[j]$  for all  $\sigma \in \mathcal{K}^p$  are cocycles whose cohomology classes  $\{[\phi_j]\}_{j=1\dots g}$  constitute a basis of  $H^p(\mathcal{K})$ .

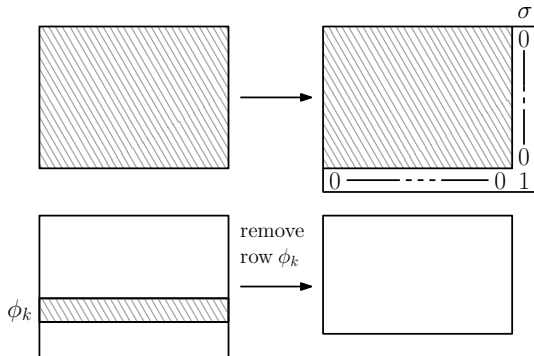
	$\sigma_1$	$\dots$	$\sigma_{ \mathcal{K}^p }$
$\phi_1$			
$\vdots$		0/1	
$\phi_g$			

# Persistent coHomology Algorithm

[de Silva, Morozov, Vejdemo-Johansson '11]; [Dey,Fan,Wang]

Insert  $\sigma = \mathcal{K}_{i+1} \setminus \mathcal{K}_i$ ; compute  $a_{\partial\sigma}$

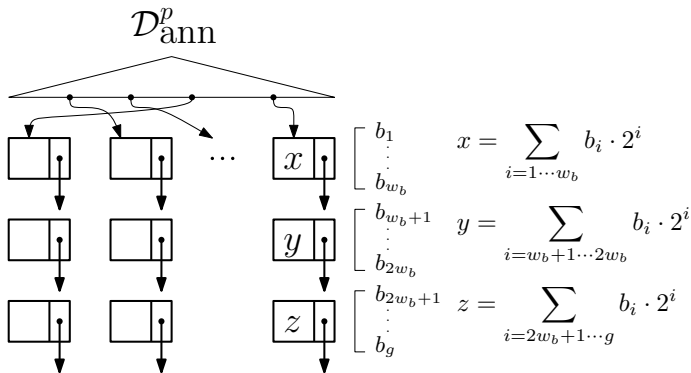
- ▶ if  $a_{\partial\sigma} = 0$ , create a class
- ▶ else kills a class  $\phi_k$  by adding  $a_{\partial\sigma}$  to all columns with a 1 at index  $k$



# Representation of the Annotation

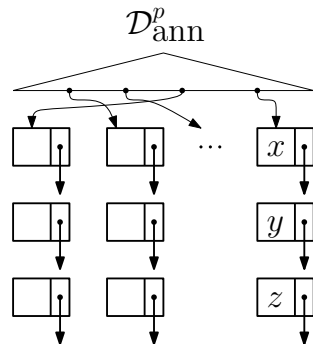
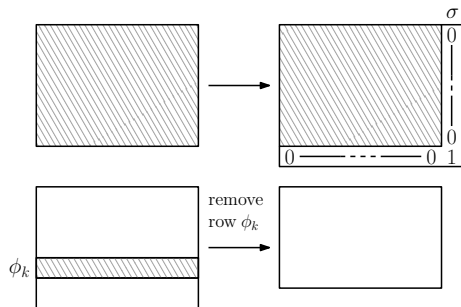
[Boissonnat, Dey, M.]

- ▶ No duplicate vector
- ▶ Bits are “packed” into integers



# Representation of the Annotation

[Boissonnat, Dey, M.]



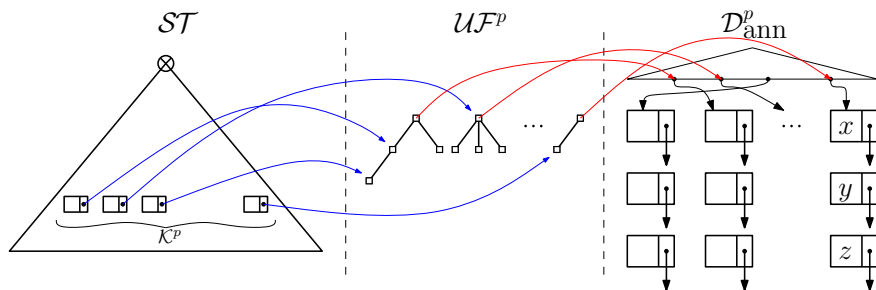
# 4

## The Implementation

[Boissonnat, Dey, M.]

# The Data Structure

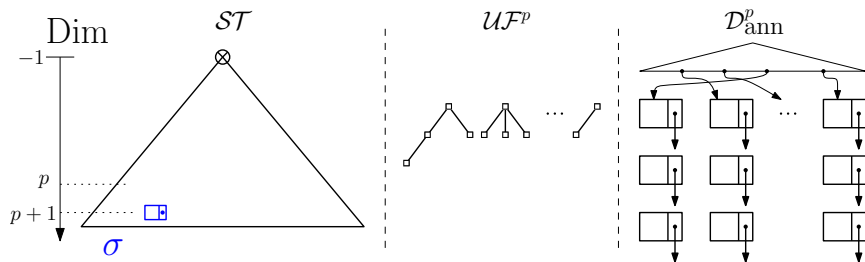
We connect simplicial complex and annotations with a union-find:



# Implementation of the Algorithm

Insert  $\sigma = \mathcal{K}_{i+1} \setminus \mathcal{K}_i$ ; compute  $a_{\partial\sigma}$

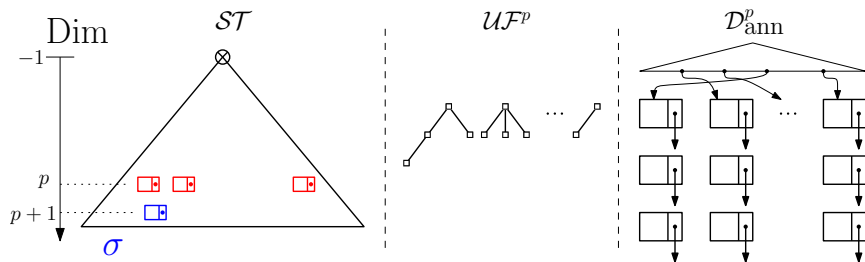
- ▶ if  $a_{\partial\sigma} = 0$ , create a class
- ▶ else kills a class  $\phi_k$  by adding  $a_{\partial\sigma}$  to all columns with a 1 at index  $k$



# Implementation of the Algorithm

Insert  $\sigma = \mathcal{K}_{i+1} \setminus \mathcal{K}_i$ ; compute  $a_{\partial\sigma}$

- ▶ if  $a_{\partial\sigma} = 0$ , create a class
- ▶ else kills a class  $\phi_k$  by adding  $a_{\partial\sigma}$  to all columns with a 1 at index  $k$

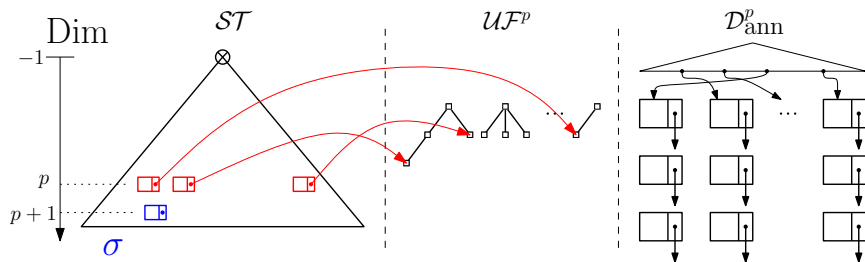




# Implementation of the Algorithm

Insert  $\sigma = \mathcal{K}_{i+1} \setminus \mathcal{K}_i$ ; compute  $a_{\partial\sigma}$

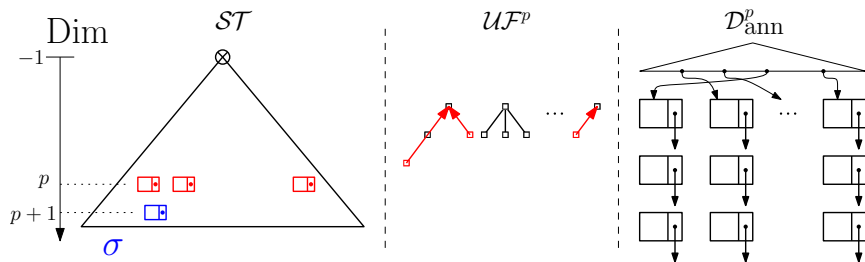
- ▶ if  $a_{\partial\sigma} = 0$ , create a class
- ▶ else kills a class  $\phi_k$  by adding  $a_{\partial\sigma}$  to all columns with a 1 at index  $k$



# Implementation of the Algorithm

Insert  $\sigma = \mathcal{K}_{i+1} \setminus \mathcal{K}_i$ ; compute  $a_{\partial\sigma}$

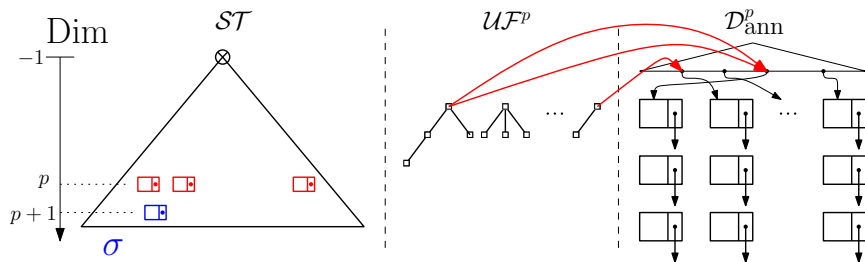
- ▶ if  $a_{\partial\sigma} = 0$ , create a class
- ▶ else kills a class  $\phi_k$  by adding  $a_{\partial\sigma}$  to all columns with a 1 at index  $k$



# Implementation of the Algorithm

Insert  $\sigma = \mathcal{K}_{i+1} \setminus \mathcal{K}_i$ ; compute  $a_{\partial\sigma}$

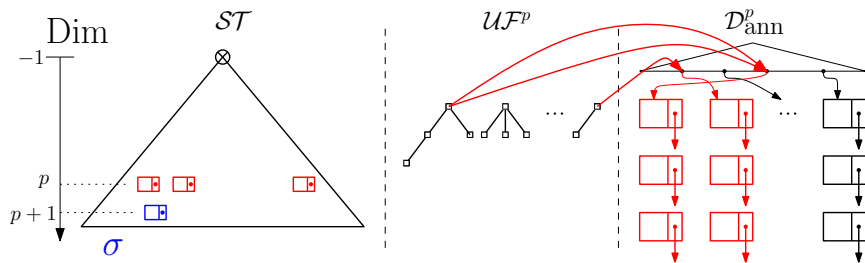
- ▶ if  $a_{\partial\sigma} = 0$ , create a class
- ▶ else kills a class  $\phi_k$  by adding  $a_{\partial\sigma}$  to all columns with a 1 at index  $k$



# Implementation of the Algorithm

Insert  $\sigma = \mathcal{K}_{i+1} \setminus \mathcal{K}_i$ ; compute  $a_{\partial\sigma}$

- ▶ if  $a_{\partial\sigma} = 0$ , create a class
- ▶ else kills a class  $\phi_k$  by adding  $a_{\partial\sigma}$  to all columns with a 1 at index  $k$



# Complexity

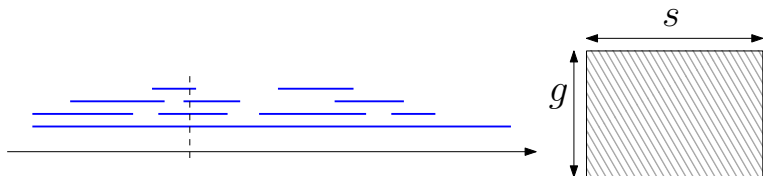
Space:

$$O\left(|\mathcal{K}| + k \frac{g_m s_m}{w_b}\right)$$

Time:

$$O\left(|\mathcal{K}| \left[ k \left( \alpha(|\mathcal{K}|) + \left\lceil \frac{g_m}{w_b} \right\rceil + k D_m \right) + s_m \left( \left\lceil \frac{g_m}{w_b} \right\rceil \log(s_m) + \alpha(|\mathcal{K}|) \right) \right] \right)$$

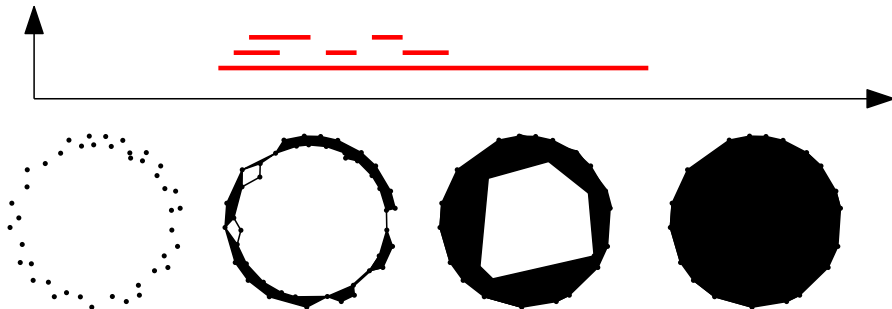
intuitively...



# Lazy Evaluation of Simplices

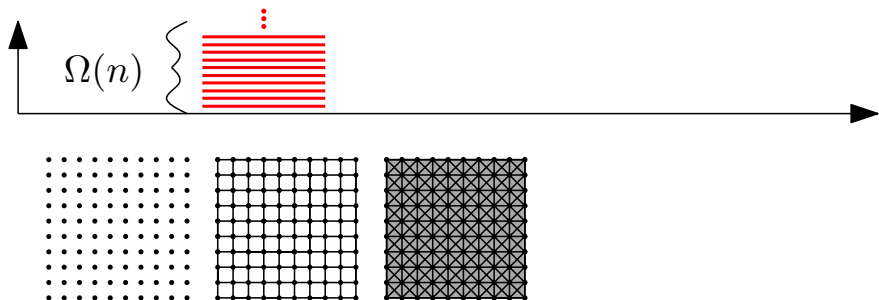
# Lazy Evaluation of Simplices

We minimize the number of “alive bars” living in parallel.



# Lazy Evaluation of Simplices

We minimize the number of “alive bars” living in parallel.





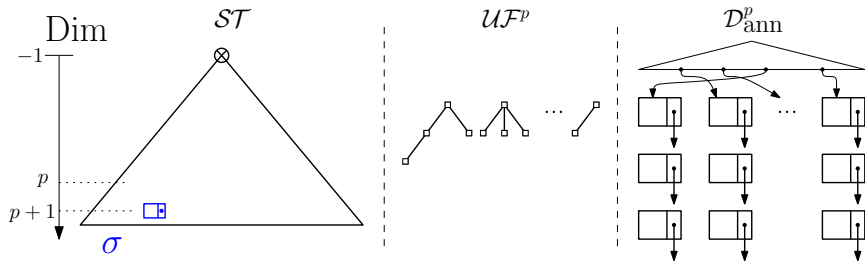
## Lazy Evaluation of Simplices

$$\emptyset = \mathcal{K}_{\rho_0} \subset \mathcal{K}_{\rho_1} \subset \dots \subset \mathcal{K}_{\rho_{m-1}} \subset \mathcal{K}_{\rho_m} = \mathcal{K}$$

$$\sigma_0 \quad \sigma_1 \quad \sigma_2 \quad \dots \quad \sigma_{m-2} \quad \sigma_{m-1} \quad \sigma_m$$

- ▶ we insert killers as soon as we encounter them
- ▶ we postpone the insertion of a creator  $\sigma$  until we encounter one of its cofaces

We mark the simplices whose insertion has been postponed. When considering  $\sigma$  in the filtration:



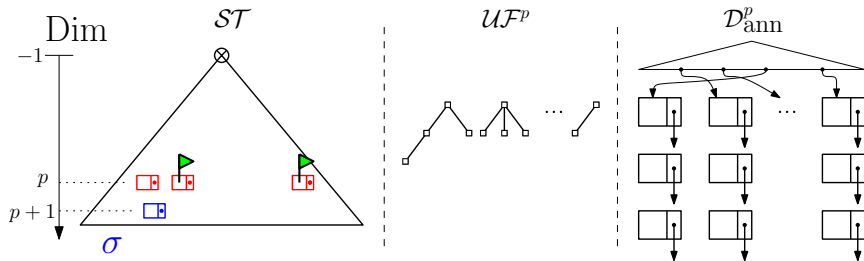
# Lazy Evaluation of Simplices

$$\emptyset = \mathcal{K}_{\rho_0} \subset \mathcal{K}_{\rho_1} \subset \dots \subset \mathcal{K}_{\rho_{m-1}} \subset \mathcal{K}_{\rho_m} = \mathcal{K}$$

$$\sigma_0 \quad \sigma_1 \quad \sigma_2 \quad \dots \quad \sigma_{m-2} \quad \sigma_{m-1} \quad \sigma_m$$

- ▶ we insert killers as soon as we encounter them
- ▶ we postpone the insertion of a creator  $\sigma$  until we encounter one of its cofaces

We mark the simplices whose insertion has been postponed. When considering  $\sigma$  in the filtration:



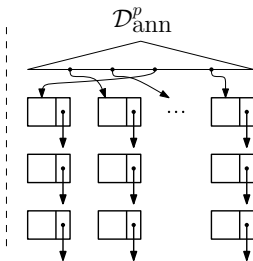
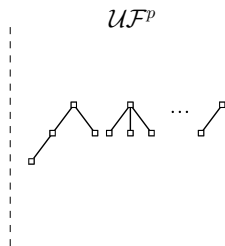
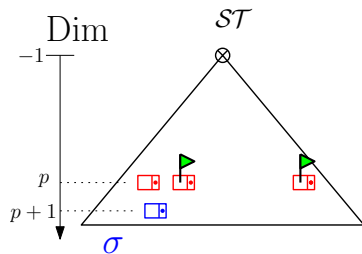
# Lazy Evaluation of Simplices

$$\emptyset = \mathcal{K}_{\rho_0} \subset \mathcal{K}_{\rho_1} \subset \dots \subset \mathcal{K}_{\rho_{m-1}} \subset \mathcal{K}_{\rho_m} = \mathcal{K}$$

$$\sigma_0 \quad \sigma_1 \quad \sigma_2 \quad \dots \quad \sigma_{m-2} \quad \sigma_{m-1} \quad \sigma_m$$

- ▶ we insert killers as soon as we encounter them
- ▶ we postpone the insertion of a creator  $\sigma$  until we encounter one of its cofaces

**No additional complexity!**  
**Same persistence diagram!**



# 5

## Experiments

## Performance

Data	$ \mathcal{P} $	$D$	$d$	$r$	$k$	$ \mathcal{K} $	$T_{\text{ann}}$	$T_{\text{ann}}/ \mathcal{K} (\text{s.})$
<b>S4</b>	507	5	4	0.75	5	$191 \cdot 10^6$	$\sim 32\text{min}$	$1.0 \cdot 10^{-5}$
<b>Bud</b>	1000	3	2	2.4	3	$283 \cdot 10^6$	$\sim 10\text{min}$	$2.2 \cdot 10^{-6}$
<b>Bro</b>	500	25	?	0.8	16	$38 \cdot 10^6$	$\sim 181\text{min}$	$2.8 \cdot 10^{-4}$
<b>Cy8</b>	500	24	2	1.3	17	$18 \cdot 10^6$	$\sim 9\text{min}$	$3.1 \cdot 10^{-5}$
<b>KI</b>	4900	5	2	0.415	5	$218 \cdot 10^6$	$\sim 29\text{min}$	$7.9 \cdot 10^{-6}$

Figure: Data, timings and statistics

$$O\left(|\mathcal{K}| + k \frac{g_m s_m}{w_b}\right)$$

$$O\left(|\mathcal{K}| \left[ k \left( \alpha(|\mathcal{K}|) + \left\lceil \frac{g_m}{w_b} \right\rceil + k D_m \right) + s_m \left( \left\lceil \frac{g_m}{w_b} \right\rceil \log(s_m) + \alpha(|\mathcal{K}|) \right) \right] \right)$$

# Homology vs coHomology

Comparison with JP<sub>LEX</sub> and DIONYSUS: Persistent Homology vs Persistent coHomology

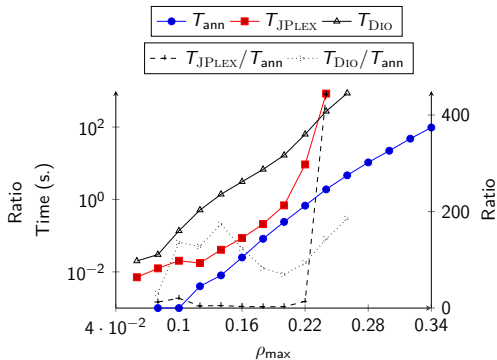
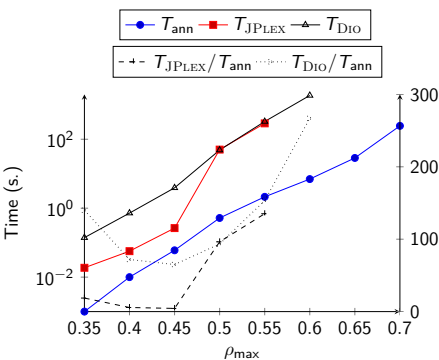


Figure: S4 and KI

# coHomology vs coHomology

Comparison with DIONYSUS: coHomology vs coHomology

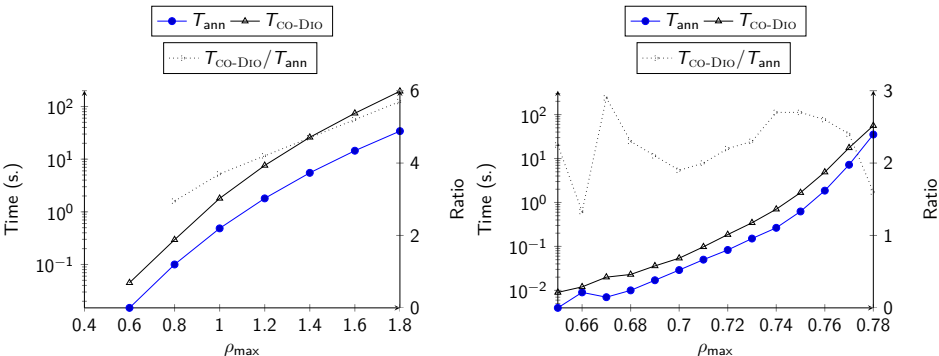


Figure: Bud and Bro

# Memory

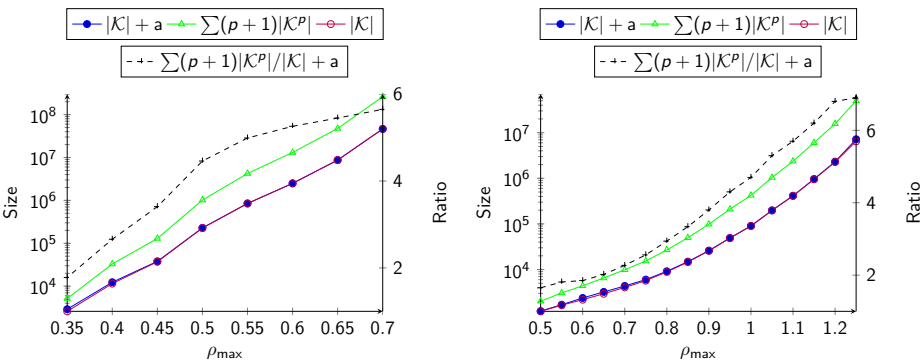


Figure: S4 and Cy8



# Numbers!

Dim:	<b>S4:</b> 1	2	3	4	<b>Bud:</b> 1	2	<b>KI:</b> 1	2	3	4						
$G_m/g_m$	4.9	1.2	6.2	6.1	2.1	8.4	11.8	10.3	10.3	8.9						
$ \mathcal{K}^P /s_m$	$10^2$	$10^3$	$10^4$	$10^3$	$10^3$	$10^5$	$10^3$	$10^4$	$10^4$	$10^4$						
$\lceil g_m/w_b \rceil$	5	4	10	86	5	2	4	1	11	87						
Dim:	<b>Bro:</b> 1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	
$G_m/g_m$	2.1	1.8	5.4	5.7	5.2	5.6	4.0	4.2	3.0	3.7	3.7	2.9	3.0	3.7	3.0	
$ \mathcal{K}^P /s_m$	$10^1$	$10^3$	$10^3$	$10^3$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	
$\lceil g_m/w_b \rceil$	6	6	6	36	150	330	703	694	726	343	133	46	8	228	342	
Dim:	<b>Cy8:</b> 1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
$G_m/g_m$	2.3	1.7	3.9	3.8	4.1	4.1	3.5	3.5	3.9	2.7	2.6	3.0	2.8	1.9	2.7	1
$ \mathcal{K}^P /s_m$	$10^3$	$10^3$	$10^3$	$10^3$	$10^3$	$10^3$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$	$10^2$
$\lceil g_m/w_b \rceil$	3	1	3	12	33	74	147	186	160	168	93	33	10	3	1	1

Figure: Statistics about the evolution of the dimension of the homology groups we maintain during the computation.

# Conclusion

## Future Work

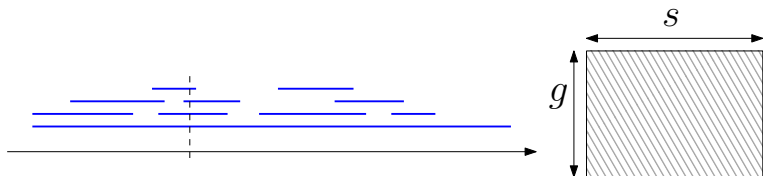
Space:

$$O\left(|\mathcal{K}| + k \frac{g_m s_m}{w_b}\right)$$

Time:

$$O\left(|\mathcal{K}| \left[ k \left( \alpha(|\mathcal{K}|) + \left\lceil \frac{g_m}{w_b} \right\rceil + k D_m \right) + s_m \left( \left\lceil \frac{g_m}{w_b} \right\rceil \log(s_m) + \alpha(|\mathcal{K}|) \right) \right] \right)$$

intuitively...



Thank you!

Question?