

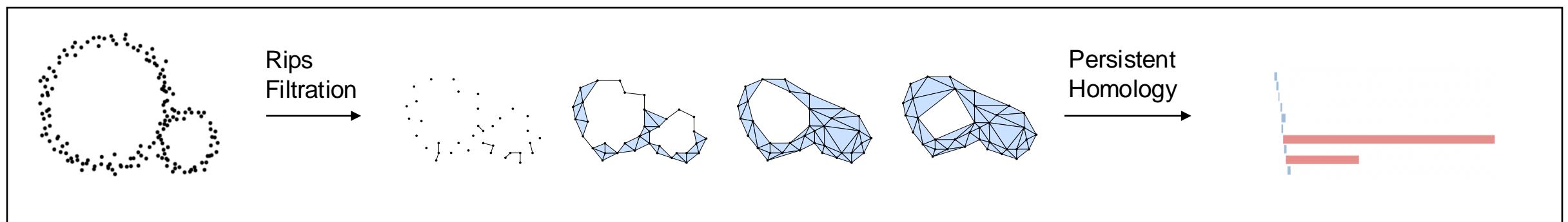
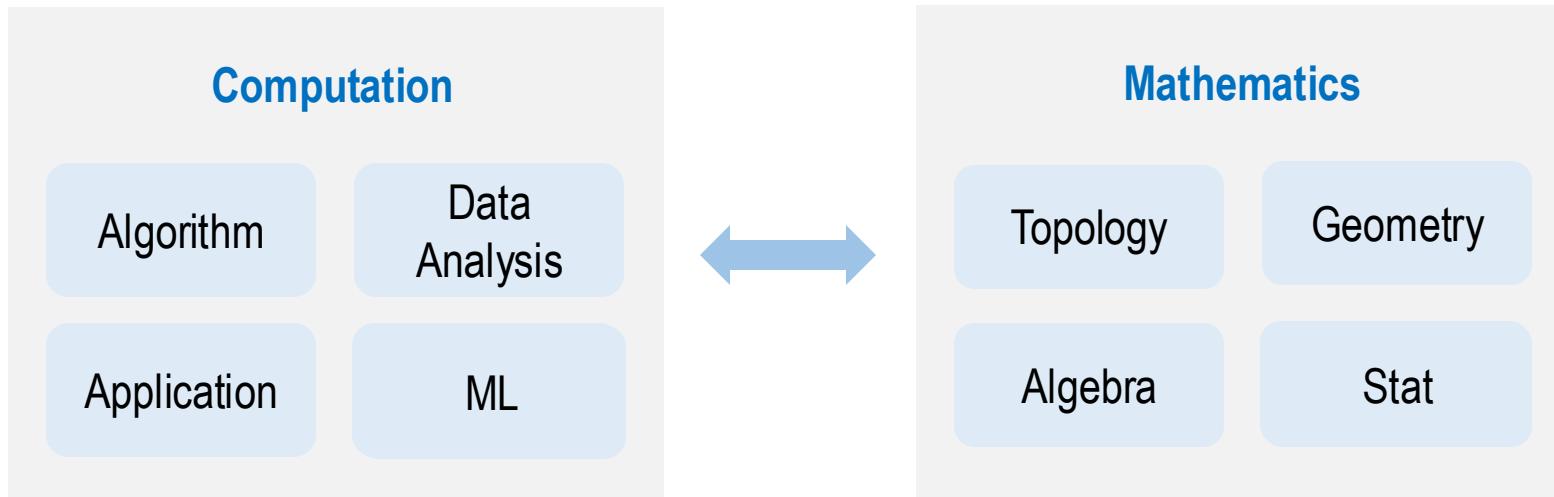
# Can zigzag persistence be computed as efficiently as the standard version?

*Geometry and Topology Seminar, Oregon State University*

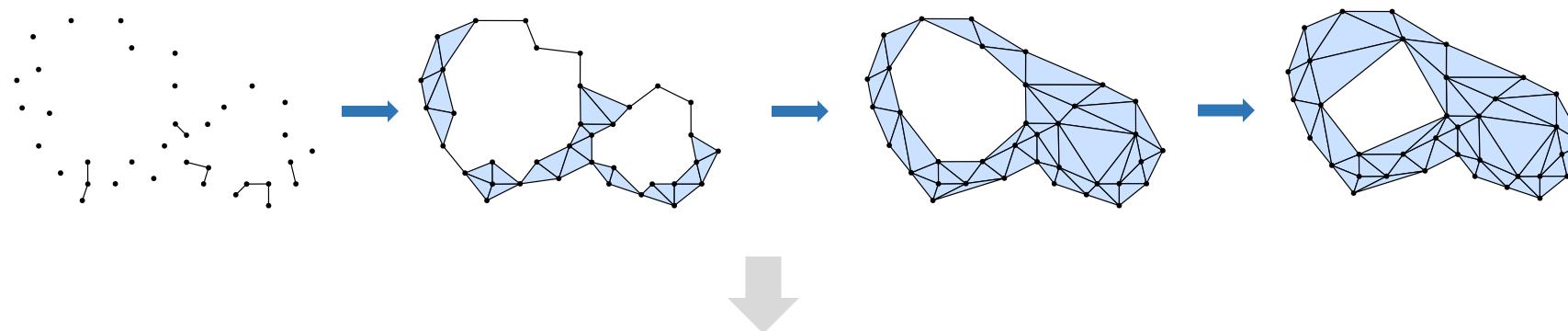
**Tao Hou**, CS Department  
University of Oregon

Joint work with **Tamal K. Dey, Dmitriy Morozov, Salman Parsa**

# Topological data analysis (TDA)

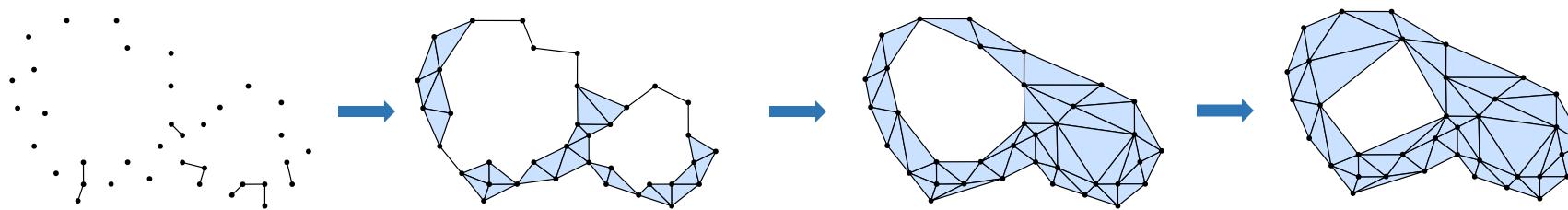


# Persistent homology

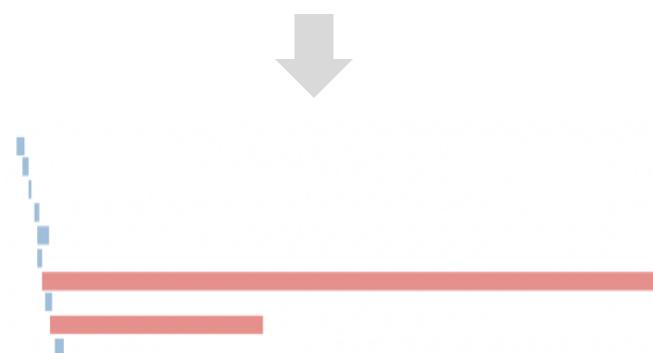


- As we add each simplex in the sequence, the homology of the complex changes, with:
  - Birth: betti number increased by 1
  - Death: betti number decreased by 1

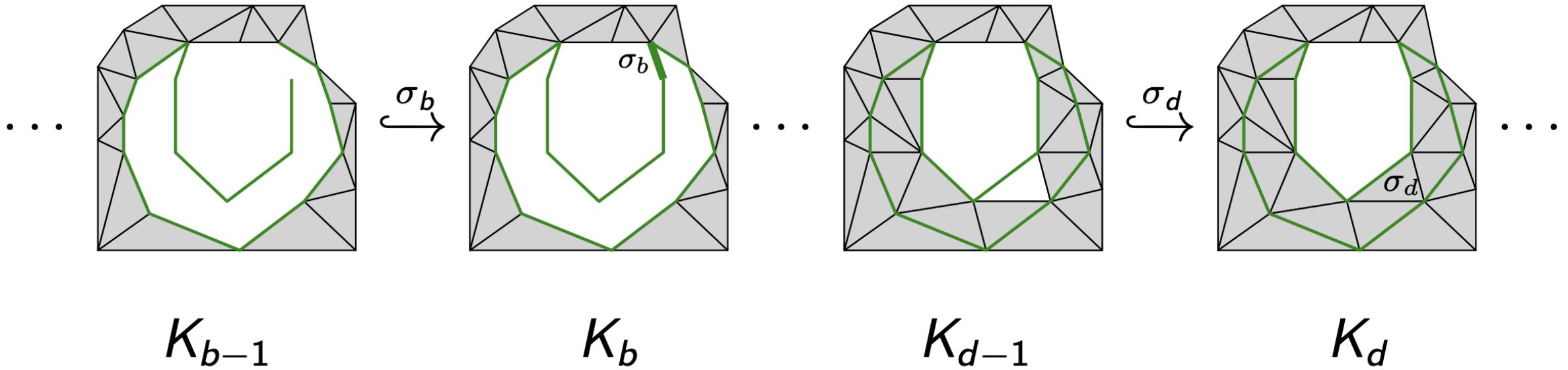
# Persistent homology



- As we add each simplex in the sequence, the homology of the complex changes, with:
  - Birth: betti number increased by 1
  - Death: betti number decreased by 1
- The birth and death points can be canonically paired, resulting in [persistence barcode](#):

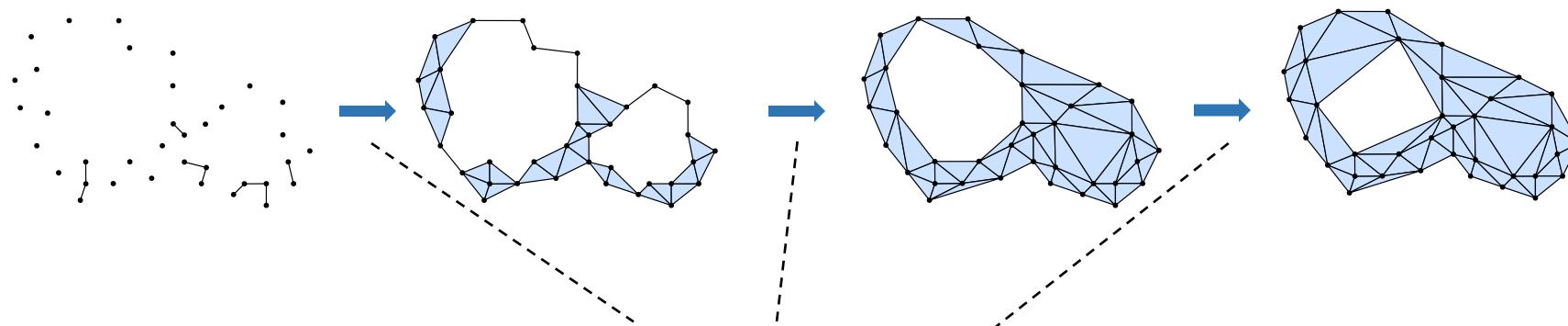


# Persistent homology: example



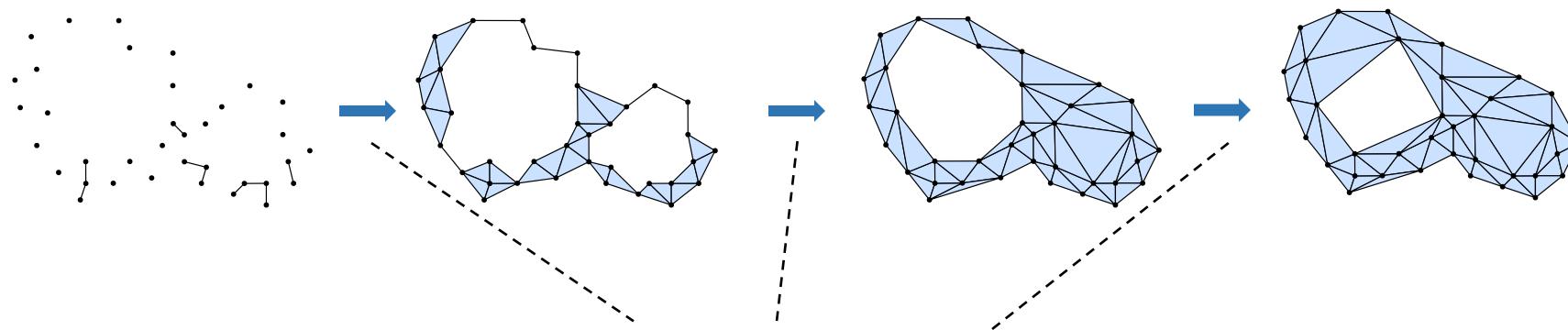
An interval:  $[b, d) = [b, d - 1]$

# Persistent homology: Simplex-wise filtration



Expand each arrow into a sequence of additions of a single simplex

# Persistent homology: Simplex-wise filtration



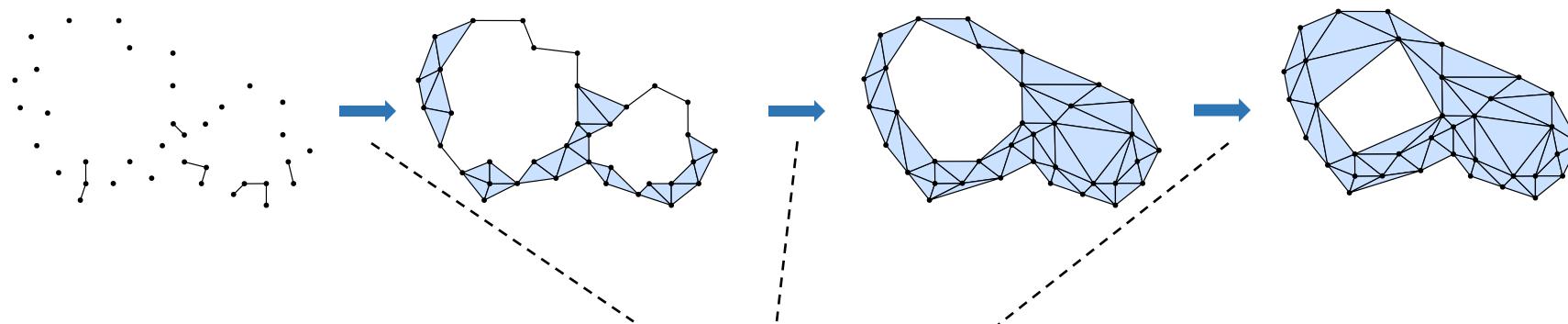
Expand each arrow into a sequence of additions of a single simplex



Simplex-wise filtration: a sequence of additions of a single simplex

$$\mathcal{F} : \emptyset = K_0 \xhookrightarrow{\sigma_0} K_1 \xhookrightarrow{\sigma_1} \dots \xhookrightarrow{\sigma_{m-1}} K_{m-1} \xhookrightarrow{\sigma_m} K_m$$

# Persistent homology: Simplex-wise filtration



Expand each arrow into a sequence of additions of a single simplex



Simplex-wise filtration: a sequence of additions of a single simplex

$$\mathcal{F} : \emptyset = K_0 \xhookrightarrow{\sigma_0} K_1 \xhookrightarrow{\sigma_1} \dots \xhookrightarrow{\sigma_{m-1}} K_{m-1} \xhookrightarrow{\sigma_m} K_m$$



# Persistent homology: Pipeline

Standard filtration:

$$\mathcal{F} : K_0 \xrightarrow{\sigma_0} K_1 \xrightarrow{\sigma_1} \cdots \xrightarrow{\sigma_{m-2}} K_{m-1} \xrightarrow{\sigma_{m-1}} K_m$$



Induced module:

$$H_p(\mathcal{F}) : H_p(K_0) \rightarrow H_p(K_1) \rightarrow \cdots \rightarrow H_p(K_{m-1}) \rightarrow H_p(K_m)$$



Interval decomposition: [Gabriel 72]

$$H_p(\mathcal{F}) = \bigoplus_{\alpha \in \mathcal{A}} \mathcal{I}^{[b_\alpha, d_\alpha]}$$



$p$ -th persistence barcode:

$$\text{Pers}_p(\mathcal{F}) = \{[b_\alpha, d_\alpha] \mid \alpha \in \mathcal{A}\}$$

# Persistent homology: Pipeline

Standard filtration:

$$\mathcal{F} : K_0 \xrightarrow{\sigma_0} K_1 \xrightarrow{\sigma_1} \cdots \xrightarrow{\sigma_{m-2}} K_{m-1} \xrightarrow{\sigma_{m-1}} K_m$$



Induced module:

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assumes  $\mathbb{Z}_2$  as coefficients

Interval decomposition: [Gabriel 72]

$$H_p(\mathcal{F}) = \bigoplus_{\alpha \in \mathcal{A}} \mathcal{I}^{[b_\alpha, d_\alpha]}$$

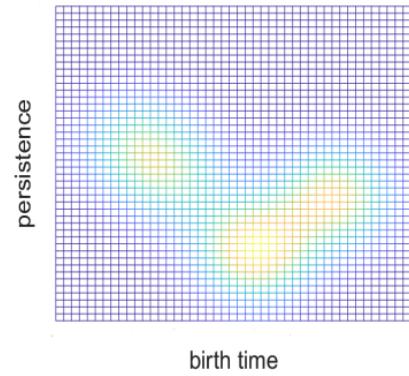
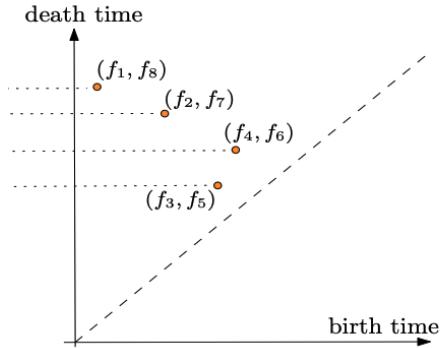


$p$ -th persistence barcode:

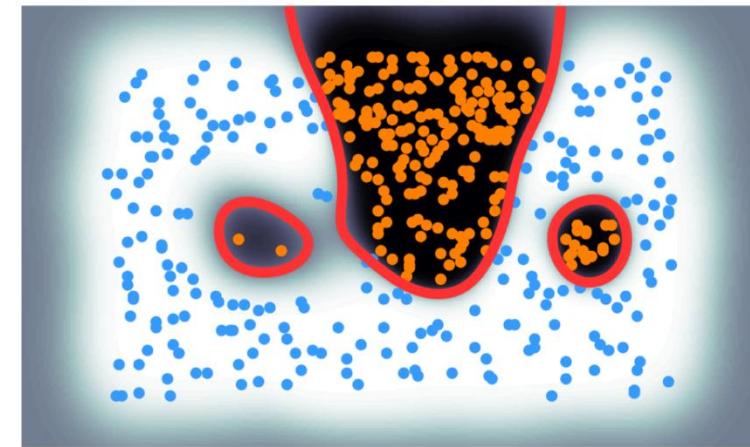
$$\text{Pers}_p(\mathcal{F}) = \{[b_\alpha, d_\alpha] \mid \alpha \in \mathcal{A}\}$$

starts and ends with indices in the filtration

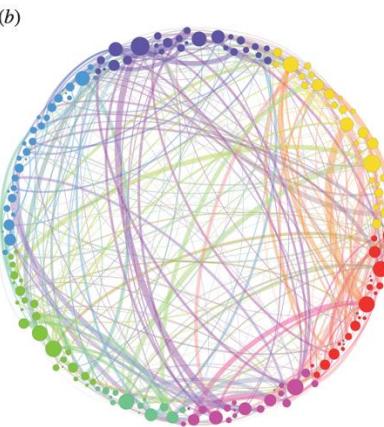
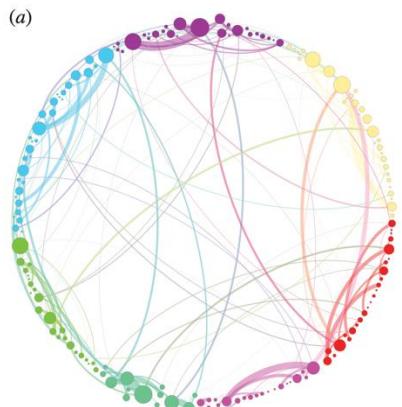
# Persistent homology: Applications



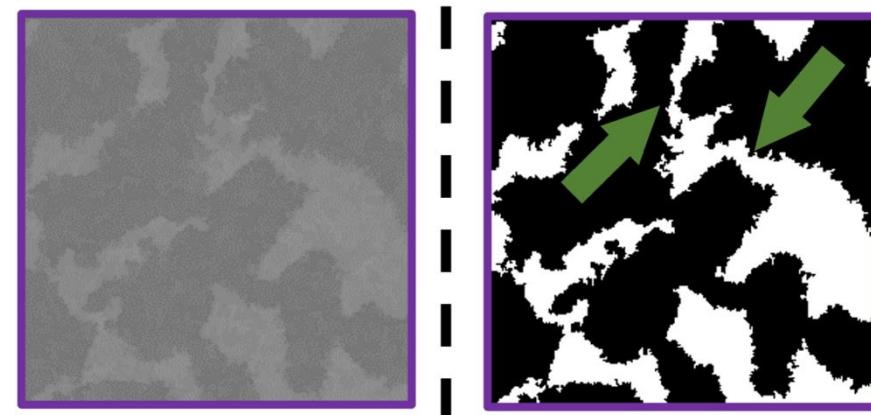
Features for ML [Zhao & Wang 19]



Topological regularizer for ML [Chen et al. 20]



Brain functional networks [Petri et al. 14]



Binarizing microstructures [Patel et al. 22]

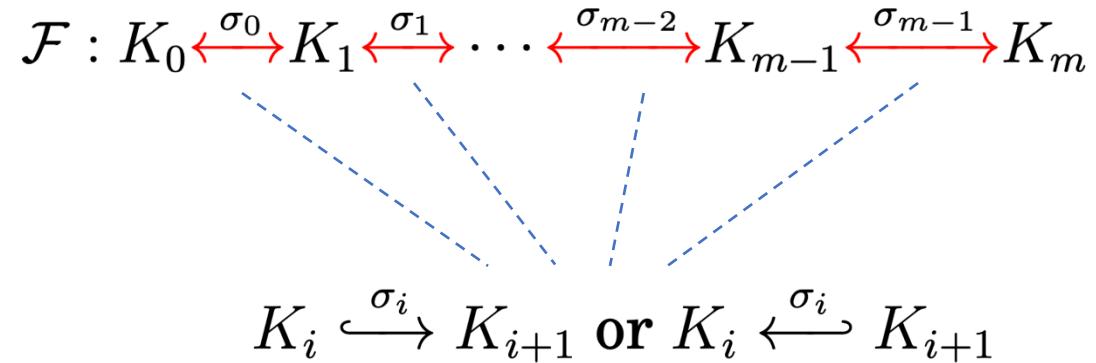
# Zigzag persistence

Zigzag filtration:

$$\mathcal{F} : K_0 \xleftarrow{\sigma_0} K_1 \xleftarrow{\sigma_1} \cdots \xleftarrow{\sigma_{m-2}} K_{m-1} \xleftarrow{\sigma_{m-1}} K_m$$

# Zigzag persistence

Zigzag filtration:



# Zigzag persistence

Zigzag filtration:

$$\mathcal{F} : K_0 \xleftrightarrow{\sigma_0} K_1 \xleftrightarrow{\sigma_1} \cdots \xleftrightarrow{\sigma_{m-2}} K_{m-1} \xleftrightarrow{\sigma_{m-1}} K_m$$



Induced module:

$$H_p(\mathcal{F}) : H_p(K_0) \leftrightarrow H_p(K_1) \leftrightarrow \cdots \leftrightarrow H_p(K_{m-1}) \leftrightarrow H_p(K_m)$$



Interval decomposition: [Gabriel 72]

$$H_p(\mathcal{F}) = \bigoplus_{\alpha \in \mathcal{A}} \mathcal{I}^{[b_\alpha, d_\alpha]}$$

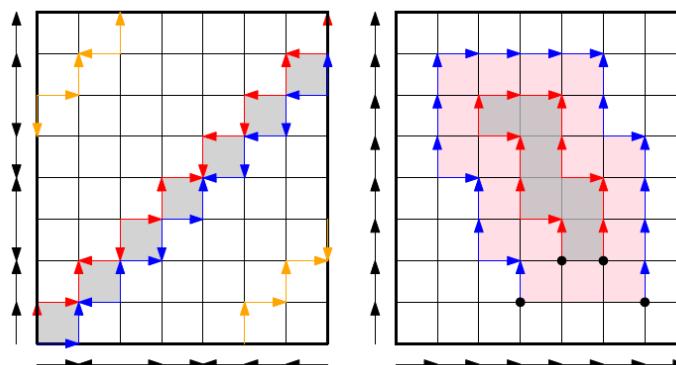


*p*-th persistence barcode:

$$\text{Pers}_p(\mathcal{F}) = \{[b_\alpha, d_\alpha] \mid \alpha \in \mathcal{A}\}$$

# Applications of Zigzag Persistence

- In time varying settings: functions, point cloud, vector field
  - G. Carlsson, V. de Silva, and D. Morozov. Zigzag persistent homology and real-valued functions. SoCG 2009.
  - W. Kim and F. Mémoli. Spatiotemporal persistent homology for dynamic metric spaces. DCG 2020.
  - T. Dey, M. Lipinsky, M. Mrozek, R. Słoboda. Tracking dynamical features via continuation and persistence. SoCG 2022.
- In multiparameter persistence



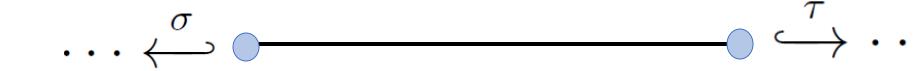
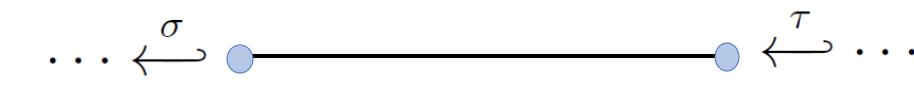
# Non-Zigzag vs. Zigzag persistence

1 Bars in **non-zigzag**: 1 type  
• closed-open



2 Bars in **zigzag**: 4 types

- closed-open
- closed-closed
- open-closed
- open-open



2 Simplices( $\sigma$ ) in **zigzag**: insertion( $\downarrow\sigma$ ), deletion( $\uparrow\sigma$ ), repeated( $\downarrow\sigma$ )

$$\mathcal{F} : \emptyset = K_0 \leftrightarrow \cdots \xleftarrow{\downarrow\sigma} \cdots \xleftarrow{\uparrow\sigma} \cdots \xleftarrow{\downarrow\sigma} \cdots \leftrightarrow K_m = \emptyset$$

# Non-Zigzag vs. Zigzag: Computing

## Non-zigzag [ELZ2000]

```

integer YOUNGEST (simplex  $\sigma^j$ )
 $\Lambda = \{\sigma \in \partial_{k+1}(\sigma^j) \mid \sigma \text{ positive}\};$ 
loop
     $i = \max(\Lambda);$ 
    if  $T[i]$  is unoccupied then
        store  $j$  and  $\Lambda$  in  $T[i]$ ; exit
    endif;
     $\Lambda = \Lambda + \Lambda^i$ 
forever;
return  $i$ .

```

**Case  $f_i$ :** We compute the representation of the boundary of simplex  $\sigma$  in terms of the cycles  $Z_i$ , and then reduce the result among the boundaries, obtaining:  $\partial\sigma = Z_i v = Z_i(B_i u + v')$ . There are two possibilities:

**Birth:** If  $v' = 0$ , then  $\partial\sigma$  is already a boundary, and addition of  $\sigma$  creates a new cycle, for example,  $C_i u - \sigma$ . We append this cycle to the matrix  $Z_i$ , and we append  $i+1$  to both the birth vector  $\mathbf{b}_i$  and the index vector  $\mathbf{idx}_i$  to get  $\mathbf{b}_{i+1}$  and  $\mathbf{idx}_{i+1}$ , respectively.

**Death:** If  $v' \neq 0$ , then let  $j$  be the row of the lowest non-zero element in vector  $v'$ . We output a pair  $(\mathbf{b}_i[j], i)$ . We append vector  $v'$  to the matrix  $B_i$ , and the corresponding chain  $(C_i u - \sigma)$  to the matrix  $C_i$  to obtain matrices  $B_{i+1}$  and  $C_{i+1}$ , respectively.

**Case  $g_i$ :** There are once again two possibilities:

**Birth:** There is no cycle in matrix  $Z_i$  that contains simplex  $\sigma$ . Let  $j$  be the index of the first column in  $C_i$  that contains  $\sigma$ , let  $l$  be the index of the row of the lowest non-zero element in  $B_i[j]$ .

1. Prepend  $D_i C_i[j]$  to  $Z_i$  to get  $Z'_i$ . Prepend  $i+1$  to the birth vector  $\mathbf{b}_i$  to get  $\mathbf{b}_{i+1}$ .
2. Let  $c = C_i[j][\sigma]$  be the coefficient of  $\sigma$  in the chain  $C_i[j]$ . Let  $\mathbf{r}_\sigma$  be the row of  $\sigma$  in matrix  $C_i$ . We prepend the row  $-\mathbf{r}_\sigma/c$  to the matrix  $B_i$  to get  $B'_i$ .
3. Subtract  $(\mathbf{r}_\sigma[k]/c) \cdot C_i[j]$  from every column  $C_i[k]$  to get  $C'_i$ .
4. Subtract  $(B'_i[k][l]/B'_i[j][l]) \cdot B'_i[j]$  from every other column  $B'_i[k]$ .

## Zigzag [CdSM2009]

5. Drop row  $l$  and column  $j$  from  $B'_i$  to get  $B_{i+1}$ , drop column  $l$  from  $Z'_i$ , and drop column  $j$  from  $C_i$  to get  $C_{i+1}$ .
6. Reduce  $Z_{i+1}$  initially set to  $Z'_i$ :
  - 1: **while**  $\exists k < j$  s.t.  $\text{low } Z_{i+1}[j] = \text{low } Z_{i+1}[k]$  **do**
  - 2:    $s = \text{low } Z_{i+1}[j]$ ,  $s_k^j = Z_{i+1}[j][s]/Z_{i+1}[k][s]$
  - 3:    $Z_{i+1}[j] = Z_{i+1}[j] - s_k^j \cdot Z_{i+1}[k]$
  - 4:   In  $B_{i+1}$ , add row  $j$  multiplied by  $s_k^j$  to row  $k$

We set the index  $\mathbf{idx}_{i+1}$  of the prepended cycle to be 1, and increase the index of every other column by 1. Figure 5 illustrates the changes made in this case.

**Death:** Let  $Z_i[j]$  be the first cycle that contains simplex  $\sigma$ . Output  $(\mathbf{b}_i[j], i)$ .

1. Change basis to remove  $\sigma$  from matrix  $Z_i$ :
  - 1: **for** increasing  $k > j$  s.t.  $\sigma \in Z_i[k]$  **do**
  - 2:   Let  $\sigma_j^k = Z_i[k][\sigma]/Z_i[j][\sigma]$
  - 3:    $Z_{i+1}[k] = Z_i[k] - \sigma_j^k \cdot Z_i[j]$
  - 4:   In  $B_i$ , add row  $k$  multiplied by  $\sigma_j^k$  to row  $j$
  - 5:   **if**  $\text{low } Z_{i+1}[k] > \text{low } Z_i[k]$  **then**
  - 6:        $j = k$
2. Subtract cycle  $(C_i[k][\sigma]/Z_i[j][\sigma]) \cdot Z_i[j]$  from every chain  $C_i[k]$ .
3. Drop  $Z_{i+1}[j]$ , the corresponding entry in vectors  $\mathbf{b}_i$  and  $\mathbf{idx}_i$ , row  $j$  from  $B_i$ , row  $\sigma$  from  $C_i$  and  $Z_i$  (as well as row and column of  $\sigma$  from  $D_i$ ).

We increase the index of every column by 1,  $\mathbf{idx}_{i+1}(l) = \mathbf{idx}_i(l) + 1$ .

# Outline

1. An algorithm for computing zigzag persistence (FastZigzag)
  - Converts to a computation of non-zigzag persistence
  - Bridges gap of efficiency for computing the two versions
2.  $O(m \log m)$  algorithm for computing graph zigzag persistence
3. Algorithms for updating zigzag persistence over local changes
  - Focus on contractions and expansions
  - Match the  $O(m^2)$  complexity of the non-zigzag version
4. Algorithms for updating graph persistence (over switches)
  - Non-zigzag:  $O(\log m)$
  - Zigzag:  $O(\sqrt{m} \log m)$
5.  $O(m^2 n)$  algorithm for computing zigzag representatives

# Fast computation of zigzag persistence

# Complexities of persistence computing

	Theoretical	In Practice
Standard	$O(m^\omega)$	<i>Various optimizations</i>
Zigzag	$O(m^\omega)$	<i>Much slower</i>

$\omega \approx 2.37286$ , matrix multiplication exponent

Edelsbrunner, Letscher, Zomorodian. Topological persistence and simplification. FoCS 2000.

Carlsson, de Silva, Morozov. Zigzag persistent homology and real-valued functions. SoCG 2009.

Milosavljević, Morozov, Skraba. Zigzag persistent homology in matrix multiplication time. SoCG 2011.

Clément Maria and Steve Y. Oudot. Zigzag persistence via reflections and transpositions. SODA 2015.

# Overview of FastZigzag

- Input zigzag filtration

$$\mathcal{F} : \emptyset = K_0 \xleftarrow{\sigma_0} K_1 \xleftarrow{\sigma_1} \cdots \xleftarrow{\sigma_{m-1}} K_m = \emptyset$$

- Convert to a **non-zigzag filtration** of **same length** (linear time)

$$\mathcal{F}' : K'_0 \xrightarrow{\sigma'_0} K'_1 \xrightarrow{\sigma'_1} \cdots \xrightarrow{\sigma'_{m-1}} K'_m$$

- Compute barcode for **non-zigzag filtration**  $\mathcal{F}'$ 
  - Fast software [Gudhi, Phat, Dionysus etc.]
- Convert barcode of  $\mathcal{F}'$  to that of  $\mathcal{F}$ 
  - $O(1)$  conversion per bar

*Overall conversion has very little cost*

# Conversion of Filtrations in FastZigzag

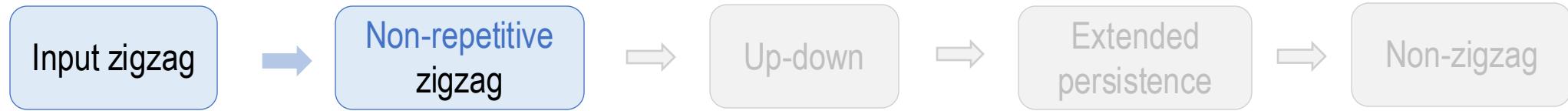


All filtrations have the **same** length (the same number of addition/deletions)

Conversions 1,2,3,4:

- Done by a simple **linear scan** of the input filtration

# Conversion of Filtrations in FastZigzag



**Non-repetitive filtration:** A simplex is added at most one time

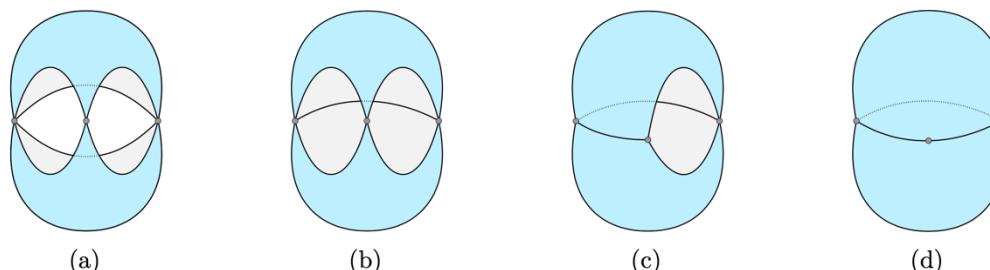
- Idea: Treat new occurrence of simplex  $\sigma$  as a **new copy** (barcodes stay the same)

$$\mathcal{F} : \emptyset = K_0 \leftrightarrow \dots \xrightarrow{\sigma} \dots \xleftarrow{\sigma} \dots \xrightarrow{\sigma} \dots \xleftarrow{\sigma} \dots \leftrightarrow K_m = \emptyset$$

↓

$$\hat{\mathcal{F}} : \emptyset = \hat{K}_0 \leftrightarrow \dots \xrightarrow{\hat{\sigma}_1} \dots \xleftarrow{\hat{\sigma}_1} \dots \xrightarrow{\hat{\sigma}_2} \dots \xleftarrow{\hat{\sigma}_2} \dots \leftrightarrow \hat{K}_m = \emptyset$$

- Simplices with the same vertex set shall occur in same complex in later filtration:  
use  **$\Delta$ -complex** [Hatcher02]



Two triangles sharing 0,1,2,3 edges

# Conversion of Filtrations in FastZigzag

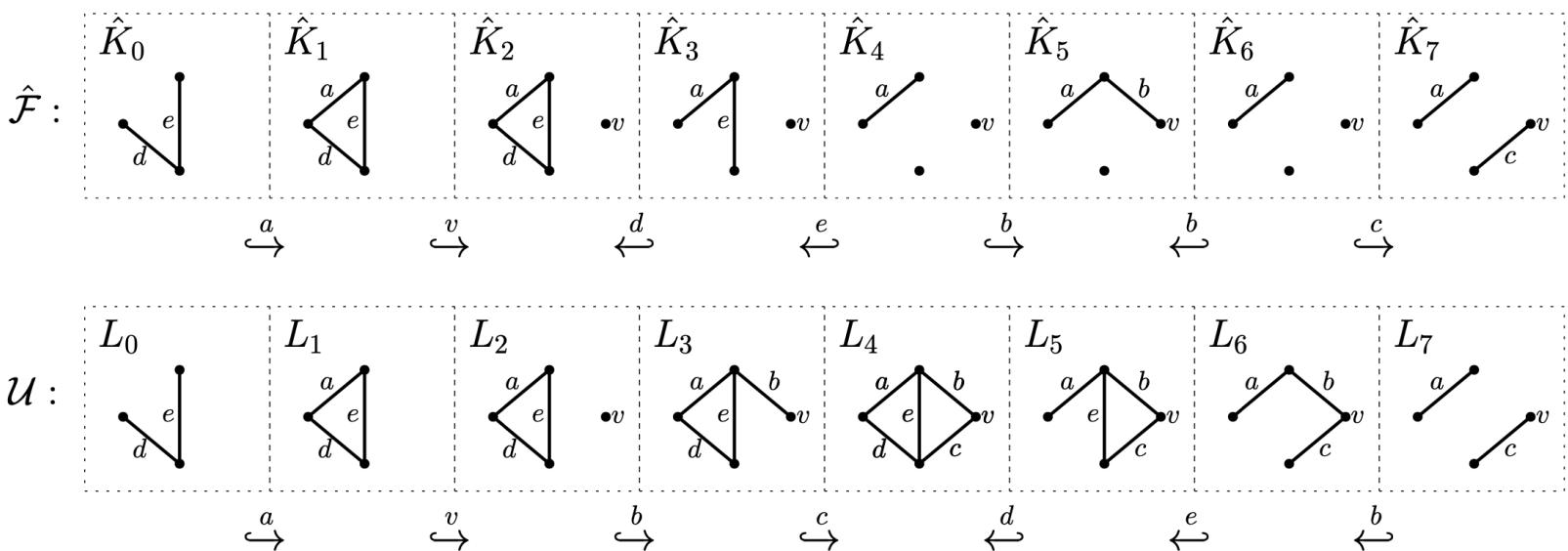


$$\hat{\mathcal{F}} : \emptyset = \hat{K}_0 \xleftarrow{\hat{\sigma}_0} \hat{K}_1 \xleftarrow{\hat{\sigma}_1} \cdots \xleftarrow{\hat{\sigma}_{m-1}} \hat{K}_m = \emptyset$$

↓

$$\mathcal{U} : \emptyset = L_0 \xrightarrow{\tau_0} \cdots \xrightarrow{\tau_{n-1}} L_n \xleftarrow{\tau_n} \cdots \xleftarrow{\tau_{2n-1}} L_{2n} = \emptyset \quad (m = 2n)$$

List the additions in  $\hat{\mathcal{F}}$  first and then the deletions in  $\hat{\mathcal{F}}$ , following the orders in  $\hat{\mathcal{F}}$



# Conversion of Filtrations in FastZigzag



$$\mathcal{U} : \emptyset = L_0 \xleftarrow{\tau_0} \cdots \xleftarrow{\tau_{n-1}} L_n \xleftarrow{\tau_n} \cdots \xleftarrow{\tau_{2n-2}} L_{2n-1} \xleftarrow{\tau_{2n-1}} L_{2n} = \emptyset$$

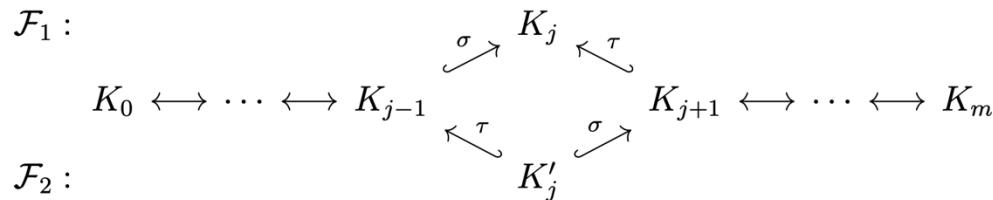


$$\mathcal{E} : \emptyset = L_0 \xleftarrow{\tau_0} \cdots \xleftarrow{\tau_{n-1}} L_n = (\hat{K}, L_{2n}) \xleftarrow{\tau_{2n-1}} (\hat{K}, L_{2n-1}) \xleftarrow{\tau_{2n-2}} \cdots \xleftarrow{\tau_n} (\hat{K}, L_n) = (\hat{K}, \hat{K})$$

# Conversion of Filtrations in FastZigzag



Mayer-Vietoris Diamond [CdS10]

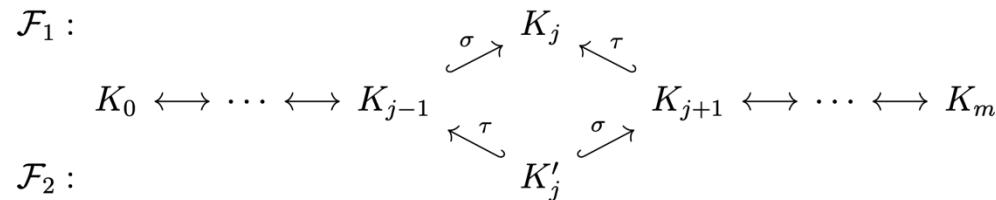


- $\mathcal{F}_1$  to  $\mathcal{F}_2$ : Outward switch
- $\mathcal{F}_2$  to  $\mathcal{F}_1$ : Inward switch

# Conversion of Filtrations in FastZigzag



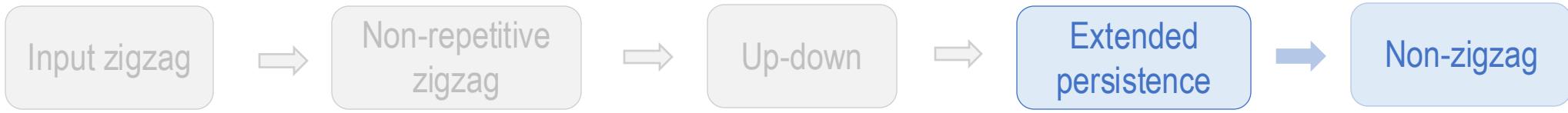
Mayer-Vietoris Diamond [CdS10]



- $\mathcal{F}_1$  to  $\mathcal{F}_2$ : Outward switch
- $\mathcal{F}_2$  to  $\mathcal{F}_1$ : Inward switch

Major takeaway: there is a bijection between the barcodes of the filtrations s.t. corresponding intervals have same creator and destroyer simplices (cells)

# Conversion of Filtrations in FastZigzag

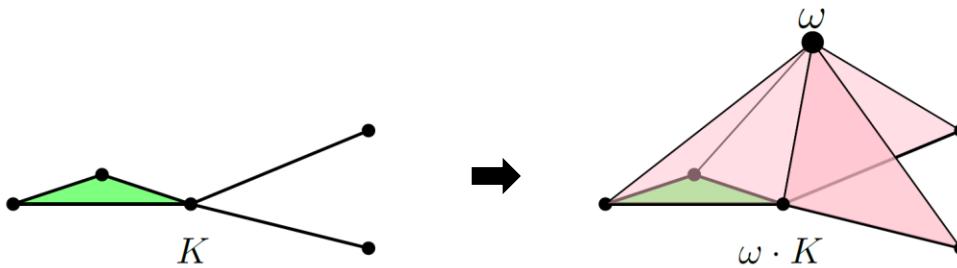


$$\mathcal{E} : \emptyset = L_0 \hookrightarrow \cdots \hookrightarrow L_n = (\hat{K}, L_{2n}) \hookrightarrow (\hat{K}, L_{2n-1}) \hookrightarrow \cdots \hookrightarrow (\hat{K}, L_n) = (\hat{K}, \hat{K})$$

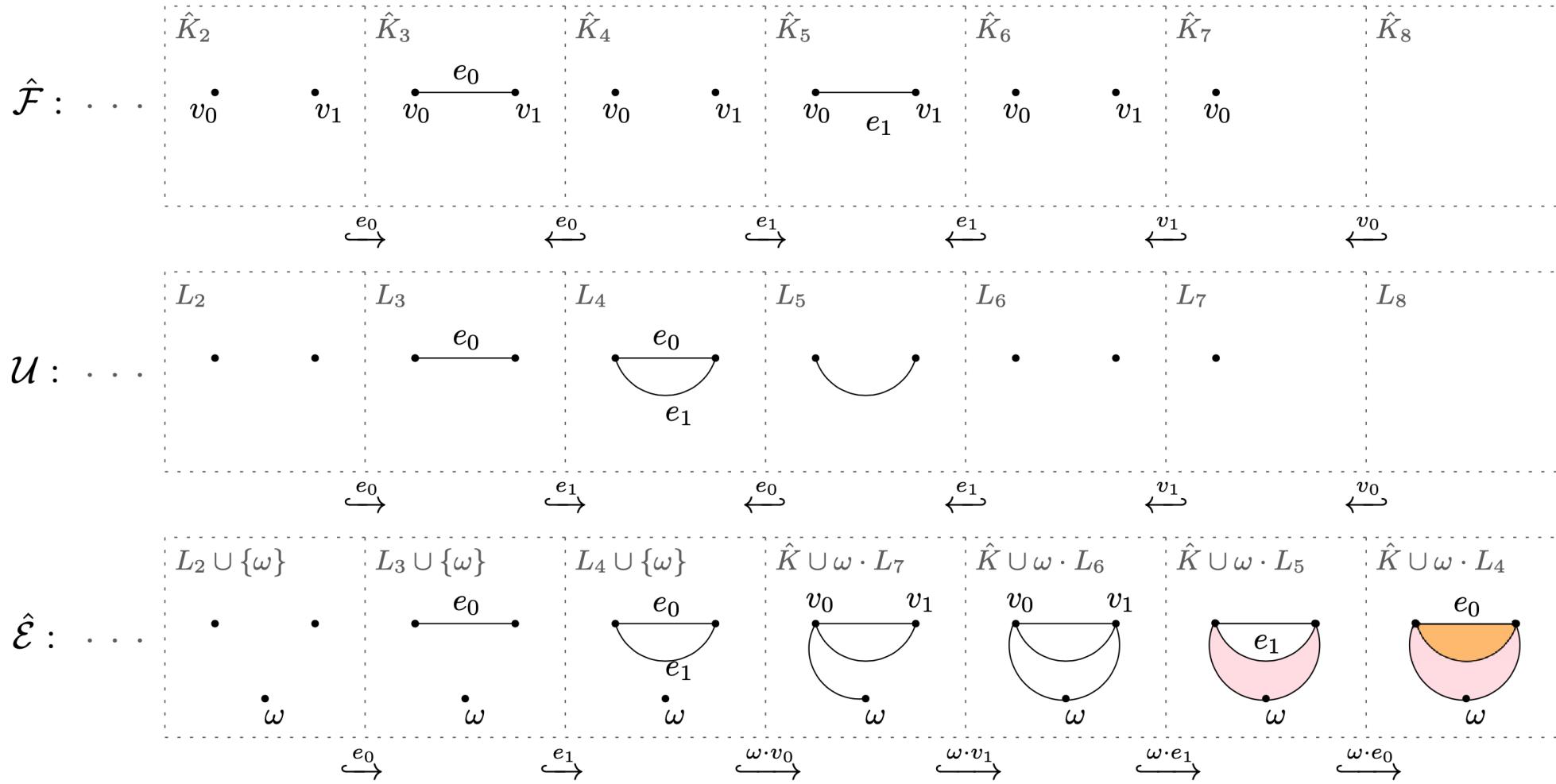


$$\hat{\mathcal{E}} : L_0 \cup \{\omega\} \hookrightarrow \cdots \hookrightarrow L_n \cup \{\omega\} = \hat{K} \cup \omega \cdot L_{2n} \hookrightarrow \hat{K} \cup \omega \cdot L_{2n-1} \hookrightarrow \cdots \hookrightarrow \hat{K} \cup \omega \cdot L_n$$

Use ‘Coning’ [CEH09]: No change in barcode



# Overall Conversions



# Pseudocodes for the conversion

---

**Algorithm 3.1** Pseudocode for converting input filtration

---

```
1: procedure CONVERTFILT( $\mathcal{F}$ )
2:   initialize boundary matrix  $D$ , cell-id map  $\text{cid}$ , deleted cell list  $\text{del\_list}$  as empty
3:   append an empty column to  $D$  representing vertex  $\omega$  for coning
4:    $\text{id} \leftarrow 1$                                       $\triangleright$  variable keeping track of id for cells
5:   for each  $K_i \xrightarrow{\sigma_i} K_{i+1}$  in  $\mathcal{F}$  do
6:     if  $\sigma_i$  is being inserted then
7:        $\text{cid}[\sigma_i] = \text{id}$                           $\triangleright$  get a new cell as a copy of simplex  $\sigma_i$ 
8:        $\text{col} \leftarrow \text{CELLBOUNDARY}(\sigma_i, \text{cid})$ 
9:       append  $\text{col}$  to  $D$ 
10:       $\text{id} \leftarrow \text{id} + 1$ 
11:    else
12:      append  $\text{cid}[\sigma_i]$  to  $\text{del\_list}$ 
13:   initialize map  $\text{cone\_id}$  as empty            $\triangleright$   $\text{cone\_id}$  tracks id for coned cells
14:   for each  $\text{del\_id}$  in  $\text{del\_list}$  (accessed reversely) do
15:      $\text{cone\_id}[\text{del\_id}] \leftarrow \text{id}$            $\triangleright$  get a new coned cell
16:      $\text{col} \leftarrow \text{CONEDCELLBOUNDARY}(\text{del\_id}, D, \text{cone\_id})$ 
17:     append  $\text{col}$  to  $D$ 
18:      $\text{id} \leftarrow \text{id} + 1$ 
19:   return  $D$ 
```

---

# Running time comparison

1-2: Non-repetitive random shuffles from height functions on triangular meshes

3-8: Clique complexes from random edge additions/deletions

9-11: Oscillating Rips zigzag from point clouds of 2000 – 4000 sampled from triangular meshes

	No.	Length	D	Rep	MaxK	T <sub>DIO2</sub>	T <sub>GUDHI</sub>	T <sub>FZZ</sub>	SU
	1	5,260,700	5	1.0	883,350	2h02m46.0s	—	8.9s	873
	2	5,254,620	4	1.0	1,570,326	19m36.6s	—	11.0s	107
	3	5,539,494	5	1.3	1,671,047	3h05m00.0s	45m47.0s	3m20.8s	13.7
	4	5,660,248	4	2.0	1,385,979	2h59m57.0s	29m46.7s	4m59.5s	6.0
	5	5,327,422	4	3.5	760,098	43m54.8s	10m35.2s	3m32.1s	3.0
	6	5,309,918	3	5.1	523,685	5h46m03.0s	1h32m37.0s	19m30.2s	4.7
	7	5,357,346	3	7.3	368,830	3h37m54.0s	57m28.4s	30m25.2s	1.9
	8	6,058,860	4	9.1	331,211	53m21.2s	7m19.0s	3m44.4s	2.0
	9	5,135,720	3	21.9	11,859	23.8s	15.6s	8.6s	1.9
	10	5,110,976	3	27.7	11,435	36.2s	39.9s	8.5s	4.3
	11	5,811,310	4	44.2	7,782	38.5s	36.9s	23.9s	1.5

- All run on Intel(R), Core™, i5-9500 CPU@3.00GHz, 16GB memory, Linux OS
- Software **FZZ** using **Phat** software for non-zigzag (<https://github.com/taohou01/fzz>)

# Running time comparison

1-2: Non-repetitive random shuffles from height functions on triangular meshes

3-8: Clique complexes from random edge additions/deletions

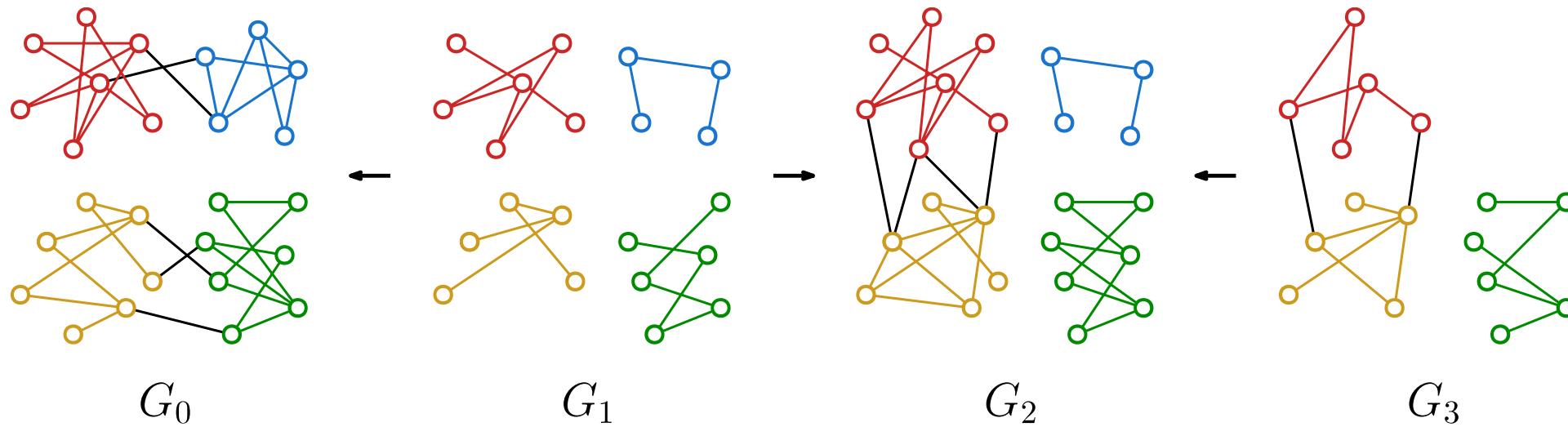
9-11: Oscillating Rips zigzag from point clouds of 2000 – 4000 sampled from triangular meshes

	No.	Length	D	Rep	MaxK	T <sub>DIO2</sub>	T <sub>GUDHI</sub>	T <sub>FZZ</sub>	SU
1-2: Non-repetitive random shuffles from height functions on triangular meshes	1	5,260,700	5	1.0	883,350	2h02m46.0s	—	8.9s	873
	2	5,254,620	4	1.0	1,570,326	19m36.6s	—	11.0s	107
	3	5,539,494	5	1.3	1,671,047	3h05m00.0s	45m47.0s	3m20.8s	13.7
	4	5,660,248	4	2.0	1,385,979	2h59m57.0s	29m46.7s	4m59.5s	6.0
	5	5,327,422	4	3.5	760,098	43m54.8s	10m35.2s	3m32.1s	3.0
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- All run on Intel(R), Core™, i5-9500 CPU @ 3.00GHz, 16GB memory, Linux OS
- Software **FZZ** using **Phat** software for non-zigzag (<https://github.com/taohou01/fzz>)

$O(m \log m)$  computation of graph zigzag persistence

# An application of graph zigzag persistence: Dynamic networks



Petter Holme and Jari Saramaki. **Temporal networks**. Physics Reports, 519(3):97–125, 2012.

# Complexities of persistence computing

---

*	Graphs	
Standard	$O(m^\omega)$	$O(m \alpha(m))$
Zigzag	$O(m^\omega)$	$O(m \log^4 n)$

---

$m$ : length of filtration

$\omega \approx 2.37286$ : matrix multiplication exponent

$\alpha(m)$ : inverse Ackermann function

Input for graph zigzag:

$$\mathcal{F} : \emptyset = G_0 \xleftarrow{\sigma_0} G_1 \xleftarrow{\sigma_1} \dots \xleftarrow{\sigma_{m-1}} G_m; G = \bigcup_{i=0}^m G_i$$

$n$ : size of  $G$

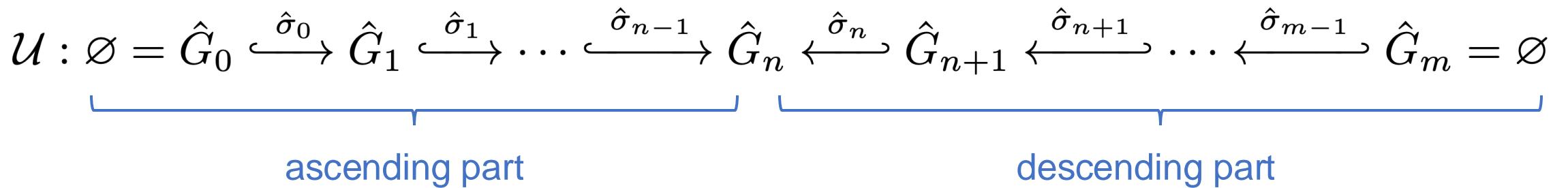
# Computation

Utilize the conversion in FastZigzag to convert the input zigzag into an up-down filtration  $\mathcal{U}$ , with the following barcode mapping:

	$\mathcal{U}$		$\mathcal{F}$
1	$\text{Pers}_0^{\text{co}}(\mathcal{U})$	$\leftrightarrow$	$\text{Pers}_0^{\text{co}}(\mathcal{F})$
2	$\text{Pers}_0^{\text{oc}}(\mathcal{U})$	$\leftrightarrow$	$\text{Pers}_0^{\text{oc}}(\mathcal{F})$
3	$\text{Pers}_0^{\text{cc}}(\mathcal{U})$	$\leftrightarrow$	$\text{Pers}_0^{\text{cc}}(\mathcal{F})$
Compute them separately	$\text{Pers}_1^{\text{cc}}(\mathcal{U})$	$\leftrightarrow$	$\text{Pers}_0^{\text{oo}}(\mathcal{F}) \cup \text{Pers}_1^{\text{cc}}(\mathcal{F})$

# Computation

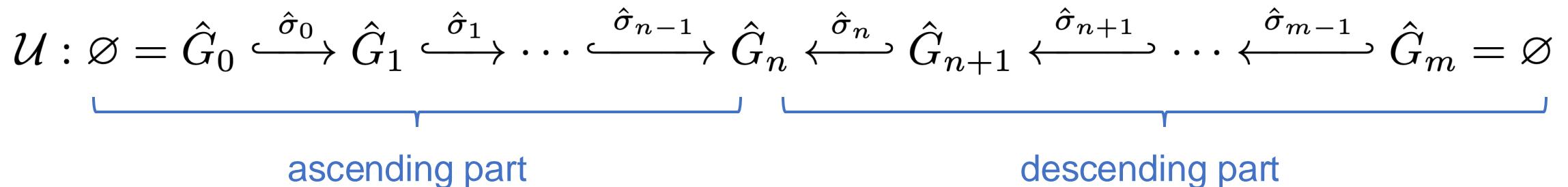
1.  $\text{Pers}_0^{\text{CO}}(\mathcal{U})$ ,  $\text{Pers}_0^{\text{OC}}(\mathcal{U})$ : run the persistence pairing for 0-dimensional standard persistence with Union-Find on the *ascending* and *descending* parts of  $\mathcal{U}$  in  $O(m \alpha(m))$  time



# Computation

2.  $\text{Pers}_0^{\text{CC}}(\mathcal{U})$  :

- Identify each connected component  $C$  of  $\hat{G}_n$
- Pair:
  - first vertex in the **ascending** part coming from  $C$
  - first vertex in the **descending** part coming from  $C$
- Can be done in linear time



# Computation

3.  $\text{Pers}_1^{\text{CC}}(\mathcal{U})$  : from the edge-edge pairs; the first edge is a positive edge from the ascending part  $\mathcal{U}_u$ , the second edge is a positive edge from the descending part  $\mathcal{U}_d$ .

*Positive* edge: connect to the same connected component

*Negative* edge: connect to the different connected components

# Computation

3.  $\text{Pers}_1^{\text{CC}}(\mathcal{U})$  : from the edge-edge pairs; the first edge is a positive edge from the ascending part  $\mathcal{U}_u$ , the second edge is a positive edge from the descending part  $\mathcal{U}_d$ .

## ► Algorithm.

1. Maintain a spanning forest  $T$  of  $\hat{G}_n$  while processing  $\mathcal{U}_d$ . Initially,  $T$  consists of all vertices of  $\hat{G}_n$  and all negative edges in  $\mathcal{U}_d$ .
2. For every positive edge  $e$  in  $\mathcal{U}_d$ :
  - a. Add  $e$  to  $T$  and check the *unique* cycle  $c$  formed by  $e$  in  $T$ .
  - b. Determine the edge  $e'$  which is the youngest edge of  $c$  with respect to the filtration  $\mathcal{U}_u$ .  
The edge  $e'$  has to be positive in  $\mathcal{U}_u$ .
  - c. Delete  $e'$  from  $T$ . This maintains  $T$  to be a tree all along.
  - d. Pair the positive edge  $e$  from  $\mathcal{U}_d$  with the positive edge  $e'$  from  $\mathcal{U}_u$ .

# Computation

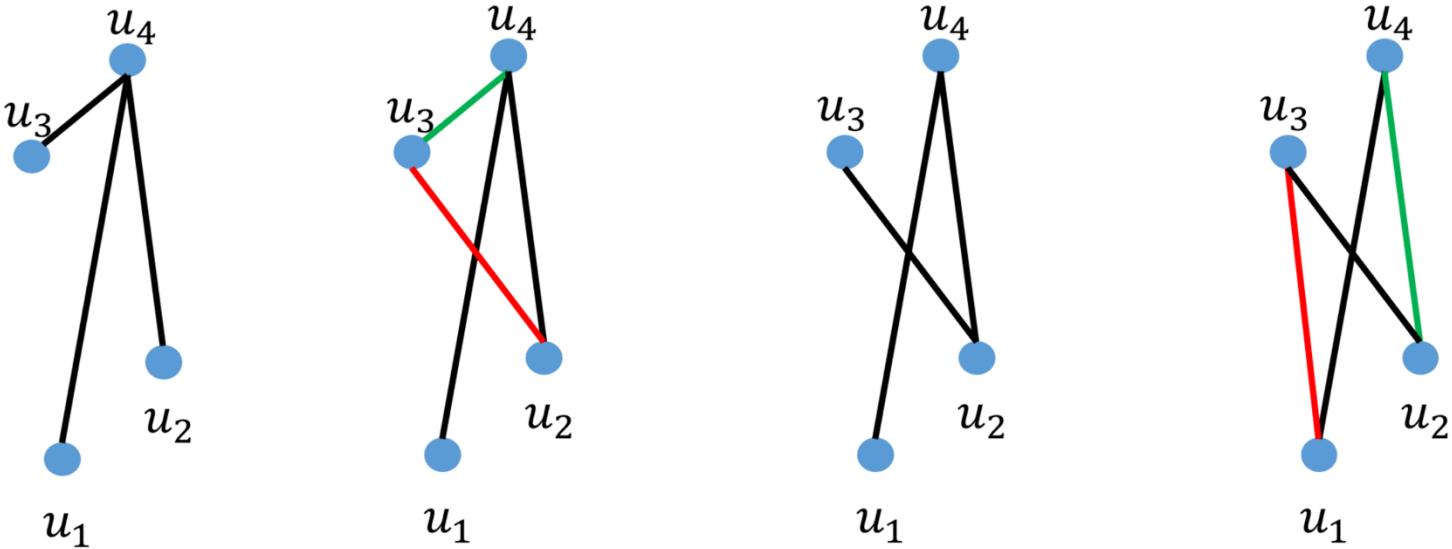


Figure from [Yan et al. 2021]

# Computation

## 3. $\text{Pers}_1^{\text{CC}}(\mathcal{U})$ :

- The algorithm in [Yan et al. 2021] runs in  $O(m^2)$  time using a direct implementation for the trees
- We propose to use the Link-Cut tree [Sleator, Tarjan, 1981] so that finding the edge-edge pairs runs in  $O(m \log m)$  time.
- Since the conversions between input zigzag and up-down are linear time, the overall complexity is  $O(m \log m)$ .

Daniel D. Sleator and Robert Endre Tarjan. A data structure for dynamic trees. 1981.

# Updating zigzag persistence

# Background

Consider local changes on filtration and update the barcode accordingly

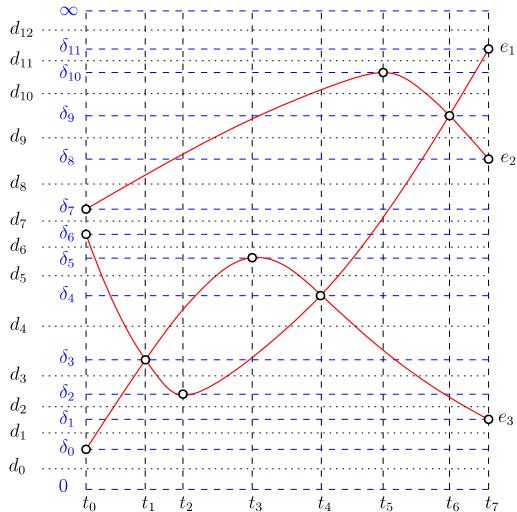
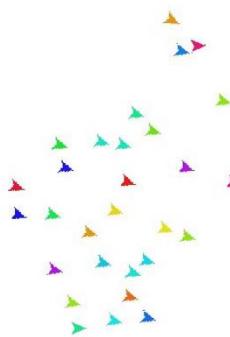
- Produces *vineyard* (a stack of barcodes)

# Background

Consider local changes on filtration and update the barcode accordingly

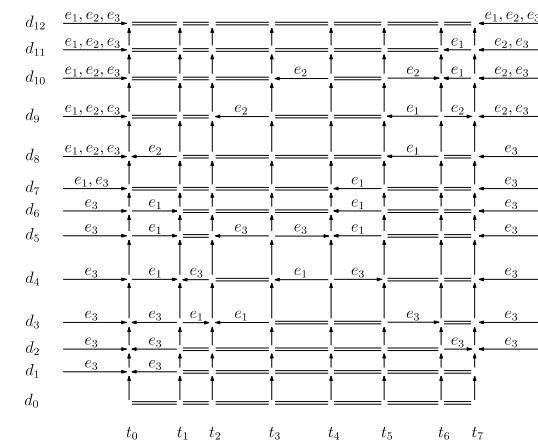
- Produces **vineyard** (a stack of barcodes)

## Example: Dynamic point cloud

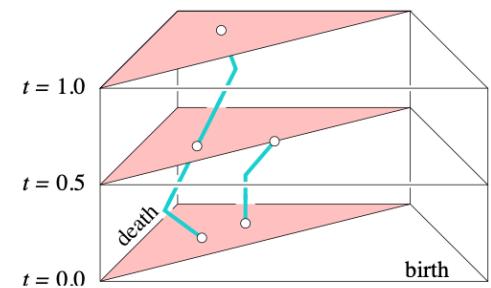


Point cloud moving  
with time

Distance-time curves of all pairs



Zigzag filtrations changing  
over distance



Vineyard  
(figure from Computational  
topology: An Introduction)

# Background

Consider local changes on filtration and update the barcode accordingly

- Produces *vineyard* (a stack of barcodes)

**Operations in standard persistence, computed in  $O(m)$  time [CEM06]**

Switch (transposition)

$$\mathcal{F} : \emptyset = K_0 \hookrightarrow \cdots \hookrightarrow K_{i-1} \xrightarrow{\sigma} K_i \xleftarrow{\tau} K_{i+1} \hookrightarrow \cdots \hookrightarrow K_m$$
$$\mathcal{F}' : \emptyset = K_0 \hookrightarrow \cdots \hookrightarrow K_{i-1} \xleftarrow{\tau} K'_i \xrightarrow{\sigma} K_{i+1} \hookrightarrow \cdots \hookrightarrow K_m$$

# Background

Consider local changes on filtration and update the barcode accordingly

- Produces **vineyard** (a stack of barcodes)

## Operations we consider

### Forward switch

$$\mathcal{F} : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-1} \xrightarrow{\sigma} K_i \xrightarrow{\tau} K_{i+1} \leftrightarrow \cdots \leftrightarrow K_m$$
$$\mathcal{F}' : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-1} \xrightarrow{\tau} K'_i \xrightarrow{\sigma} K_{i+1} \leftrightarrow \cdots \leftrightarrow K_m$$

### Backward switch

$$\mathcal{F} : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-1} \xleftarrow{\sigma} K_i \xleftarrow{\tau} K_{i+1} \leftrightarrow \cdots \leftrightarrow K_m$$
$$\mathcal{F}' : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-1} \xleftarrow{\tau} K'_i \xleftarrow{\sigma} K_{i+1} \leftrightarrow \cdots \leftrightarrow K_m$$

### Outward/inward switch

$$\mathcal{F} : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-1} \xrightarrow{\sigma} K_i \xleftarrow{\tau} K_{i+1} \leftrightarrow \cdots \leftrightarrow K_m$$
$$\mathcal{F}' : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-1} \xleftarrow{\tau} K'_i \xrightarrow{\sigma} K_{i+1} \leftrightarrow \cdots \leftrightarrow K_m$$

Keep filtration size

### Inward contraction/expansion

$$\mathcal{F} : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-2} \leftrightarrow K_{i-1} \xrightarrow{\sigma} K_i \xleftarrow{\sigma} K_{i+1} \leftrightarrow K_{i+2} \leftrightarrow \cdots \leftrightarrow K_m$$
$$\mathcal{F}' : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-2} \leftrightarrow K'_i \leftrightarrow K_{i+2} \leftrightarrow \cdots \leftrightarrow K_m$$

### Outward contraction/expansion

$$\mathcal{F} : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-2} \leftrightarrow K_{i-1} \xleftarrow{\sigma} K_i \xrightarrow{\sigma} K_{i+1} \leftrightarrow K_{i+2} \leftrightarrow \cdots \leftrightarrow K_m$$
$$\mathcal{F}' : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-2} \leftrightarrow K'_i \leftrightarrow K_{i+2} \leftrightarrow \cdots \leftrightarrow K_m$$

Increase/decrease filtration size

# Easy updates

For following **switch** operations (do not change input length):

- Barcodes can be easily updated in  $O(m)$  time
- Using the conversion in FastZigzag

## Forward switch

$$\begin{aligned}\mathcal{F} : K_0 \leftrightarrow \dots \leftrightarrow K_{i-1} &\xrightarrow{\sigma} K_i \xleftarrow{\tau} K_{i+1} \leftrightarrow \dots \leftrightarrow K_m \\ \mathcal{F}' : K_0 \leftrightarrow \dots \leftrightarrow K_{i-1} &\xleftarrow{\tau} K'_i \xrightarrow{\sigma} K_{i+1} \leftrightarrow \dots \leftrightarrow K_m\end{aligned}$$

## Backward switch

$$\begin{aligned}\mathcal{F} : K_0 \leftrightarrow \dots \leftrightarrow K_{i-1} &\xleftarrow{\sigma} K_i \xleftarrow{\tau} K_{i+1} \leftrightarrow \dots \leftrightarrow K_m \\ \mathcal{F}' : K_0 \leftrightarrow \dots \leftrightarrow K_{i-1} &\xleftarrow{\tau} K'_i \xleftarrow{\sigma} K_{i+1} \leftrightarrow \dots \leftrightarrow K_m\end{aligned}$$

## Outward/inward switch

$$\begin{aligned}\mathcal{F} : K_0 \leftrightarrow \dots \leftrightarrow K_{i-1} &\xrightarrow{\sigma} K_i \xleftarrow{\tau} K_{i+1} \leftrightarrow \dots \leftrightarrow K_m \\ \mathcal{F}' : K_0 \leftrightarrow \dots \leftrightarrow K_{i-1} &\xleftarrow{\tau} K'_i \xrightarrow{\sigma} K_{i+1} \leftrightarrow \dots \leftrightarrow K_m\end{aligned}$$

# Difficulties with (outward) contraction/expansion

Difficulties lie in (outward) contraction/expansion (input length changes)

Inward contraction/expansion

$$\mathcal{F} : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-2} \leftrightarrow K_{i-1} \xleftarrow{\sigma} K_i \xleftarrow{\sigma} K_{i+1} \leftrightarrow K_{i+2} \leftrightarrow \cdots \leftrightarrow K_m \leftarrow \text{---}$$
$$\mathcal{F}' : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-2} \leftrightarrow K'_i \leftrightarrow K_{i+2} \leftrightarrow \cdots \leftrightarrow K_m \leftarrow \text{---}$$

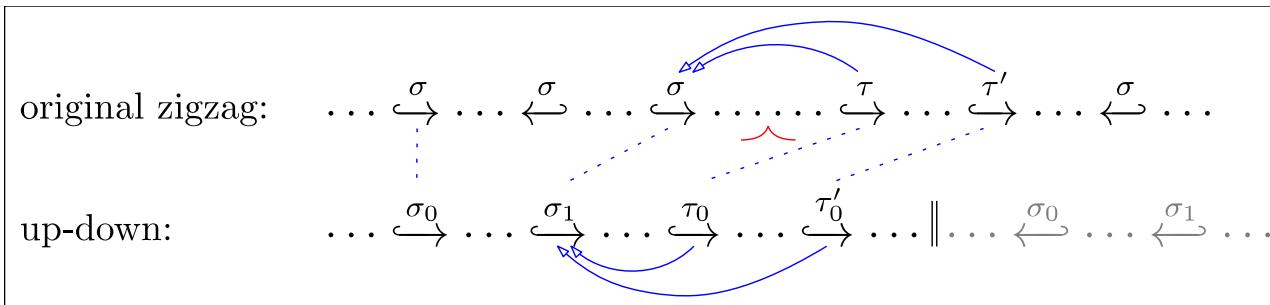
Outward contraction/expansion

$$\mathcal{F} : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-2} \leftrightarrow K_{i-1} \xleftarrow{\sigma} K_i \xrightarrow{\sigma} K_{i+1} \leftrightarrow K_{i+2} \leftrightarrow \cdots \leftrightarrow K_m \leftarrow \text{---}$$
$$\mathcal{F}' : K_0 \leftrightarrow \cdots \leftrightarrow K_{i-2} \leftrightarrow K'_i \leftrightarrow K_{i+2} \leftrightarrow \cdots \leftrightarrow K_m \leftarrow \text{---}$$

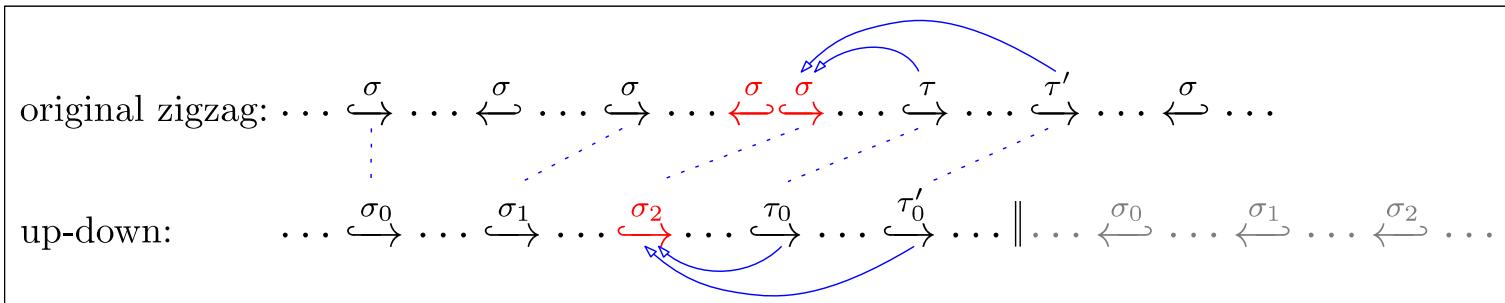
# Difficulties with (outward) contraction/expansion

- If we convert the zigzag filtrations into up-down/non-zigzag filtrations, there are some **adjacency change** on the cells:
  - *Before and after the operation, boundary faces of certain  $(p + 1)$ -cells change into other  $p$ -cells (which come in earlier/later in the up-down/non-zigzag filtration)*
- Straightforward approach takes  $O(m^3)$  time

Before outward expansion

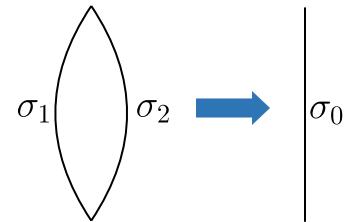


After outward expansion



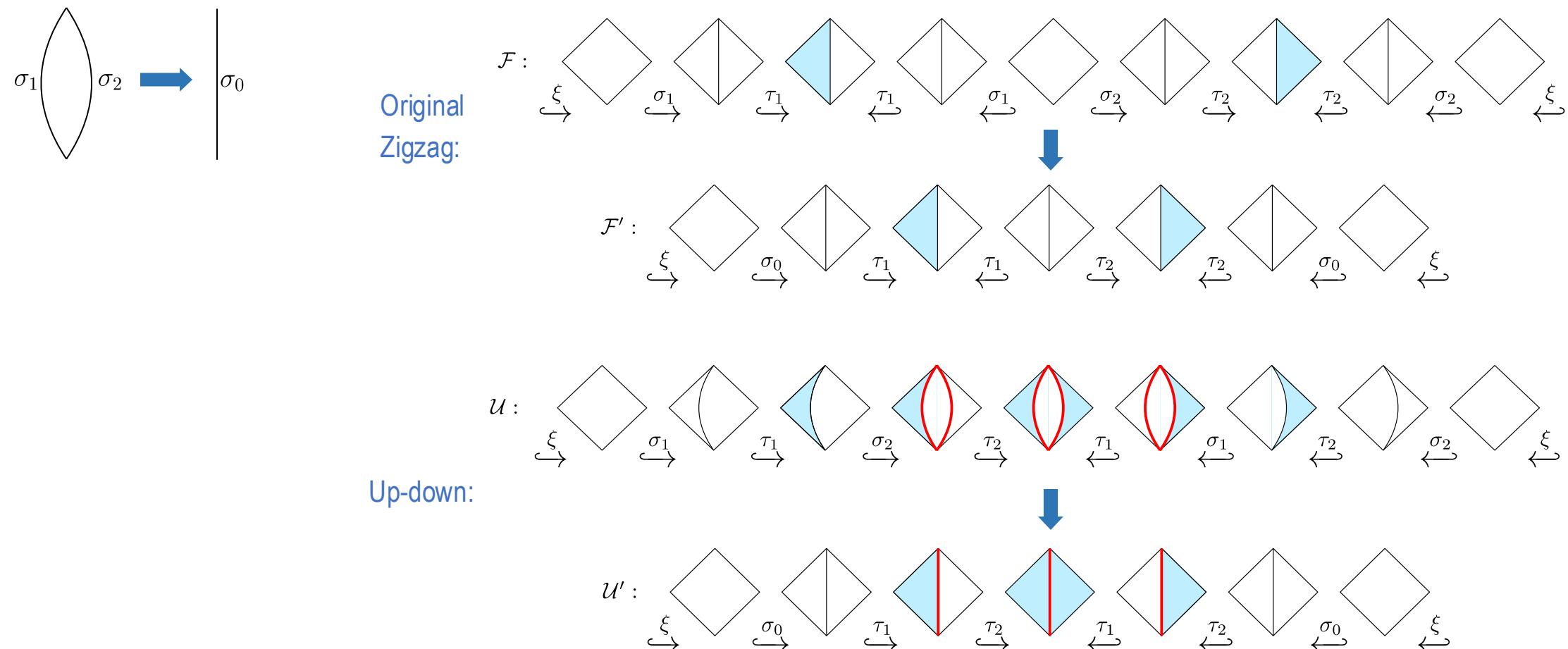
# Idea of the computation for outward contraction

- Convert input zigzag filtration into **up-down filtration**
- [Observation] The boundary change of cells in the up-down filtration in the contraction:
  - Two  $p$ -cells  $\sigma_1, \sigma_2$  are identified as the same  $p$ -cell  $\sigma_0$

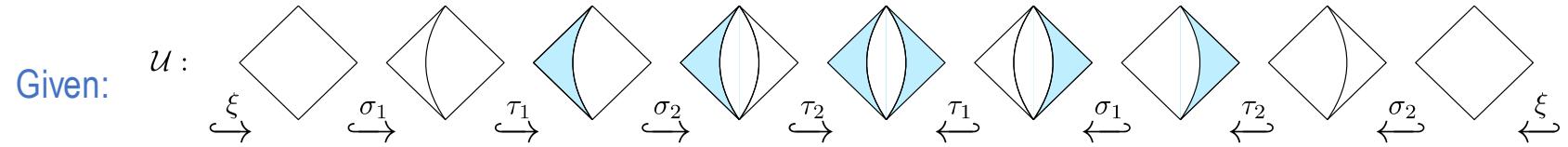


# Idea of the computation for outward contraction

- Convert input zigzag filtration into **up-down filtration**
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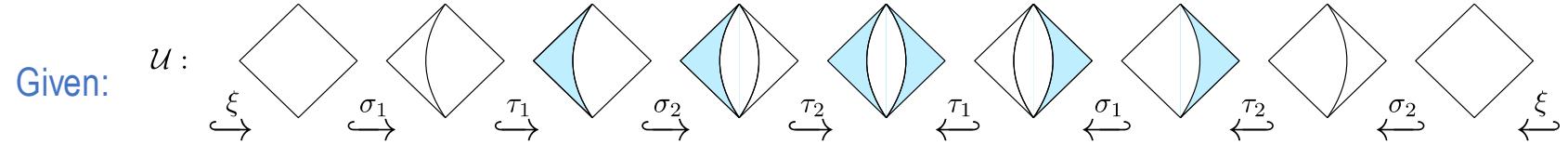


# Solution for outward contraction

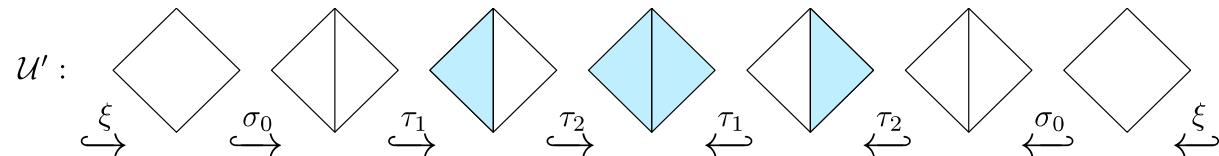


# Solution for outward contraction

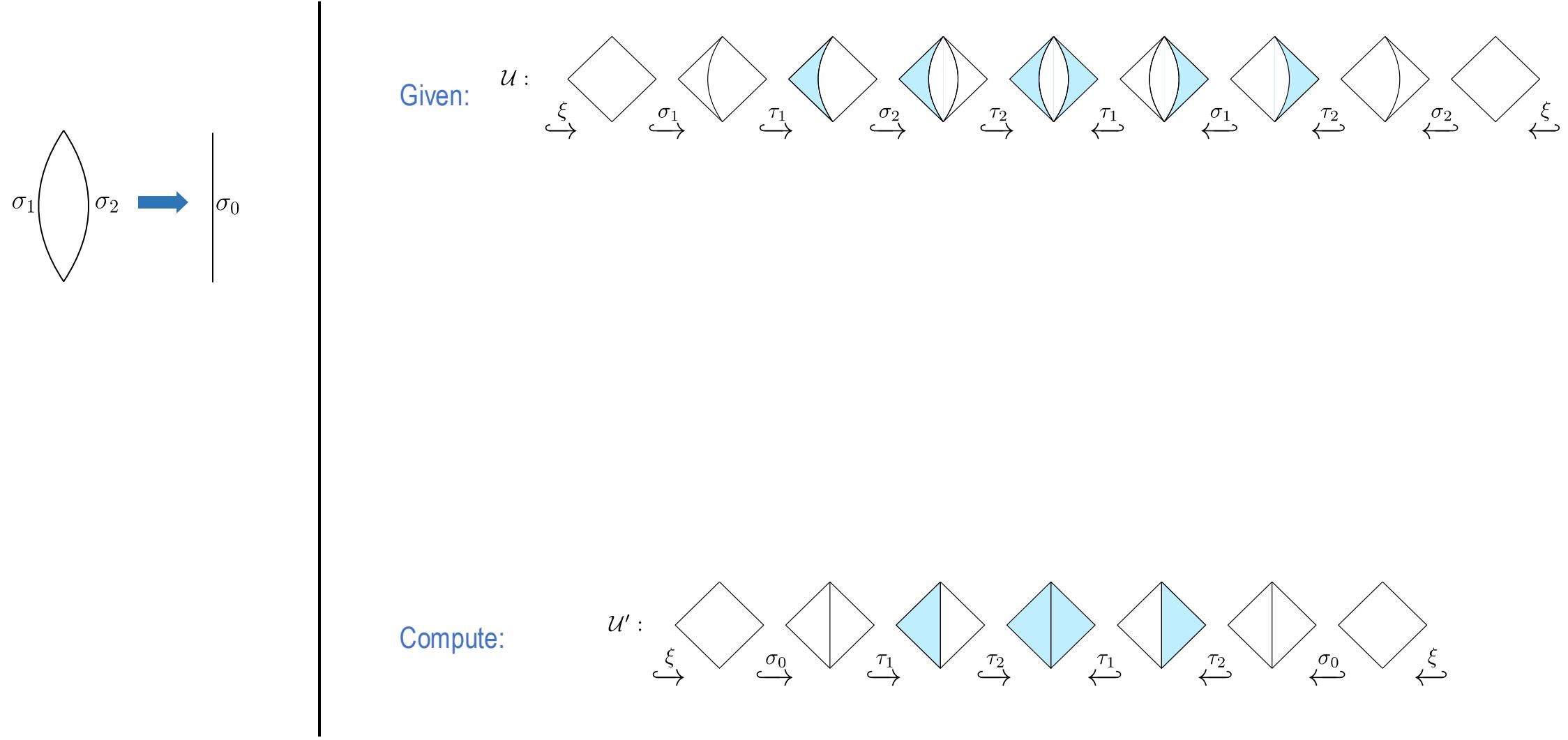
Given:



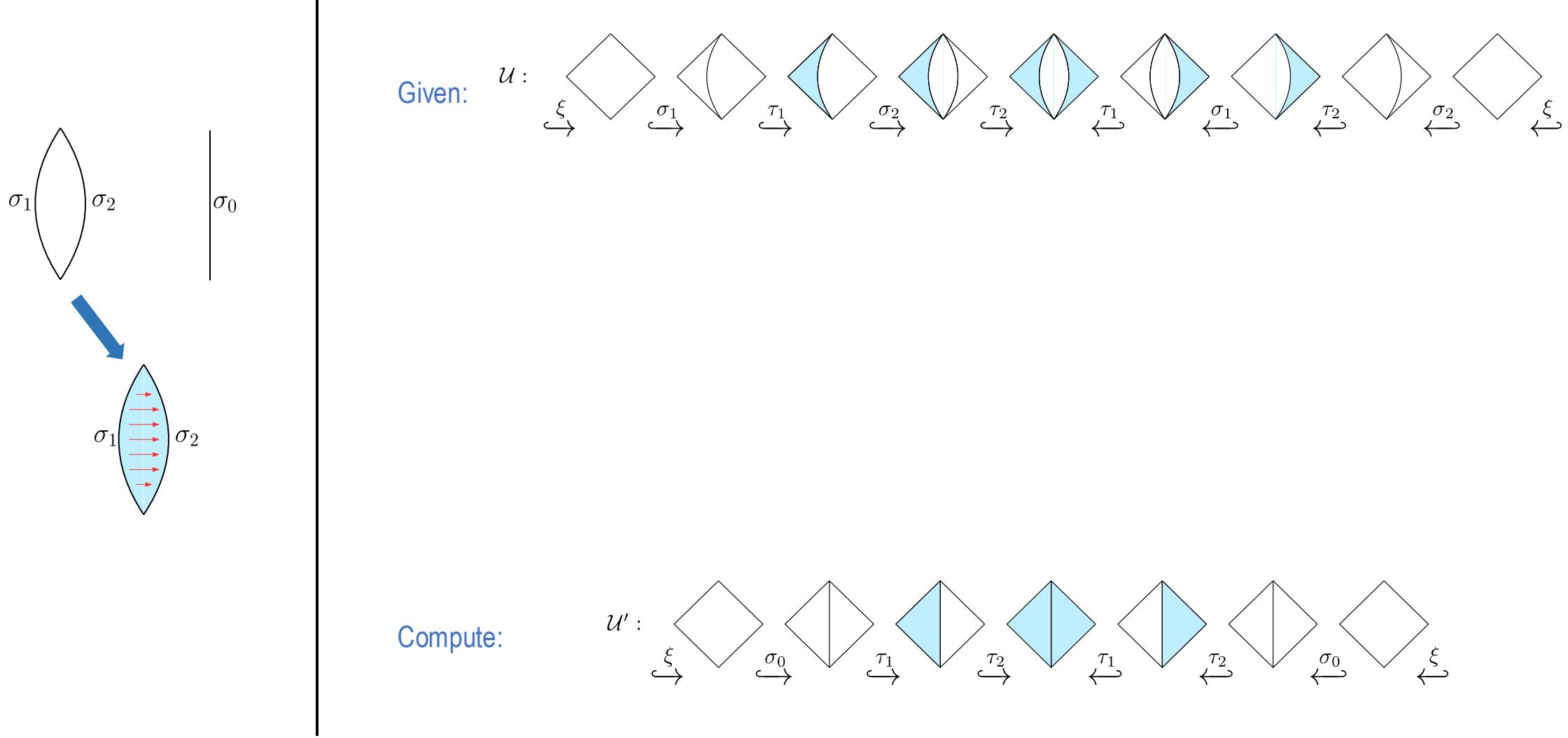
Compute:



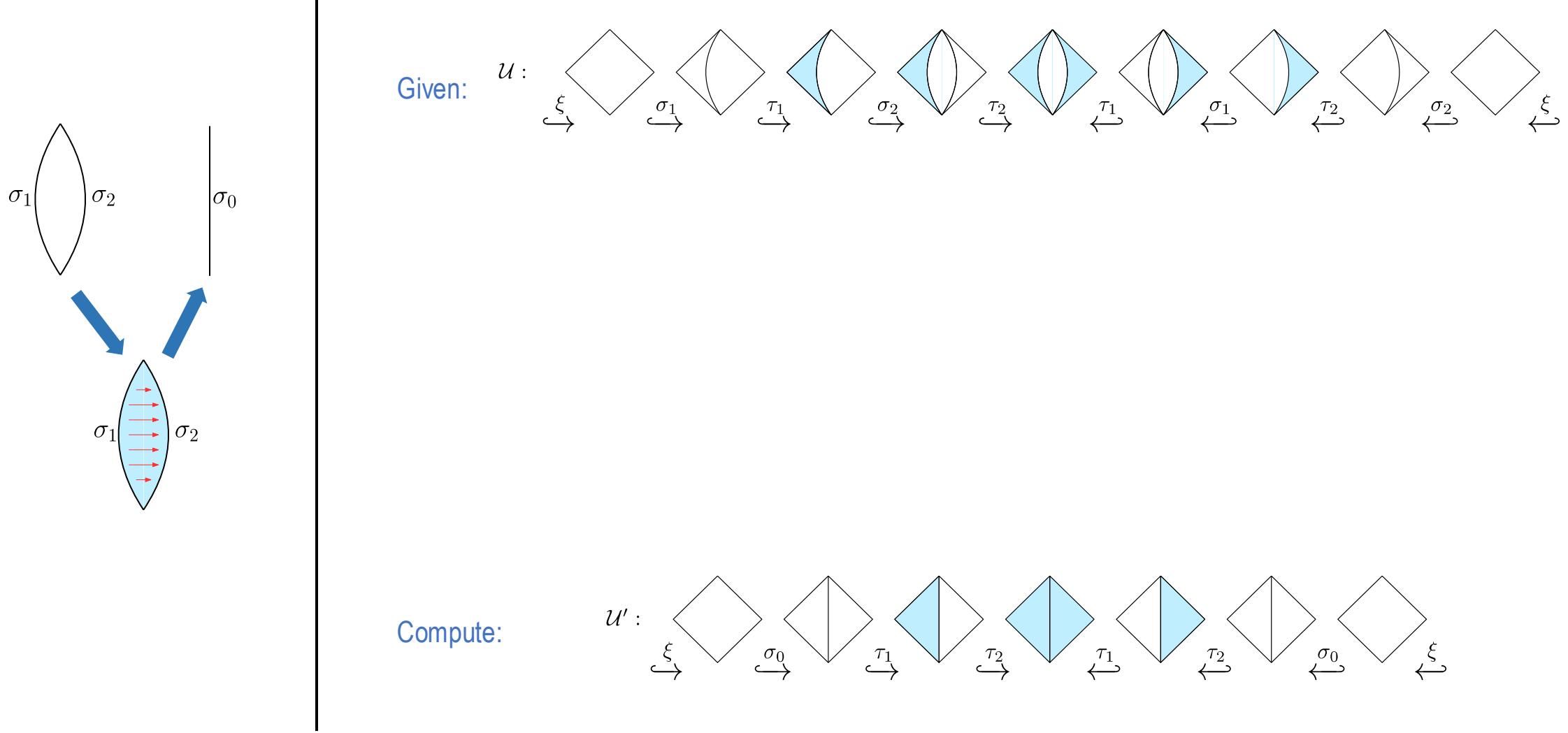
# Solution for outward contraction



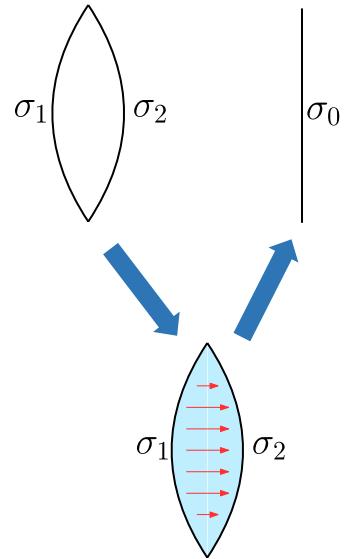
# Solution for outward contraction



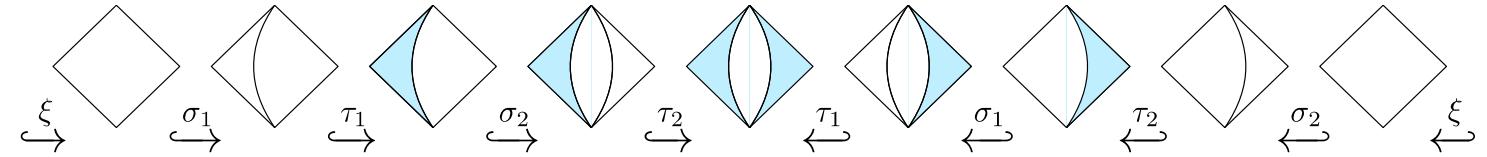
# Solution for outward contraction



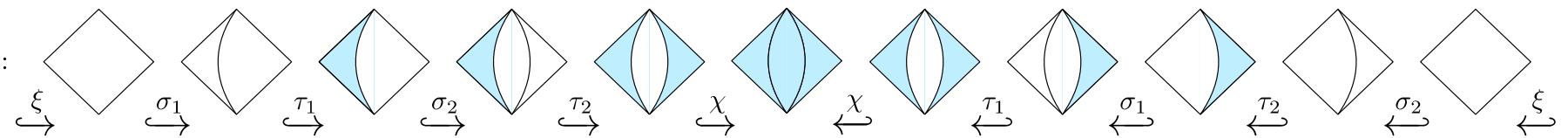
# Solution for outward contraction



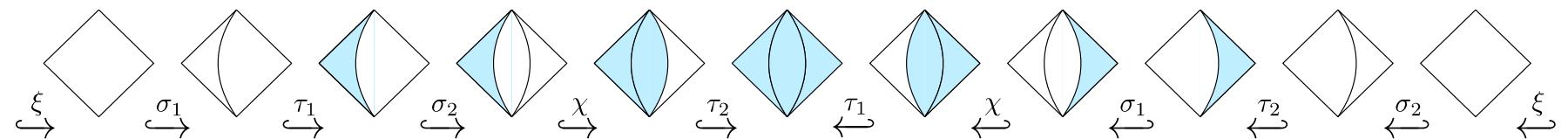
Given:



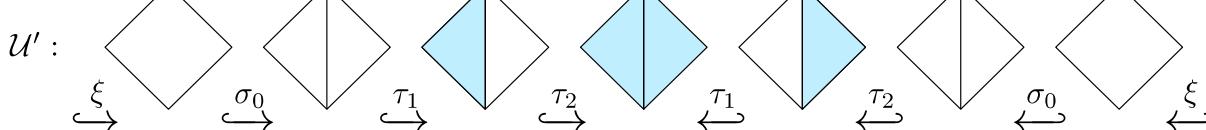
$\mathcal{U}^+ :$



$\tilde{\mathcal{U}} :$



Compute:

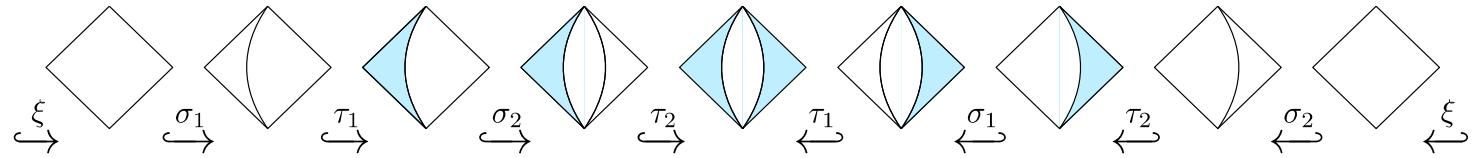


# Solution for outward contraction

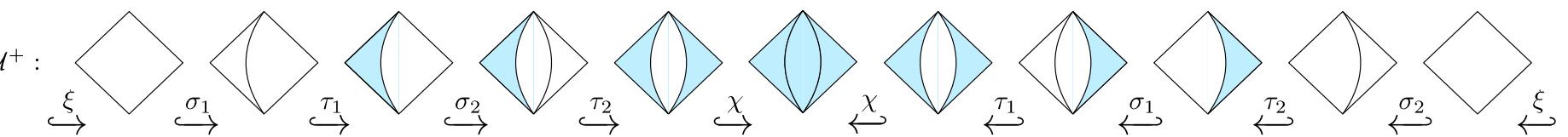
$U \Rightarrow U^+$ :

- Attaching  $\chi$
- $O(m^2)$

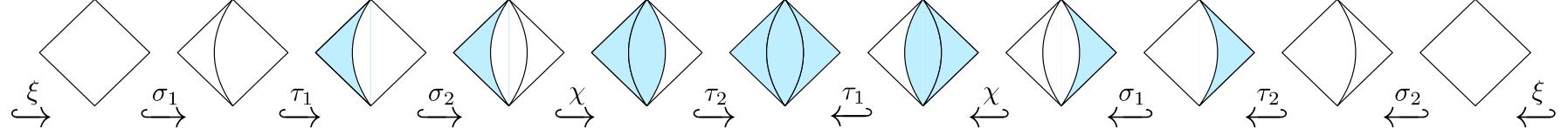
Given:



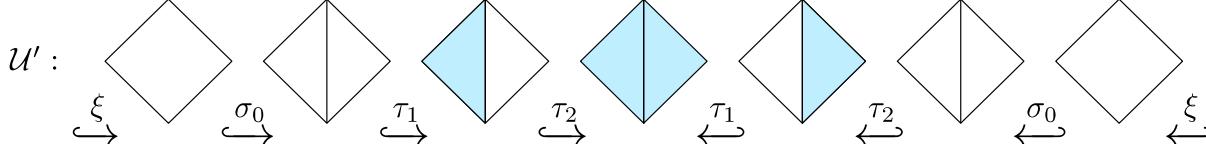
$U^+$ :



$\tilde{U}$ :



Compute:



# Solution for outward contraction

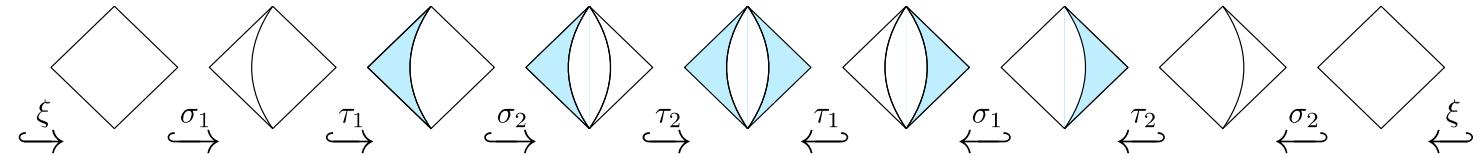
$U \Rightarrow U^+$ :

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- $O(m^2)$

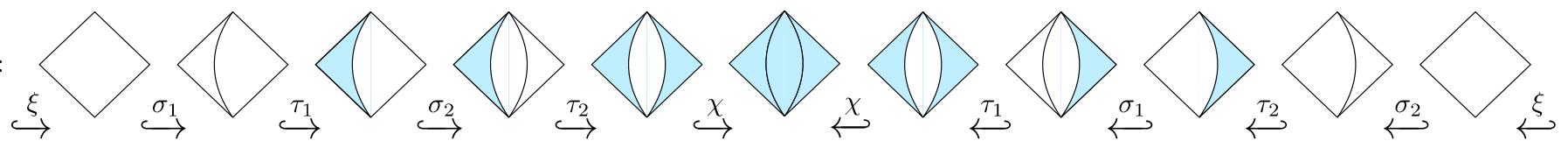
$U^+ \Rightarrow \tilde{U}$ :

- Perform switches
- $O(m^2)$

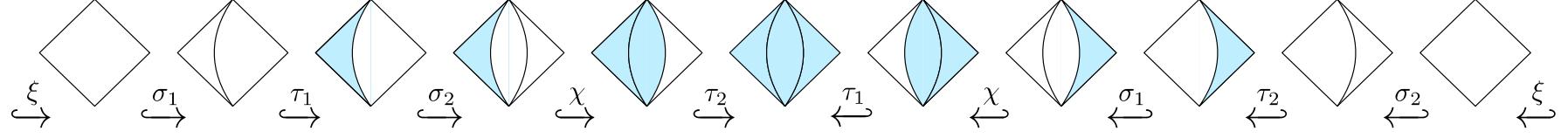
Given:



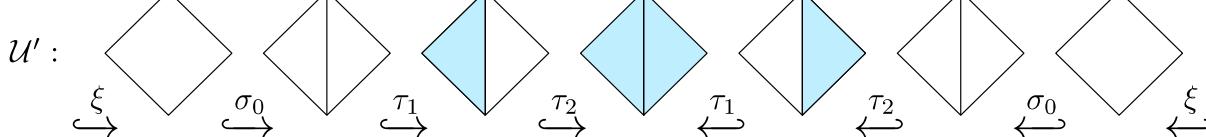
$U^+$ :



$\tilde{U}$ :



Compute:



# Solution for outward contraction

$U \Rightarrow U^+$ :

- Attaching  $\chi$
- $O(m^2)$

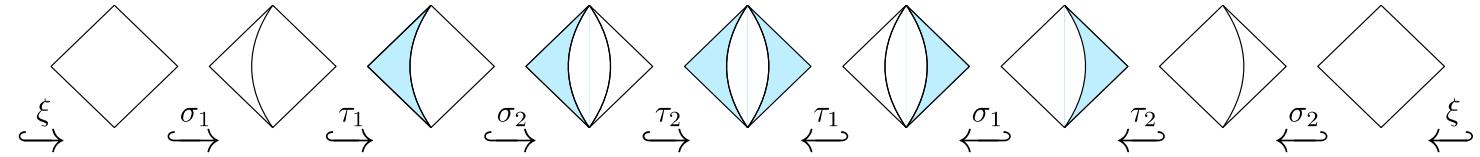
$U^+ \Rightarrow \tilde{U}$ :

- Perform switches
- $O(m^2)$

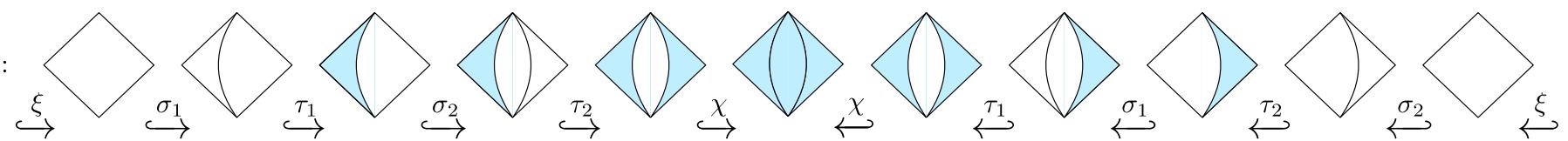
$\tilde{U} \Rightarrow U'$ :

- “Almost” the same
- $O(m)$

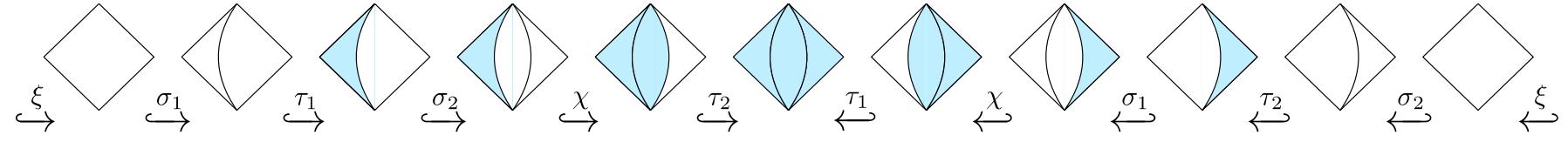
Given:



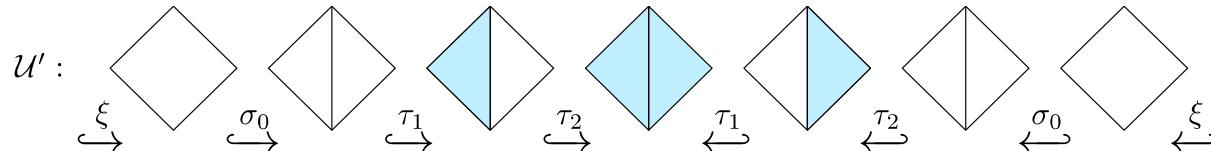
$u^+$ :



$\tilde{u}$ :



Compute:



# Solution for outward contraction

$\tilde{U} \Rightarrow U'$ , formally:

**Proposition.** *Given  $\text{Pers}_*(\tilde{\mathcal{U}})$ , one only needs to do the following to obtain  $\text{Pers}_*(\mathcal{U}')$ : Ignoring the pairs  $(\searrow\sigma_2, \searrow\xi)$  and  $(\nwarrow\xi, \nwarrow\sigma_1)$  in  $\text{Pers}_*(\tilde{\mathcal{U}})$ , for each remaining pair  $(\nwarrow\eta, \nwarrow\gamma) \in \text{Pers}_*(\tilde{\mathcal{U}})$ , produce a corresponding pair  $(\theta(\nwarrow\eta), \theta(\nwarrow\gamma)) \in \text{Pers}_*(\mathcal{U}')$ .*

# Solution for outward contraction

$\tilde{U} \Rightarrow U'$ , formally:

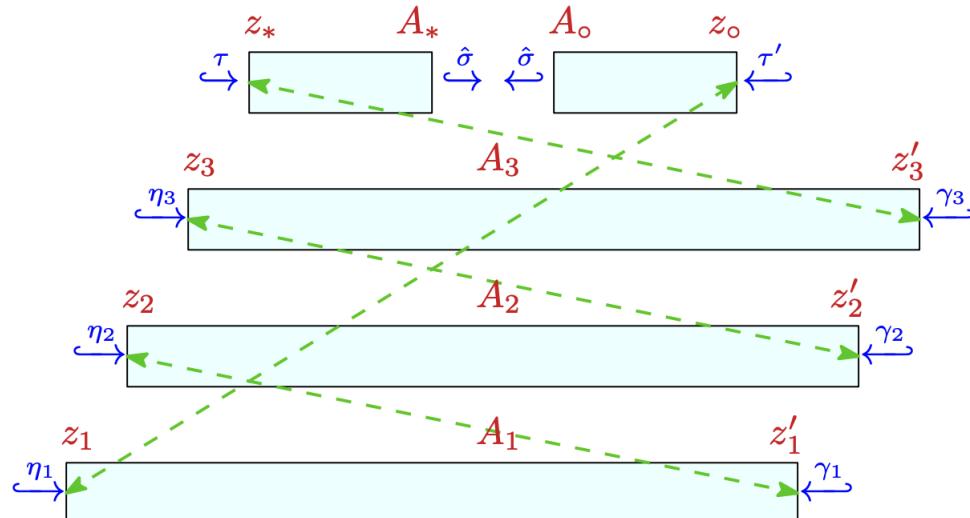
**Proposition.** *Given  $\text{Pers}_*(\tilde{\mathcal{U}})$ , one only needs to do the following to obtain  $\text{Pers}_*(\mathcal{U}')$ : Ignoring the pairs  $(\searrow\sigma_2, \searrow\xi)$  and  $(\nwarrow\xi, \nwarrow\sigma_1)$  in  $\text{Pers}_*(\tilde{\mathcal{U}})$ , for each remaining pair  $(\nwarrow\eta, \nwarrow\gamma) \in \text{Pers}_*(\tilde{\mathcal{U}})$ , produce a corresponding pair  $(\theta(\nwarrow\eta), \theta(\nwarrow\gamma)) \in \text{Pers}_*(\mathcal{U}')$ .*

Conclusion:

**Theorem.** *The barcodes for an outward contraction on zigzag filtrations can be updated in  $O(m^2)$  time, matching the complexity for a contraction on the standard filtrations.*

# Inward contraction

- While inward contraction is easy by converting to non-zigzag, it becomes non-trivial when converting to up-down
- Algorithm idea:
  - After some preprocessing, we are left with certain intervals which are not ‘settled’ (contains the cell being removed)
  - These intervals follow a fixed pattern, and we utilized an ‘alternative relinking’ to produce intervals for the new filtration



# Updating zigzag persistence on graphs over switches

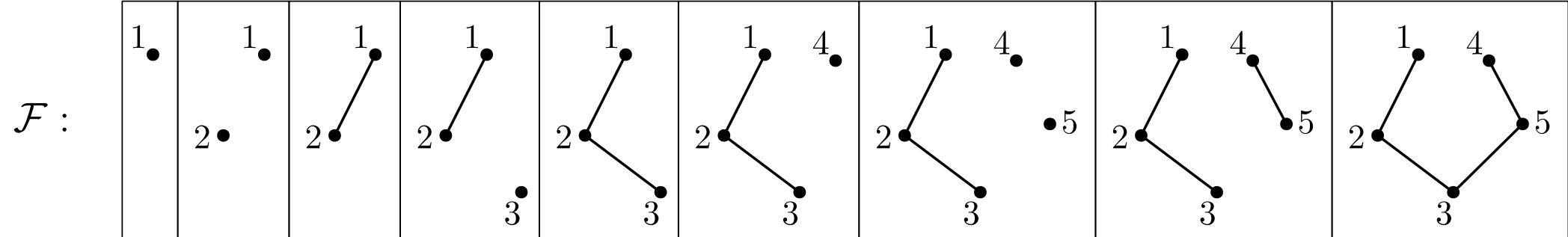
# Update for non-zigzag graph filtrations

- Propose  $O(\log m)$  algorithms for updating non-zigzag filtrations on graphs over switches

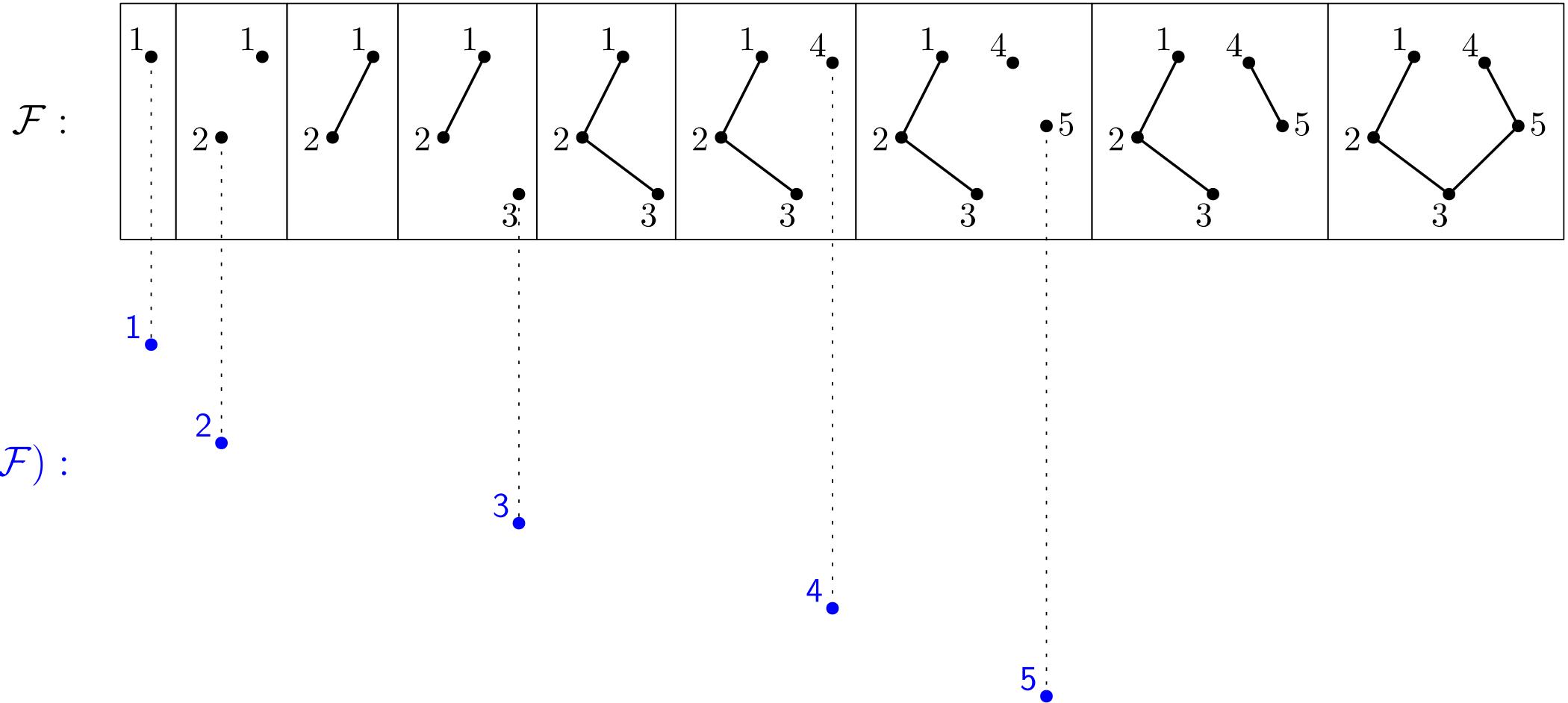
$$\mathcal{F} : \emptyset = G_0 \hookrightarrow \dots \hookrightarrow G_{i-1} \xhookrightarrow{\sigma} G_i \xhookrightarrow{\tau} G_{i+1} \hookrightarrow \dots \hookrightarrow G_m$$
$$\mathcal{F}' : \emptyset = G_0 \hookrightarrow \dots \hookrightarrow G_{i-1} \xhookrightarrow{\tau} G'_i \xhookrightarrow{\sigma} G_{i+1} \hookrightarrow \dots \hookrightarrow G_m$$

- Maintain merge forest (trees) encoding all info in the pers module
- Case analysis: Perform the update in difference cases
- Use two dynamic trees data structure (DFT tree, Link-Cut tree) to achieve the complexity

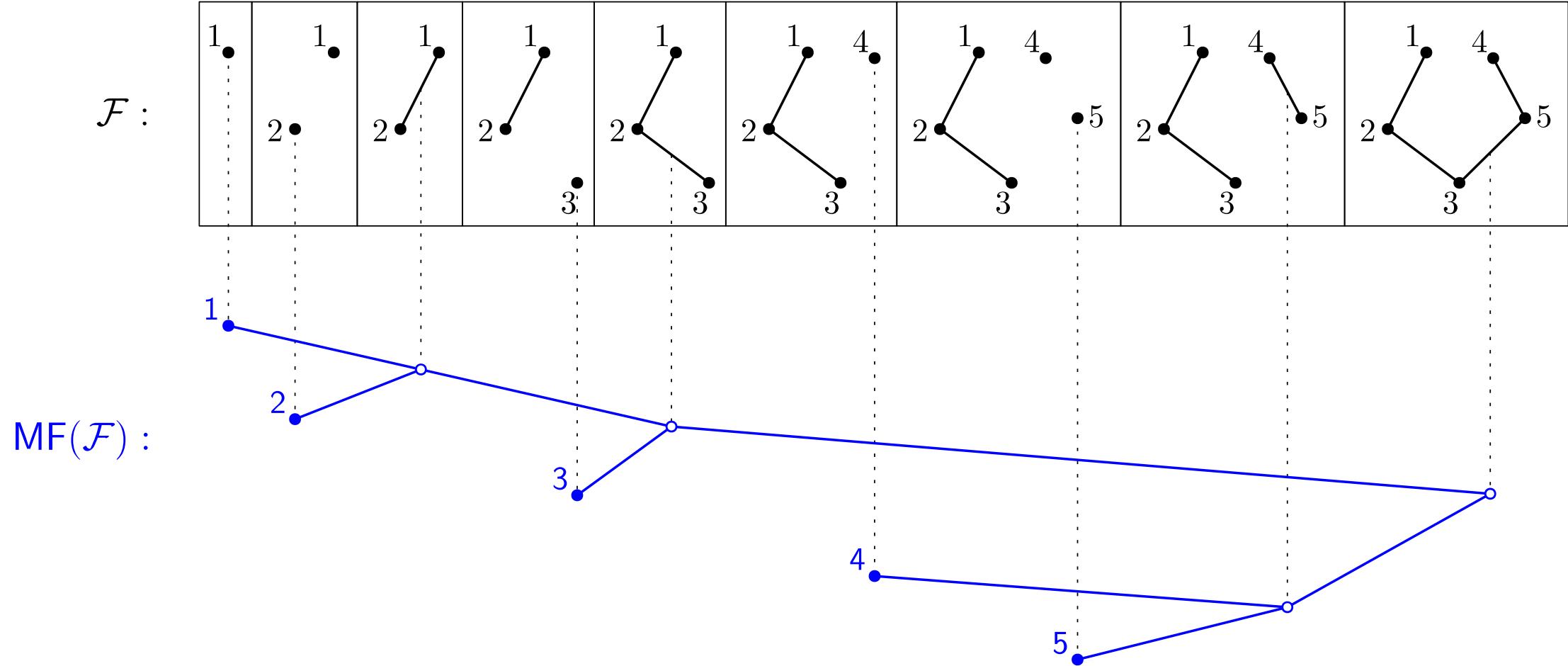
# Merge forest (tree)



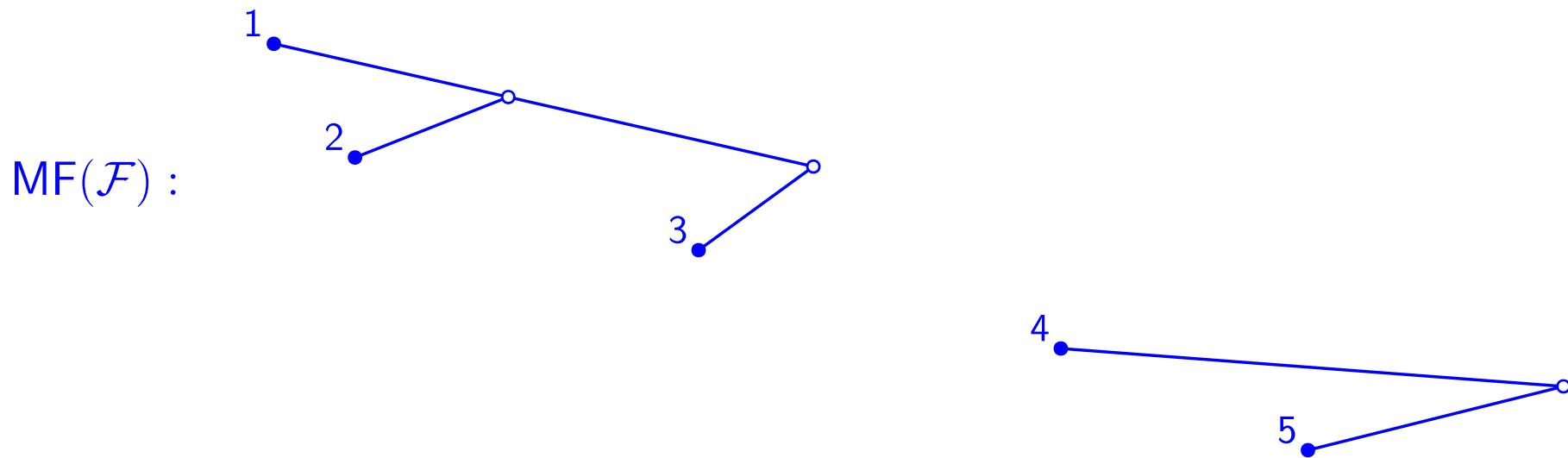
# Merge forest (tree)



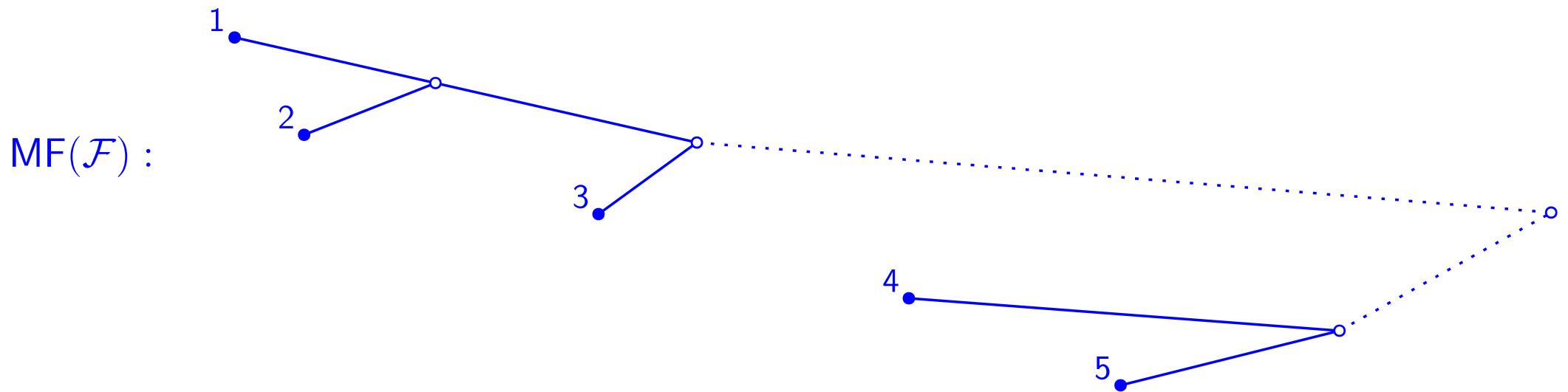
# Merge forest (tree)



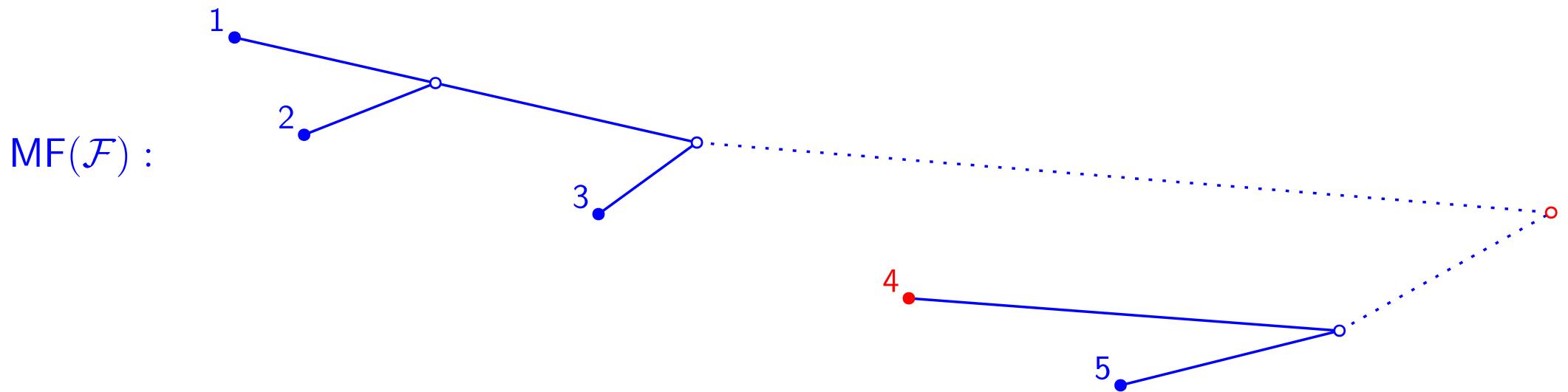
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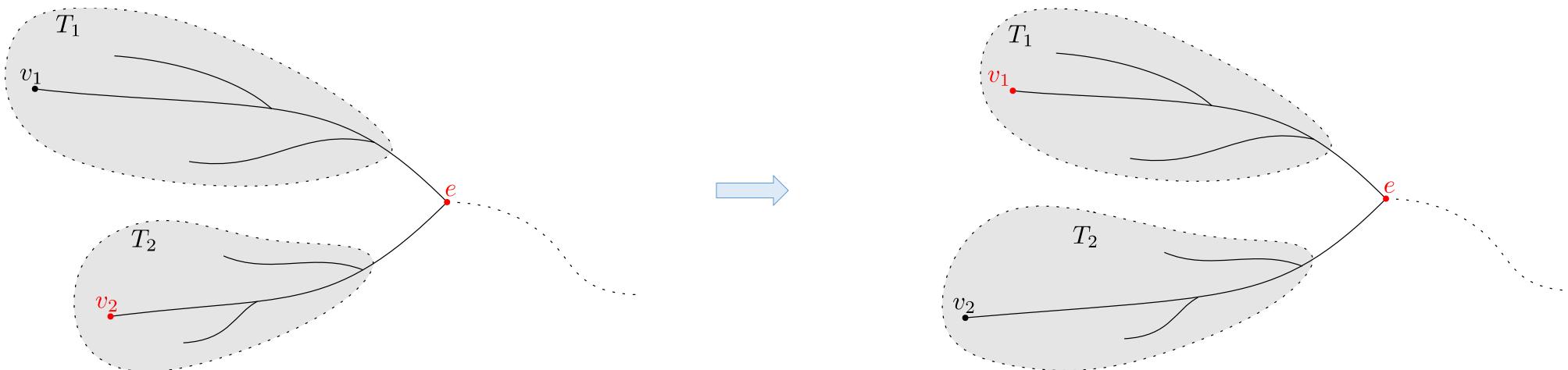


# Merge forest (tree)



# 1. Switch two vertices $v_1, v_2$

- The only situation where the pairing changes:
  - $v_1, v_2$  are in the same tree in the merge forest
  - $v_1, v_2$  are both unpaired when  $e$  is added in  $\mathcal{F}$ , where  $e$  is the edge corresponding to the *nearest common ancestor* of  $v_1, v_2$  in the merge forest
- In above case, we switch the paired edges of  $v_1, v_2$



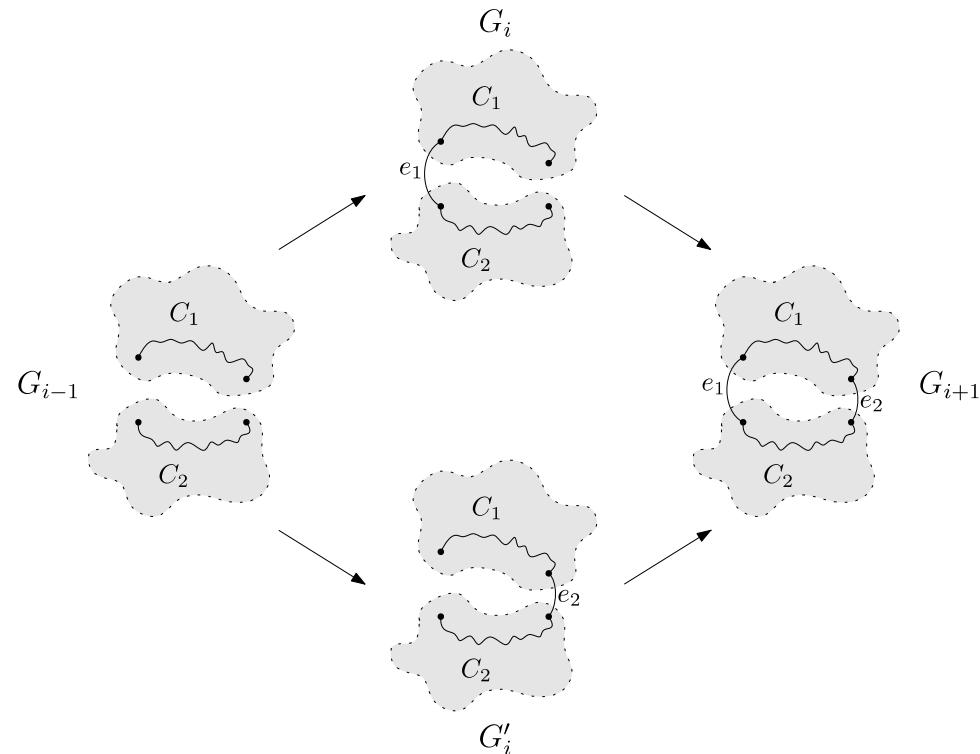
# More definitions

Two types of edges in the graph filtration:

- **Negative edge**: connect two different connected components
- **Positive edge**: connect the same connected component

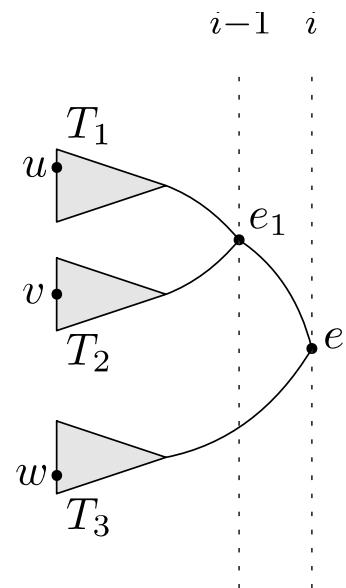
## 2. Switch a negative edge $e_1$ and a positive edge $e_2$

- If  $e_1$  is in a 1-cycle after is  $e_2$  added:
  - This is the case where  $e_1, e_2$  connect to the same two connected components
  - After the switch,  $e_1$  becomes positive and  $e_2$  becomes negative
  - We pair  $e_2$  with the vertex that  $e_1$  previously pairs with
  - The node in the merge forest corresponding to  $e_1$  should now correspond to  $e_2$



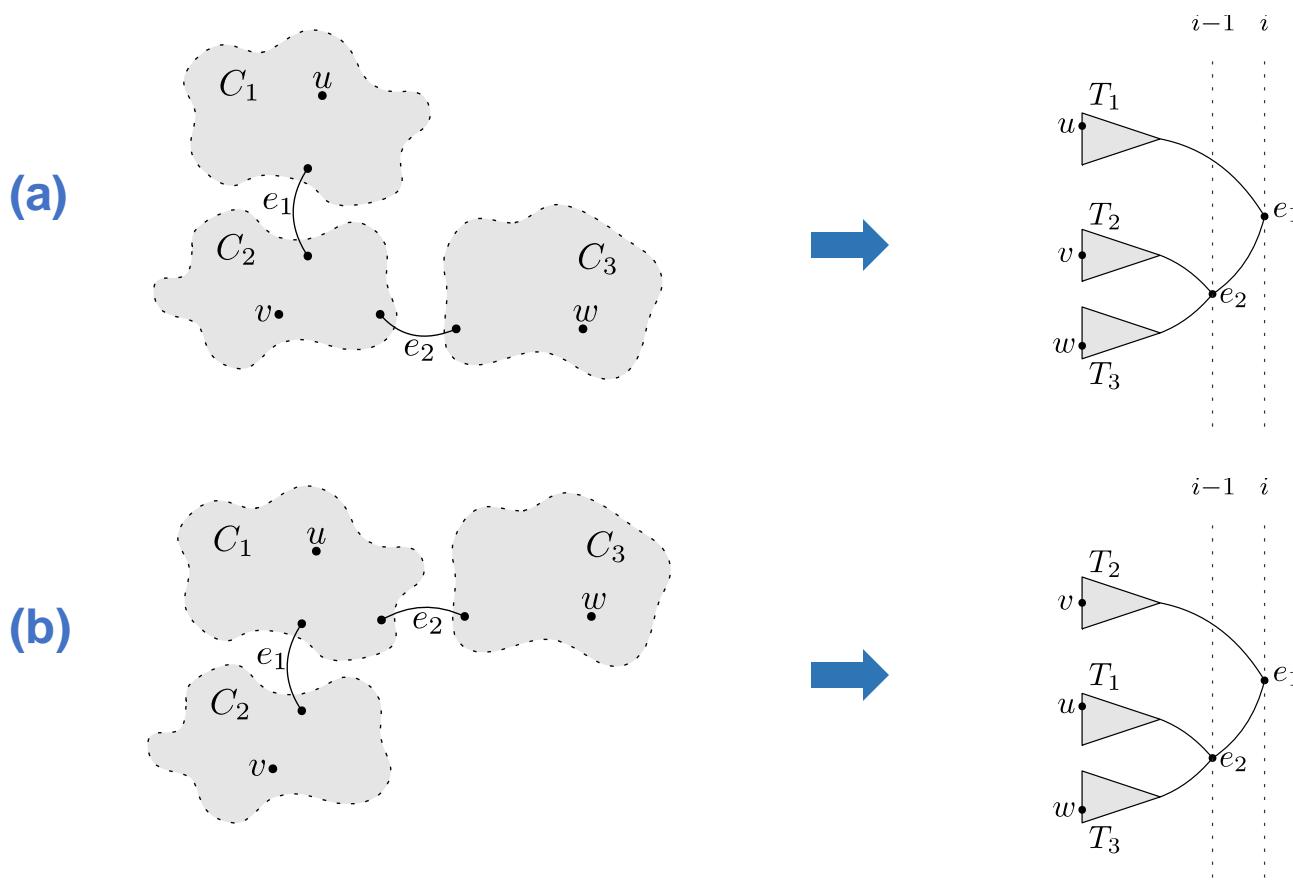
### 3. Switch two negative edges $e_1, e_2$

- Only need to make changes when the corresponding node of  $e_1$  is a child of the corresponding node of  $e_2$  in the merge forest
- Let  $u, v, w$  be the lowest leaves in  $T_1, T_2, T_3$ .
- WLOG, assume  $v$  is lower than  $u$ .



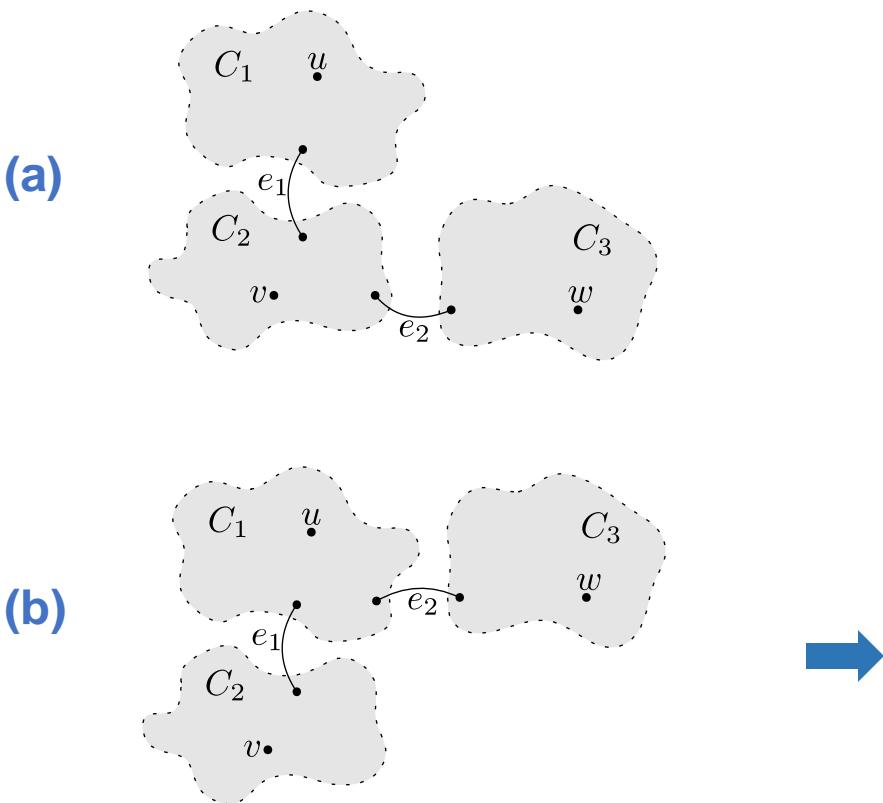
### 3. Switch two negative edges $e_1, e_2$

- Based on the structure of the merge forest, there are two connecting configurations for  $C_1, C_2, C_3$  in  $G_{i-1}$  ( $C_1, C_2, C_3$  are the connected components containing  $u, v, w$  respectively)



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If  $w$  is lower than  $u$ , then swap the paired vertices of  $e_1, e_2$

- Provable by case analysis

# DFT-Tree

Data structures implementing the merge forests [Farina, Laura, 2015]:

- $\text{ROOT}(v)$ : Returns the root of the tree containing node  $v$ .
- $\text{CUT}(v)$ : Deletes the edge connecting node  $v$  to its parent.
- $\text{LINK}(u, v)$ : Makes the root of the tree containing node  $v$  be a child of node  $u$ .
- $\text{NCA}(u, v)$ : Returns the nearest common ancestor of two nodes  $u, v$  in the same tree.
- $\text{CHANGE-VAL}(v, x)$ : Assigns the value associated to a leaf  $v$  to be  $x$ .
- $\text{SUBTREE-MIN}(v)$ : Returns the leaf with the minimum associated value in the subtree rooted at  $v$ .

>Returns the lowest leaf for a subtree

# Detecting if $e_1 = (u, v)$ is in a cycle in $G_{i+1}$

- Check if  $u, v$  are connected in  $G'_i$

$$\mathcal{F}' : \emptyset = G_0 \hookrightarrow \cdots \hookrightarrow G_{i-1} \hookrightarrow G'_i \xhookrightarrow{e_1} G_{i+1} \hookrightarrow \cdots \hookrightarrow G_m$$

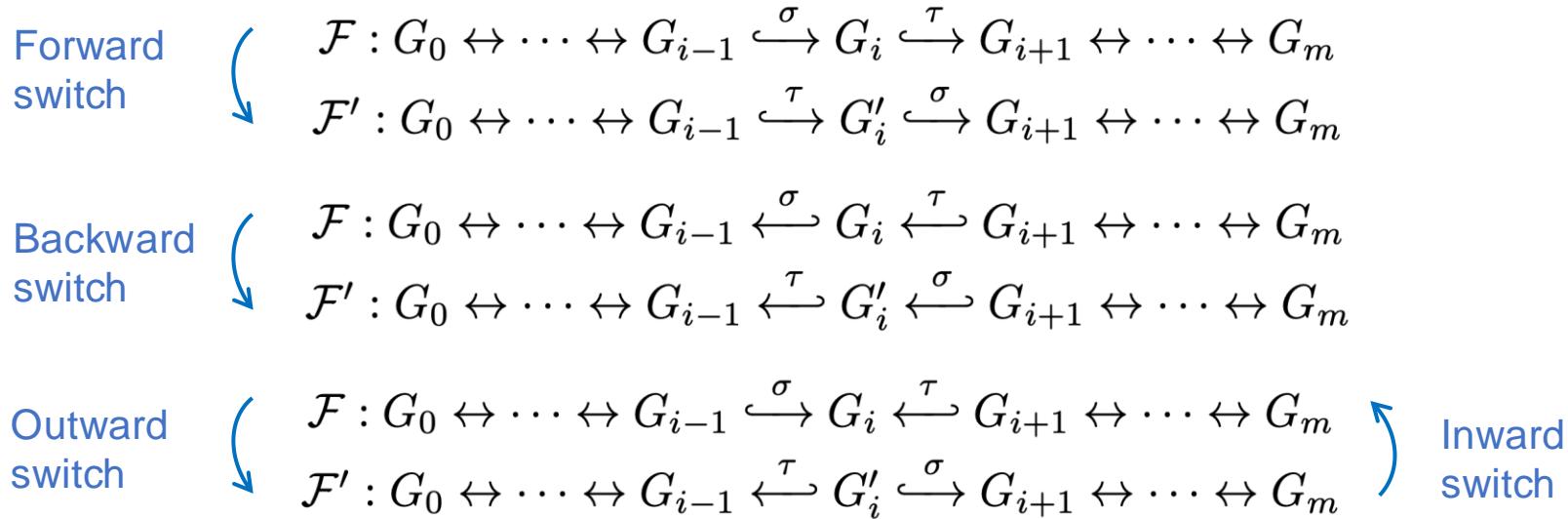
- Check the first time  $u, v$  are connected in the filtration  $\mathcal{F}'$
- Based on an idea in [DH21], do following:
  - Let edges in  $G := G_m$  be weighted by their indices in  $\mathcal{F}'$
  - The first time  $u, v$  are connected =  $1 +$  the **bottleneck weight** of the path in the MSF of  $G$  (**bottleneck weight**: max weight of edges)
- Maintain the MSF over the switch by the Link-Cut tree [ST81]:
  - Everything can be in  $O(\log m)$  time
  - This is possible because we are only doing switches (switching the weights for edges whose weights are consecutive)

- Tamal K. Dey and Tao Hou. Computing zigzag persistence on graphs in near-linear time. 2021.

- Daniel D. Sleator and Robert Endre Tarjan. A data structure for dynamic trees. 1981.

# Update for zigzag graph filtrations

- Four switch operations:



- Strategy: Convert the zigzag filtrations to up-down filtrations as previous
- Immediately, inward and outward switches take  $O(1)$  time
- Forward and backward switches: For intervals other than those from the edge-edge pairs, the update reduces to the standard persistence case, hence  $O(\log m)$  time

# $O(m)$ algorithm for updating edge-edge pairs

- Based on a direct maintenance of **representative cycles** for pairs

**Algorithm 1.** We describe the algorithm for the forward switch and the procedure for a backward switch is symmetric. Let  $\Pi$  be the set of edge-edge pairs initially for  $\mathcal{U}$ . Since a switch containing a vertex makes no changes to the edge-edge pairs, suppose that the switch is an edge-edge switch and let  $e_1, e_2$  be the two switched edges. Also, let  $\mathcal{U}_u$  be the ascending part of  $\mathcal{U}$ . We have the following cases:

- A.  $e_1$  and  $e_2$  are both negative in  $\mathcal{U}_u$ :** Do nothing.
- B.  $e_1$  is positive and  $e_2$  is negative in  $\mathcal{U}_u$ :** Do nothing.
- C.  $e_1$  is negative and  $e_2$  is positive in  $\mathcal{U}_u$ :** Let  $z$  be the representative cycle for the pair  $(e_2, \epsilon) \in \Pi$ . If  $e_1 \in z$ , pair  $e_1$  with  $\epsilon$  in  $\Pi$  with the same representative  $z$  (notice that  $e_2$  becomes unpaired).
- D.  $e_1$  and  $e_2$  are both positive in  $\mathcal{U}_u$ :** Let  $z, z'$  be the representative cycles for the pairs  $(e_1, \epsilon), (e_2, \epsilon') \in \Pi$  respectively. Do the following according to different cases:
  - If  $e_1 \in z'$  and the deletion of  $\epsilon'$  is before the deletion of  $\epsilon$  in  $\mathcal{U}$ : Let the representative for  $(e_2, \epsilon')$  be  $z + z'$ . The pairing does not change.
  - If  $e_1 \in z'$  and the deletion of  $\epsilon'$  is after the deletion of  $\epsilon$  in  $\mathcal{U}$ : Pair  $e_1$  and  $\epsilon'$  in  $\Pi$  with the representative  $z'$ ; pair  $e_2$  and  $\epsilon$  in  $\Pi$  with the representative  $z + z'$ .

# $O(\sqrt{m} \log m)$ algorithm: ideas

- Eliminate the explicit maintenance of representative cycles by observing:
  - We only need to check the connectivity of two vertices in the [intersection of two graphs](#) in the up-down
  - One graph is from the [ascending](#) part, the other is from the [descending](#) part
- Maintain the MSF's for  $\sqrt{m}$  [graphs](#) in the [ascending](#) part where the edges are weighted by indices in the [descending](#) part.
- Each MSF is a Link-Cut tree

# Computing zigzag representatives in $O(m^2n)$ time

[Dey-H-Morozov] A fast Algorithm for computing zigzag representatives. SODA25 (to appear)

# Computing representatives for persistence

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**Definition 4** (Representative). Let  $[b, d] \subseteq \{1, \dots, m - 1\}$  be an interval. A  $p$ -th representative sequence (also simply called  $p$ -th representative) for  $[b, d]$  consists of a sequence of  $p$ -cycles  $\{z_i \in Z_p(K_i) \mid b \leq i \leq d\}$  and a sequence of  $(p + 1)$ -chains  $\{c_i \mid b - 1 \leq i \leq d\}$ , typically denoted as

$$c_{b-1} \leftarrow z_b \xrightarrow{c_b} \dots \xrightarrow{c_{d-1}} z_d \dashrightarrow c_d,$$

such that for each  $i$  with  $b \leq i < d$ :

- if  $K_i \hookrightarrow K_{i+1}$  is forward, then  $c_i \in C_{p+1}(K_{i+1})$  and  $z_i + z_{i+1} = \partial(c_i)$  in  $K_{i+1}$ ;
- if  $K_i \hookleftarrow K_{i+1}$  is backward, then  $c_i \in C_{p+1}(K_i)$  and  $z_i + z_{i+1} = \partial(c_i)$  in  $K_i$ .

Furthermore, the sequence satisfies the additional conditions:

**Birth condition:** If  $K_{b-1} \xleftarrow{\sigma_{b-1}} K_b$  is backward, then  $z_b = \partial(c_{b-1})$  for  $c_{b-1}$  a  $(p + 1)$ -chain in  $K_{b-1}$  containing  $\sigma_{b-1}$ ; if  $K_{b-1} \xleftarrow{\sigma_{b-1}} K_b$  is forward, then  $\sigma_{b-1} \in z_b$  and  $c_{b-1}$  is undefined.

**Death condition:** If  $K_d \xleftarrow{\sigma_d} K_{d+1}$  is forward, then  $z_d = \partial(c_d)$  for  $c_d$  a  $(p + 1)$ -chain in  $K_{d+1}$  containing  $\sigma_d$ ; if  $K_d \xleftarrow{\sigma_d} K_{d+1}$  is backward, then  $\sigma_d \in z_d$  and  $c_d$  is undefined.

- $O(m^2 n^2)$  time by directly adapt the algorithm in [MO15]

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- $O(m^2 n^2)$  time by directly adapt the algorithm in [MO15]
- Find a way to compress the representatives to achieve the  $O(m^2 n)$  complexity

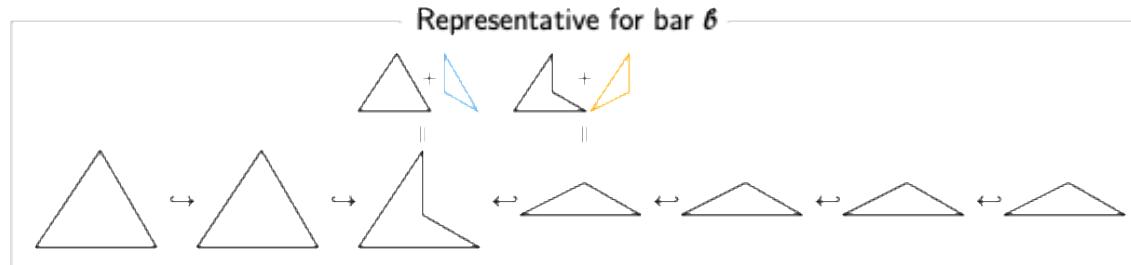
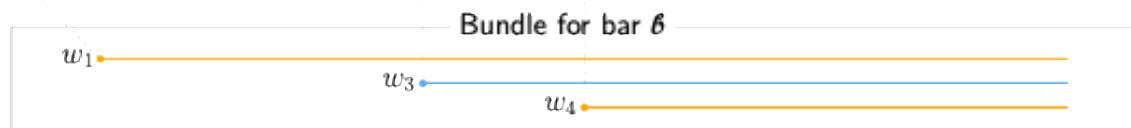
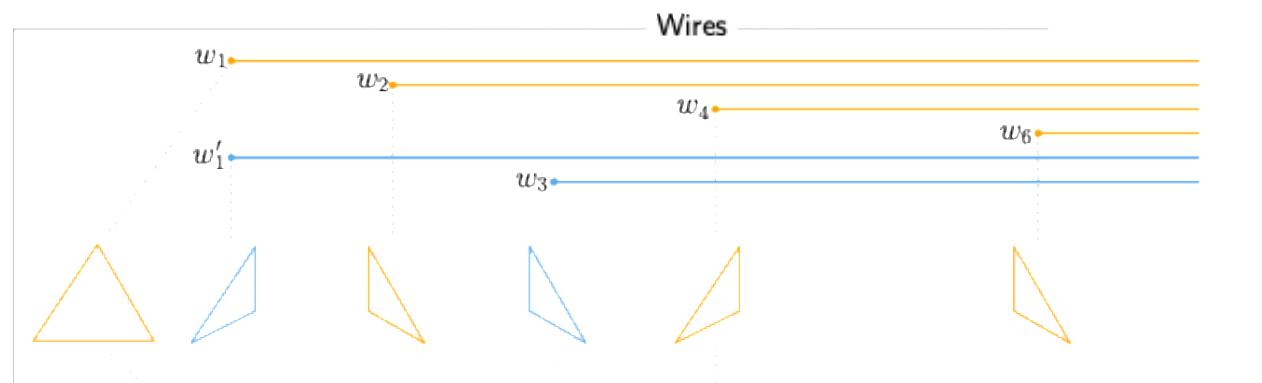
# Key to bringing down the complexity

- How to store a representative for an interval in memory:
  - The straightforward method takes  $O(mn)$  space, so that summing two representatives takes  $O(mn)$  time, and hence the  $O(m^2n^2)$  complexity

# Key to bringing down the complexity

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  - The straightforward method takes  $O(mn)$  space, so that summing two representatives takes  $O(mn)$  time, and hence the  $O(m^2n^2)$  complexity
- We find a *compressed* way to store a representative
  - The compressed method takes  $O(m)$  space, so that summing two representatives takes  $O(m)$  time, and hence the  $O(m^2n)$  complexity
  - This is by storing representatives as a set of *wires*, each a cycle born at a certain time and extending indefinitely

# An example for storing a rep. as wires



To Answer the Question in the Title

# Can zigzag persistence be computed as efficiently as the standard version?

Problems		Wall-clock time	Complexity
Compute persistence	General		Yes!
	Graph	Not far away	$O(m \alpha(m))$ vs $O(m \log m)$
Update	General	?	Yes
	Graph	No	$O(\log m)$ vs $O(\sqrt{m} \log m)$
Compute representatives		Still a gap	$O(m^\omega)$ vs $O(m^2 n)$

Thank you!