

**MATH412/COMPSCI434/MATH713**  
**Fall 2025**

***Topological Data Analysis***

**Topic 3: Simplicial Homology - Part 3**

**Topic 4: Introduction to Persistent Homology - Part 1**

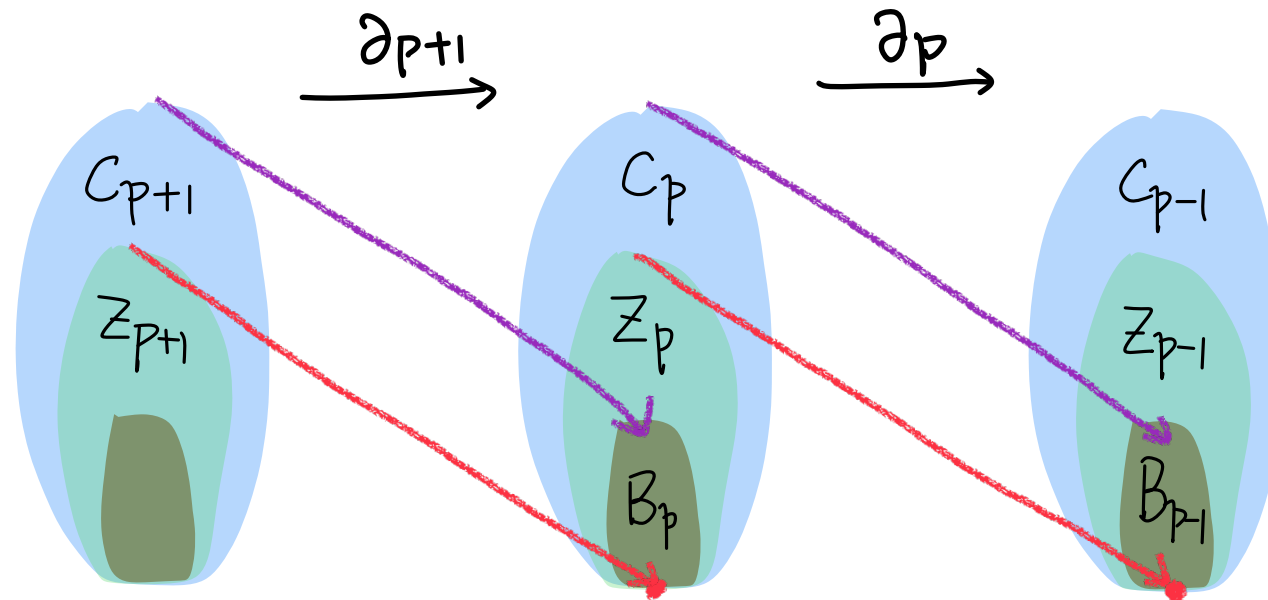
Instructor: Ling Zhou

boundaries  $\subseteq$  cycles  $\subseteq$  chains

# Review: Cycles and Boundaries

Under field (e.g.  $\mathbb{Z}_2$ ) coefficients,  $B_p$ ,  $Z_p$ ,  $C_p$  are all vector spaces.

$$\partial_p \circ \partial_{p+1} = 0 \Rightarrow B_p \subseteq Z_p \subseteq C_p$$



# Review: Homology groups

- ▶  $p$ -th **homology group** is  $H_p(K) = Z_p / B_p$ 
  - ▶  $c_1$  is **homologous to**  $c_2$  if  $c_1 - c_2 \in B_p$ , i.e,  $c_1 - c_2$  is a boundary cycle
  - ▶ The **homology class**  $[c]$  represents the family  $p$ -cycles homologous to  $c$
- ▶ **Betti number**:  $\beta_p(K) = \dim H_p(K) = \dim Z_p - \dim B_p$

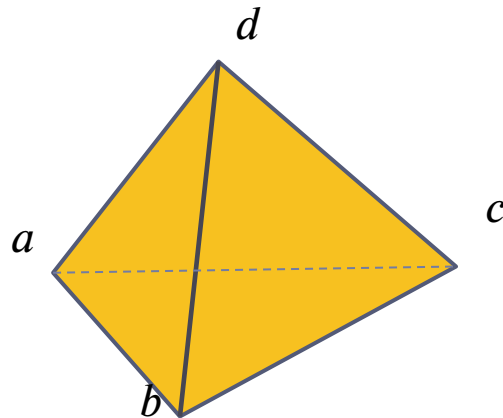
▶ Theorem: Two homotopy equivalent topological spaces have isomorphic homology groups and thus same Betti numbers.

$$\dots \xrightarrow{\partial_{p+2}} C_{p+1} \xrightarrow{\partial_{p+1}} C_p \xrightarrow{\partial_p} C_{p-1} \xrightarrow{\partial_{p-1}} \dots$$

# Matrix view and computation

# Boundary Matrix

- ▶  $K^p = \{ \alpha_1, \dots, \alpha_{n_p} \}$ ,  $K^{p-1} = \{ \tau_1, \dots, \tau_{n_{p-1}} \}$ 
  - ▶  $K^p$  forms a basis for p-th chain group  $C_p$
- ▶ For simplicity, we **use  $\mathbb{Z}_2$  coefficients**, and define the  $n_{p-1} \times n_p$  **boundary matrix**  $A_p$ 
  - ▶  $A_p[i][j] = 1$  iff  $\tau_i \subseteq \sigma_j$
  - ▶ representing  $\partial_p: C_p \rightarrow C_{p-1}$  w.r.t. ordered bases  $\{ \alpha_1, \dots, \alpha_{n_p} \}$  and  $\{ \tau_1, \dots, \tau_{n_{p-1}} \}$



$$A_2 = \begin{matrix} & \begin{matrix} abc & abd & acd & bcd \end{matrix} \\ \begin{matrix} ab \\ ac \\ ad \\ bc \\ bd \\ cd \end{matrix} & \begin{pmatrix} 1 & 1 & & \\ 1 & & 1 & \\ & 1 & 1 & \\ 1 & & & 1 \\ & 1 & & 1 \\ & & 1 & 1 \end{pmatrix} \end{matrix}$$

# Boundary Matrix

- ▶ Vector representation of a  $p$ -chain  $c = \sum_{i=1}^{n_p} c_i \alpha_i$ :
  - ▶ Under basis  $K^p$ , vector representation of  $c$  is  $\vec{c} = [c_1, c_2, \dots, c_{n_p}]^T$
- ▶ Boundary  $\partial_p c$  is a  $(p - 1)$ -chain with vector representation  $A_p \vec{c}$  w.r.t basis  $K^{p-1}$

$$A_p \vec{c} = \begin{bmatrix} a_1^1 & a_1^2 & \dots & a_1^{n_p} \\ a_2^1 & a_2^2 & \dots & a_2^{n_p} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n_p-1}^1 & a_{n_p-1}^2 & \dots & a_{n_p-1}^{n_p} \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_{n_p} \end{bmatrix}$$

# Observations

- ▶ To compute the cycle space  $Z_p = \ker \partial_p$ ,
  - ▶ solve the equation  $A_p c = 0$  to find a basis for the kernel
- ▶ To compute the boundary space  $B_p = \text{Im} \partial_{p+1}$ 
  - ▶ compute the column space of  $A_{p+1}$  and find a basis for the image
- ▶ We can do both on  $A_p$  through **column reduction**
  - ↳ { null space  
column space

# Matrix reduction

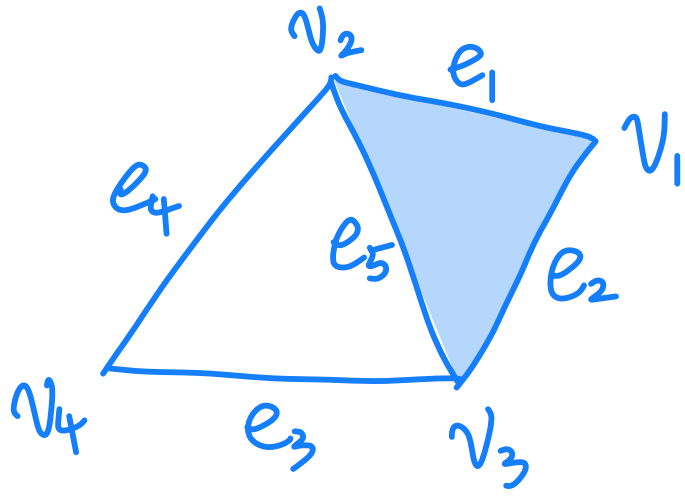
- ▶ Turn  $A_p$  into the **column reduced form**
  - ▶ Each non-zero column has a unique lowID/pivot: index of lowest 1-entry
- ▶ **Column operations** in Gaussian elimination:
  - ▶ scaling (not needed over  $\mathbb{Z}_2$ )
  - ▶ swap (not necessary)
  - ▶ add one column to another
- ▶ Do **left-to-right column reduction** and get bases of
  - ▶  $B_{p-1} = \text{Im } \partial_p$ : the reduced columns
  - ▶  $Z_p = \ker \partial_p$ : the column operations

$$\begin{bmatrix} * & * & * & 0 \\ * & 1 & * & 0 \\ 1 & 0 & * & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

Column reduced form

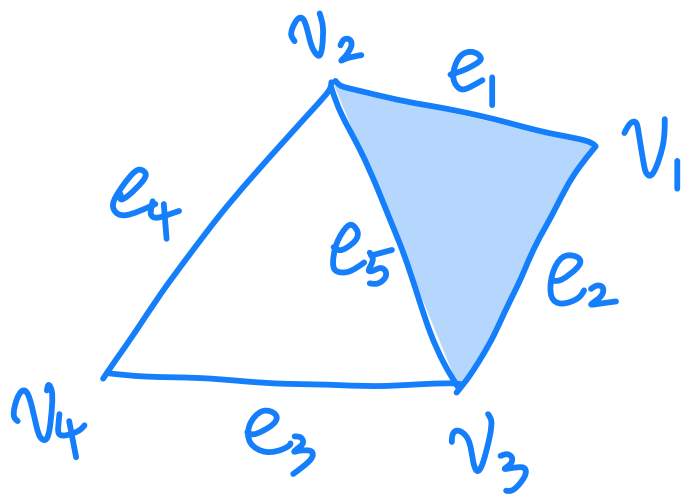
$$\text{low}[i] \neq \text{low}[j]$$





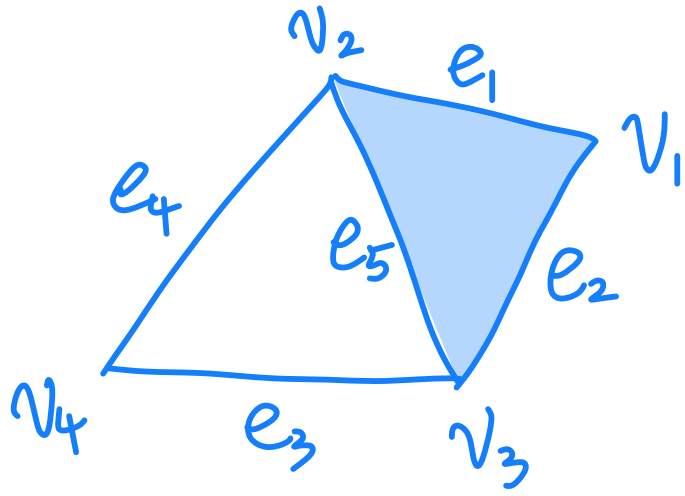
	$v_1v_2$	$v_1v_3$	$v_3v_4$	$v_2v_4$	$v_2v_3$
	e1	e2	e3	e4	e5
v1	1	1	0	0	0
v2	1	0	0	1	1
v3	0	1	1	0	1
v4	0	0	1	1	0

add  $e_3$  to  $e_4$



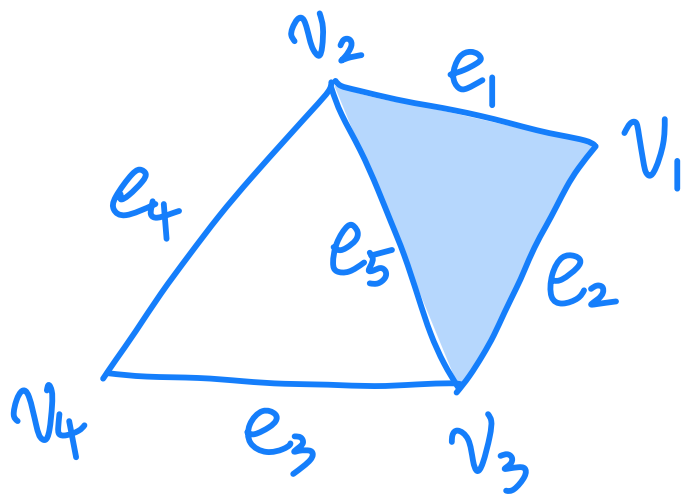
	$v_1v_2$	$v_1v_3$	$v_3v_4$	$v_2v_4$	$v_2v_3$
	e1	e2	e3	e4+e3	e5
v1	1	1	0	0	0
v2	1	0	0	1	1
v3	0	1	1	1	1
v4	0	0	1	0	0

add  $e_2$  to  $e_4 + e_3$



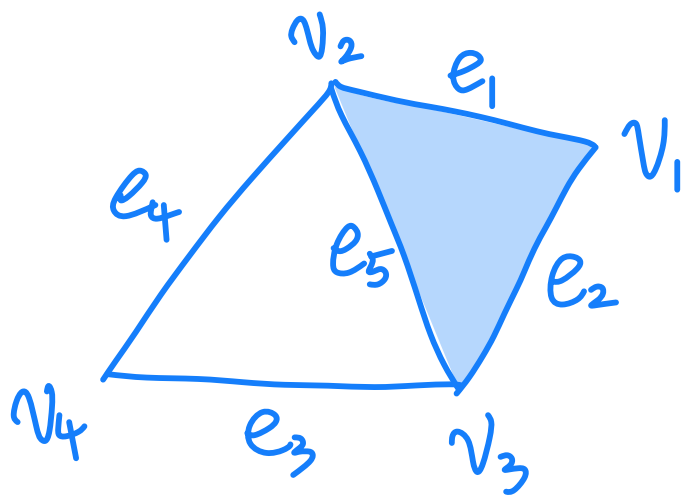
	$v_1v_2$	$v_1v_3$	$v_3v_4$	$v_2v_4$	$v_2v_3$
	e1	e2	e3	e4+e3+e2	e5
v1	1	1	0	1	0
v2	1	0	0	1	1
v3	0	1	1	0	1
v4	0	0	1	0	0

add  $e_1$  to  $e_4 + e_3 + e_2$



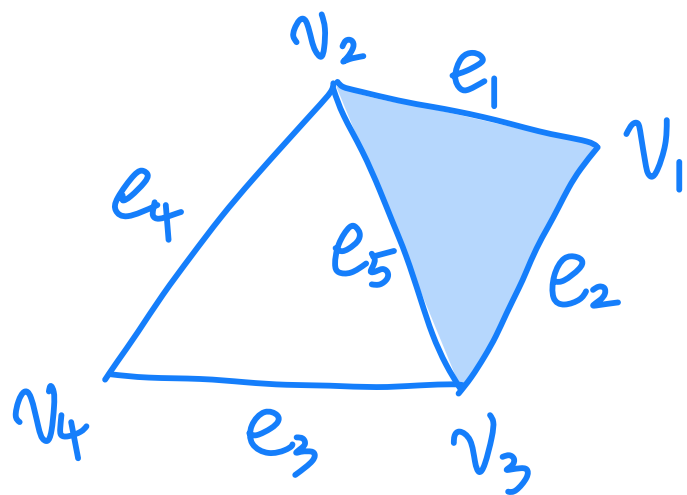
	$v_1v_2$	$v_1v_3$	$v_3v_4$	$v_2v_4$	$v_2v_3$
	e1	e2	e3	e4+e3+e2+e1	e5
v1	1	1	0	0	0
v2	1	0	0	0	1
v3	0	1	1	0	1
v4	0	0	1	0	0

add  $e_2$  to  $e_5$

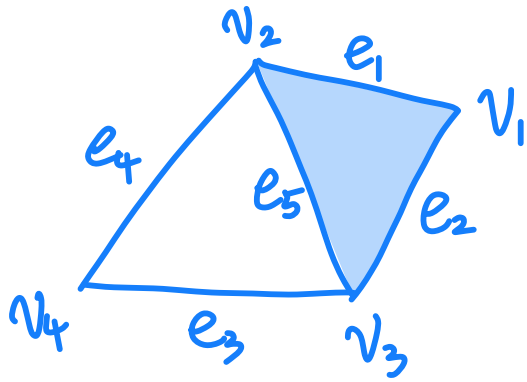


	$v_1v_2$	$v_1v_3$	$v_3v_4$	$v_2v_4$	$v_2v_3$
	e1	e2	e3	e4+e3+e2+e1	e5+e2+e1
v1	1	1	0	0	1
v2	1	0	0	0	1
v3	0	1	1	0	0
v4	0	0	1	0	0

add  $e_1$  to  $e_5 + e_2$



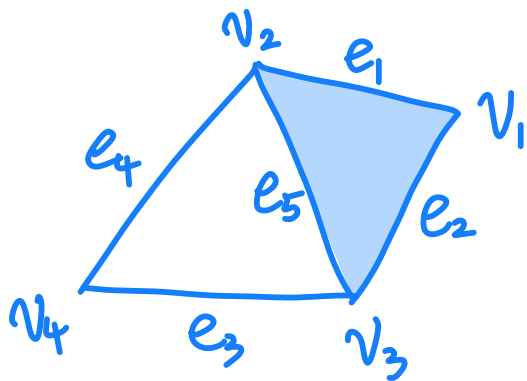
	$v_1v_2$	$v_1v_3$	$v_3v_4$	$v_2v_4$	$v_2v_3$
	e1	e2	e3	e4+e3+e2+e1	e5+e2+e1
v1	1	1	0	0	0
v2	1	0	0	0	0
v3	0	1	1	0	0
v4	0	0	1	0	0



$$A_1 \xrightarrow{\text{reduction}} \tilde{A}_1 =$$

	$v_1 v_2$ e1	$v_1 v_3$ e2	$v_3 v_4$ e3	$v_2 v_4$ e4+e3+e2+e1	$v_2 v_3$ e5+e2+e1
v1	1	1	0	0	0
v2	1	0	0	0	0
v3	0	1	1	0	0
v4	0	0	1	0	0

- ▶  $\dim B_0 = 3$ ,  $\dim(B_1) = \text{rank}(A_1) = \# \text{ of pivots}$
- ▶  $\dim Z_1 = 2$ ,  $\dim(Z_1) = \dim C_1 - \text{rank}(A_1)$   
 $= \# \text{ of columns} - \# \text{ of pivots}$



$$A_1 \xrightarrow{\text{reduction}} \tilde{A}_1 =$$

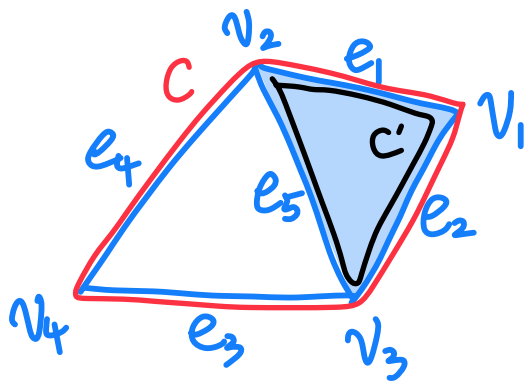
	$v_1 v_2$	$v_1 v_3$	$v_3 v_4$	$v_2 v_4$	$v_2 v_3$
	e1	e2	e3	e4+e3+e2+e1	e5+e2+e1
v1	1	1	0	0	0
v2	1	0	0	0	0
v3	0	1	1	0	0
v4	0	0	1	0	0

►  $\beta_p = \dim Z_p - \dim B_p$

$$\begin{aligned} \beta_0 &= \dim Z_0 - \dim B_0 = \dim C_0 - \dim B_0 = \# \text{ of rows} - \# \text{ of pivots} \\ &= 4 - 3 = 1 \end{aligned}$$







$$A_1 \xrightarrow{\text{reduction}} \tilde{A}_1 =$$

	$v_1 v_2$	$v_1 v_3$	$v_3 v_4$	$v_2 v_4$	$v_2 v_3$
	e1	e2	e3	e4+e3+e2+e1	e5+e2+e1
v1	1	1	0	0	0
v2	1	0	0	0	0
v3	0	1	1	0	0
v4	0	0	1	0	0
	$\parallel$ $\alpha_1$	$\parallel$ $\alpha_2$	$\parallel$ $\alpha_3$		

► Bases for  $B_0$  and  $Z_1$

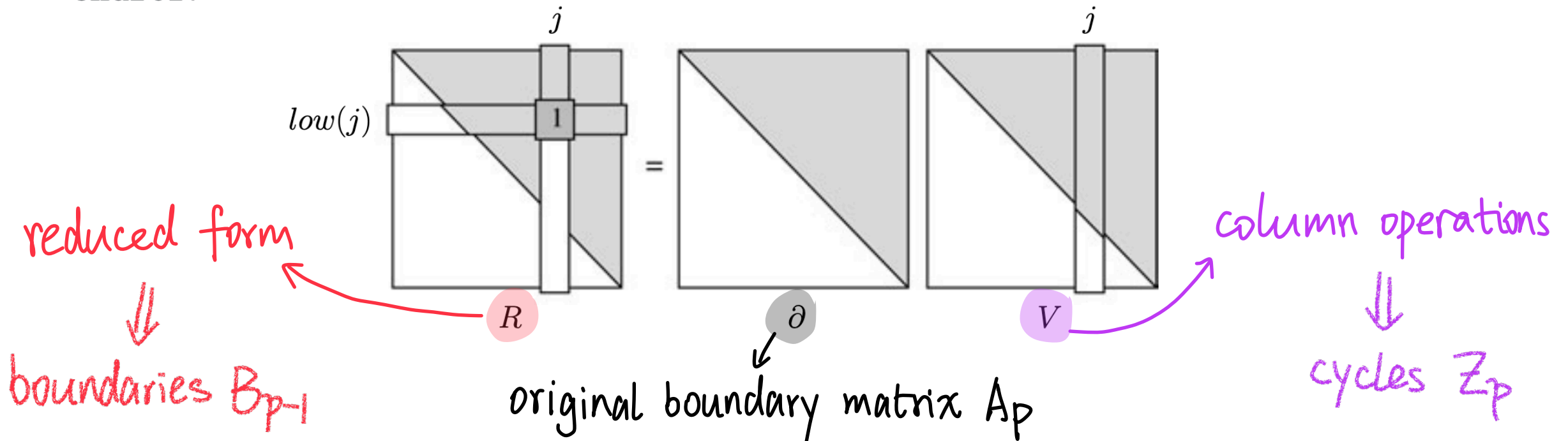
•  $\vec{v} := (v_1, v_2, v_3, v_4) \Rightarrow B_0 = \langle \alpha_1 \cdot \vec{v}, \alpha_2 \cdot \vec{v}, \alpha_3 \cdot \vec{v} \rangle$

•  $Z_1 = \langle \underbrace{e_1 + e_2 + e_3 + e_4}_{\parallel C}, \underbrace{e_1 + e_2 + e_5}_{\parallel C'} \rangle$

$Z_1$

# Left-to-right Column Reduction Algorithm

$R = \partial$ ;  $V = \text{Id}_{n_p \times n_p}$   
 for  $j = 1$  to  $m$  do  
   while there exists  $j_0 < j$  with  $\text{low}(j_0) = \text{low}(j)$  do  
     add column  $j_0$  to column  $j$ ; add  $V_{j_0}$  to  $V_j$   
   endwhile  
 endfor.



# Properties

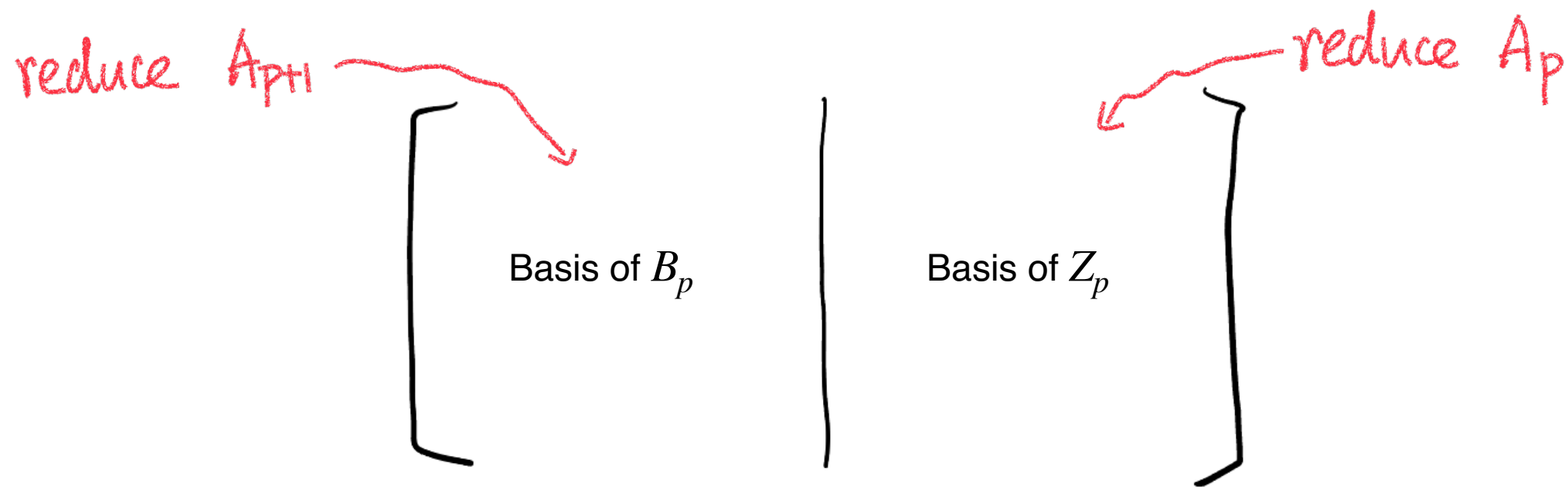
- ▶ Let  $n_p$  = the number of p-simplices. Let  $A_j$  be the j-th column of A.

- ▶ Theorem:

- ▶ The algorithm terminates in  $O(n_p n_{p-1}^2)$  time
- ▶ The output matrix  $R$  is in column reduced form
- ▶ The set  $\{R_j : R_j \neq 0\}$  form a basis for  $B_{p-1}$
- ▶ The set  $\{V_j : R_j = 0\}$  form a basis for  $Z_p$

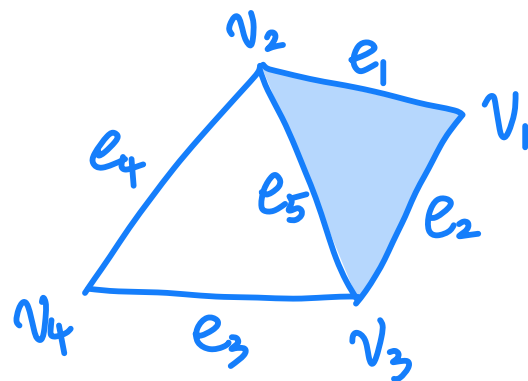
This is not the only reduction algorithm!! Any elimination via row/column additions to convert a matrix into a reduced form works!

# Computing a basis for homology



- ▶ Left part is already column reduced
  - ▶ Apply Reduction to the above matrix to obtain basis of  $H_p$
- the RHS after reduction

# Computing a basis for homology

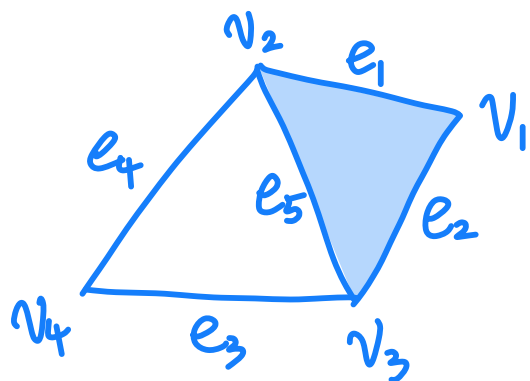


$B_1$

$Z_1$

	$e_5 + e_2 + e_1$		$e_4 + e_3 + e_2 + e_1$	$e_5 + e_2 + e_1$
E1	1		1	1
E2	1		1	1
E3	0		1	0
E4	0		1	0
E5	1		0	1

# Computing a basis for homology



a basis for  $H_1$  ←



$$H_1 = \langle [e_4 + e_3 + e_2 + e_1] \rangle$$

	$e_5 + e_2 + e_1$		$e_4 + e_3 + e_2 + e_1$	0
E1	1		1	0
E2	1		1	0
E3	0		1	0
E4	0		1	0
E5	1		0	0

# Functoriality of Homology

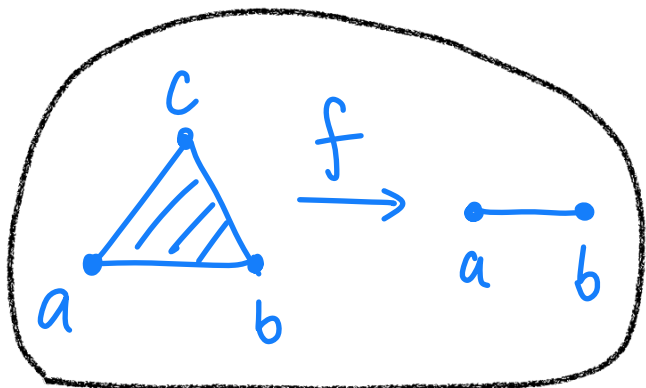


# Functoriality of Simplicial Homology

- ▶ Recall: Given simplicial complexes  $K$  and  $L$ , a function  $f : V(K) \rightarrow V(L)$  is called a **simplicial map** if for any  $\sigma = \{p_0, \dots, p_d\} \in \Sigma(K)$ ,  
 $f(\sigma) = \{f(p_0), \dots, f(p_d)\}$  spans a simplex in  $L$ , i.e.,  $f(\sigma) \in \Sigma(L)$ .
- ▶ Let  $K = (V, \Sigma)$  and  $K' = (V', \Sigma')$  and let  $f : V \rightarrow V'$  be a simplicial map. Then,
  - ▶  $f$  induces a linear map on homology groups  $f_p : H_p(K) \rightarrow H_p(K')$
  - ▶ If there exist  $K'' = (V'', \Sigma'')$  and another simplicial map  $g : V' \rightarrow V''$ , then
    - ▶  $(g \circ f)_p = g_p \circ f_p$

# Construction of $f_p$

- ▶ First,  $f$  induces linear maps on chain spaces  $\bar{f}_p : C_p(K) \rightarrow C_p(K')$ 
  - ▶  $\bar{f}_p(\sigma) = \begin{cases} f(\sigma) & \text{if } f(\sigma) \text{ is a } p\text{-simplex} \\ 0 & \text{otherwise} \end{cases}$
- ▶ Then,  $\bar{f}_p : C_p(K) \rightarrow C_p(K')$  induces  $f_p : H_p(K) \rightarrow H_p(K')$ 
  - ▶  $f_p([c]) := [\bar{f}_p(c)]$

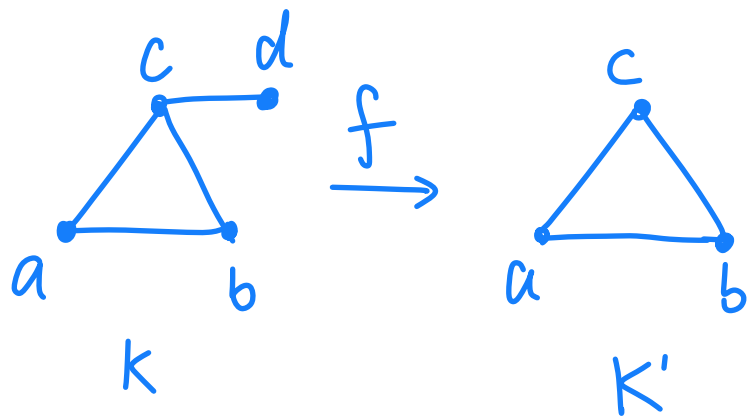


$$\begin{array}{ccccc} \cdots & \longrightarrow & C_p(K) & \longrightarrow & C_{p-1}(K) & \longrightarrow & \cdots \\ & & \bar{f}_p \downarrow & & \bar{f}_{p-1} \downarrow & & \\ \cdots & \longrightarrow & C_p(K') & \longrightarrow & C_{p-1}(K') & \longrightarrow & \cdots \end{array}$$

$$f_p : H_p(K) \rightarrow H_p(K')$$

for every  $p$

# Construction of $f_p$



$$\begin{array}{c}
 \langle ab, ac, bc, cd \rangle \\
 \parallel \\
 \dots \rightarrow C_1(K) \xrightarrow{\partial_1} C_0(K) \rightarrow \dots \\
 \bar{f}_1 \downarrow \qquad \qquad \bar{f}_0 \downarrow \\
 \dots \rightarrow C_1(K') \xrightarrow{\partial_1} C_0(K) \rightarrow \dots \\
 \parallel \\
 \langle ab, ac, bc \rangle
 \end{array}$$

$$f_1: H_1(K) \rightarrow H_1(K')$$

$$[\alpha] \mapsto [\bar{f}_1(\alpha)]$$

e.g: for  $\alpha = ab + bc + ac$

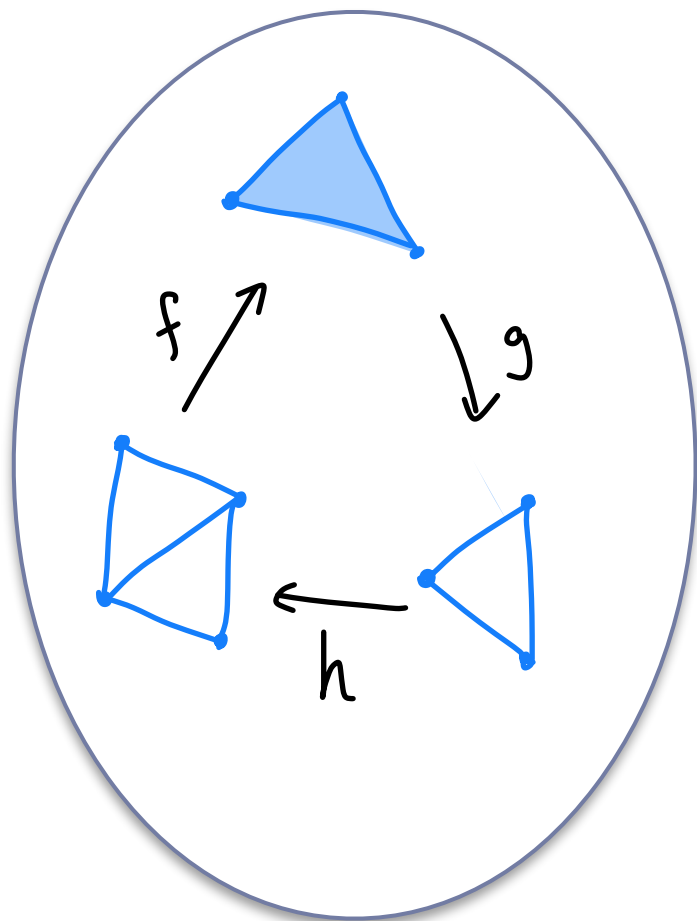
$$\Rightarrow f_1([\alpha]) = [\bar{f}_1(\alpha)] = [ab + bc + ac]$$

$$\bar{f}_0: a \mapsto a, b \mapsto b, c \mapsto c, d \mapsto c$$

$$\bar{f}_1: ab \mapsto ab, ac \mapsto ac, bc \mapsto bc$$

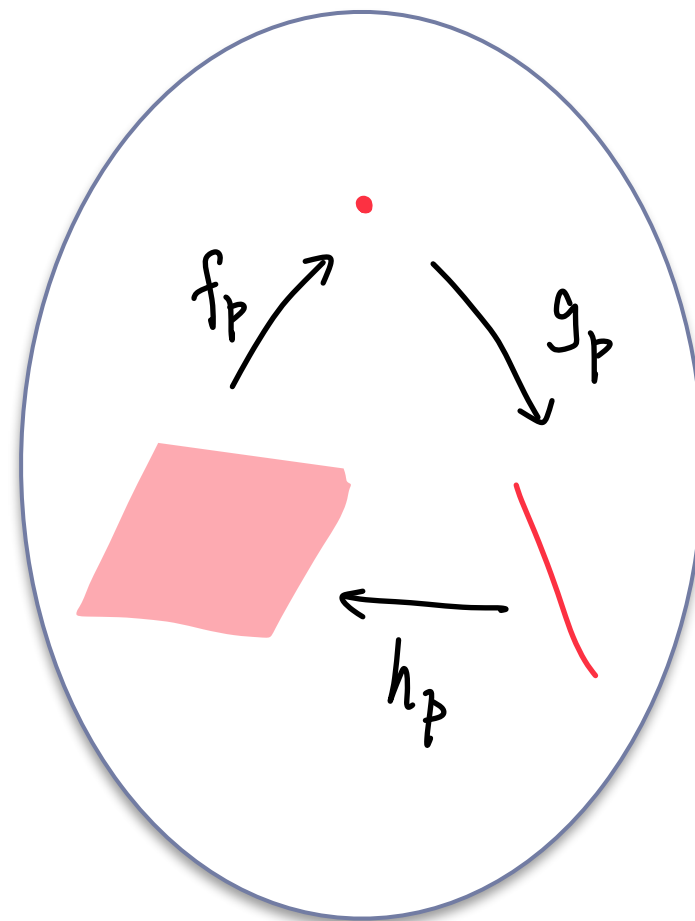
$$cd \mapsto 0$$

# Mind picture of functoriality



simplicial complexes

$$\xrightarrow[\begin{matrix} H_p(\quad; F) \end{matrix}]{\text{homology}}$$



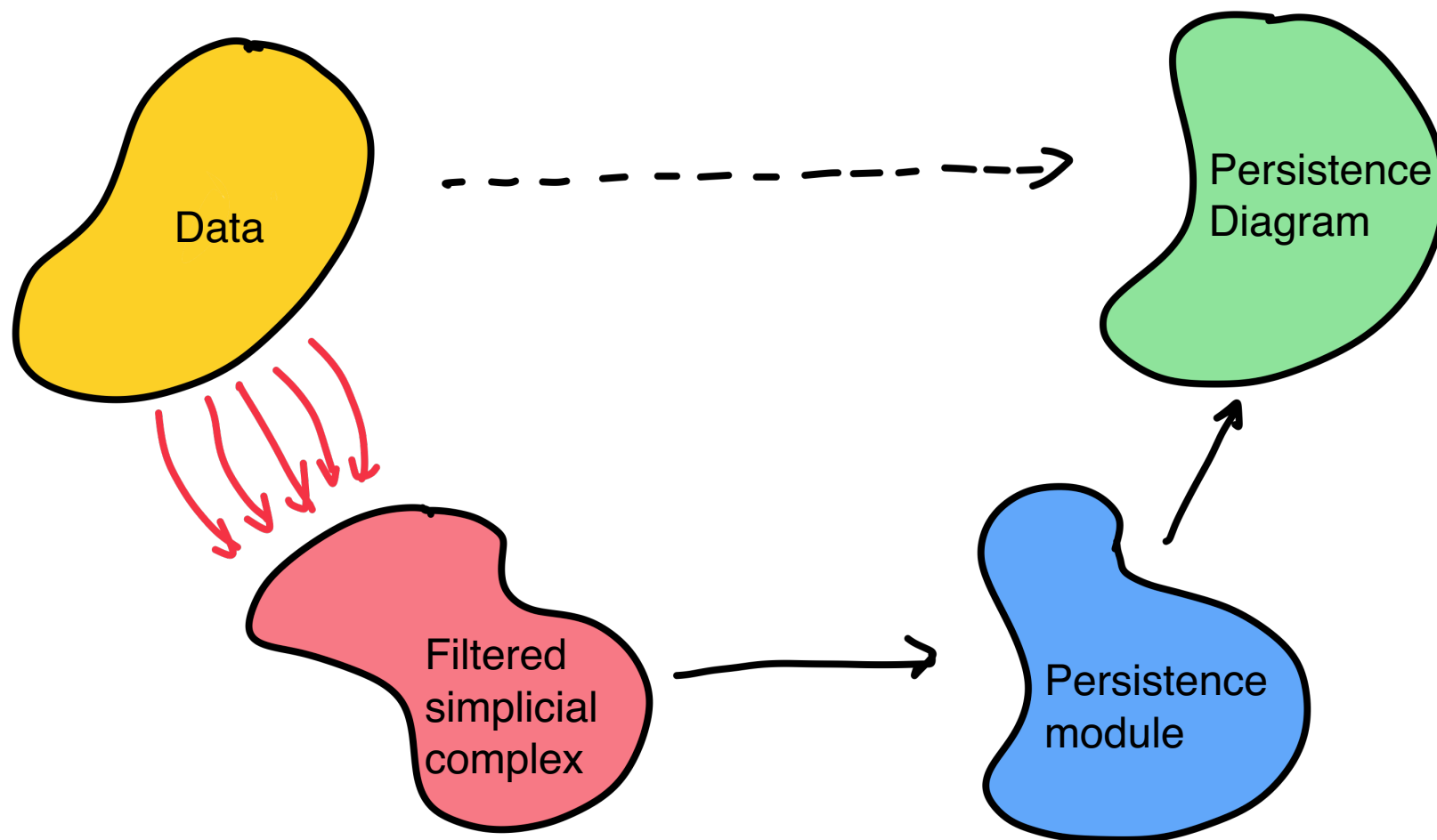
vector spaces

# Topic 4: Intro to Persistent Homology

# Persistent homology

- ▶ A modern extension of homology to “**sequence of spaces**”
  - ▶ [Edelsbrunner, Letcher, and Zomorodian, FOCS 2000]
  - ▶ Significantly broaden its practical power
- ▶ What is persistent homology (PH)
  - ▶ Motivation
  - ▶ Persistent betti numbers and persistence diagrams
- ▶ Algorithm(s) for persistent homology

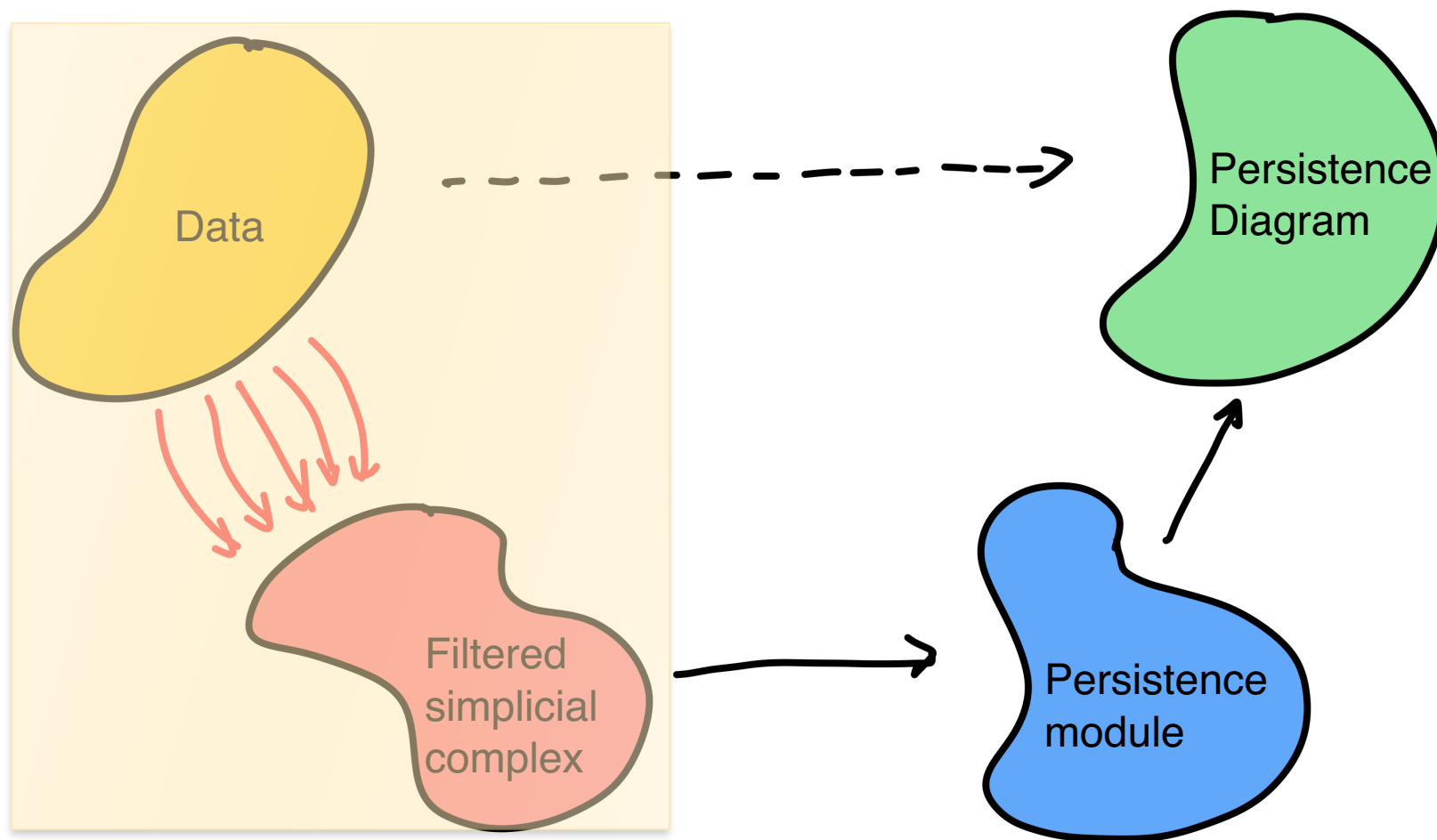
# Mind picture



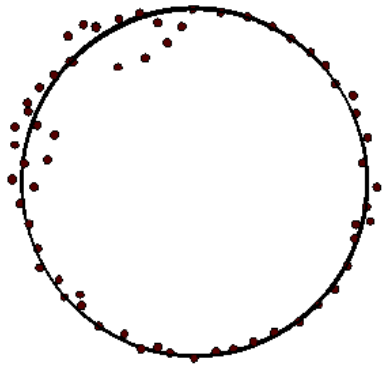
# Filtrations



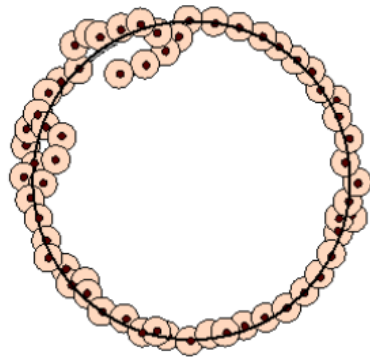
# Filtered simplicial complex



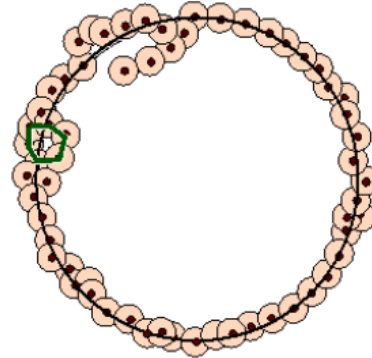
# Issue of Scale



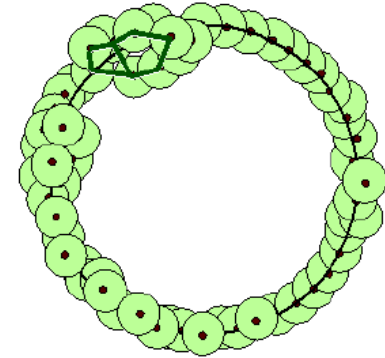
(a)



(b)



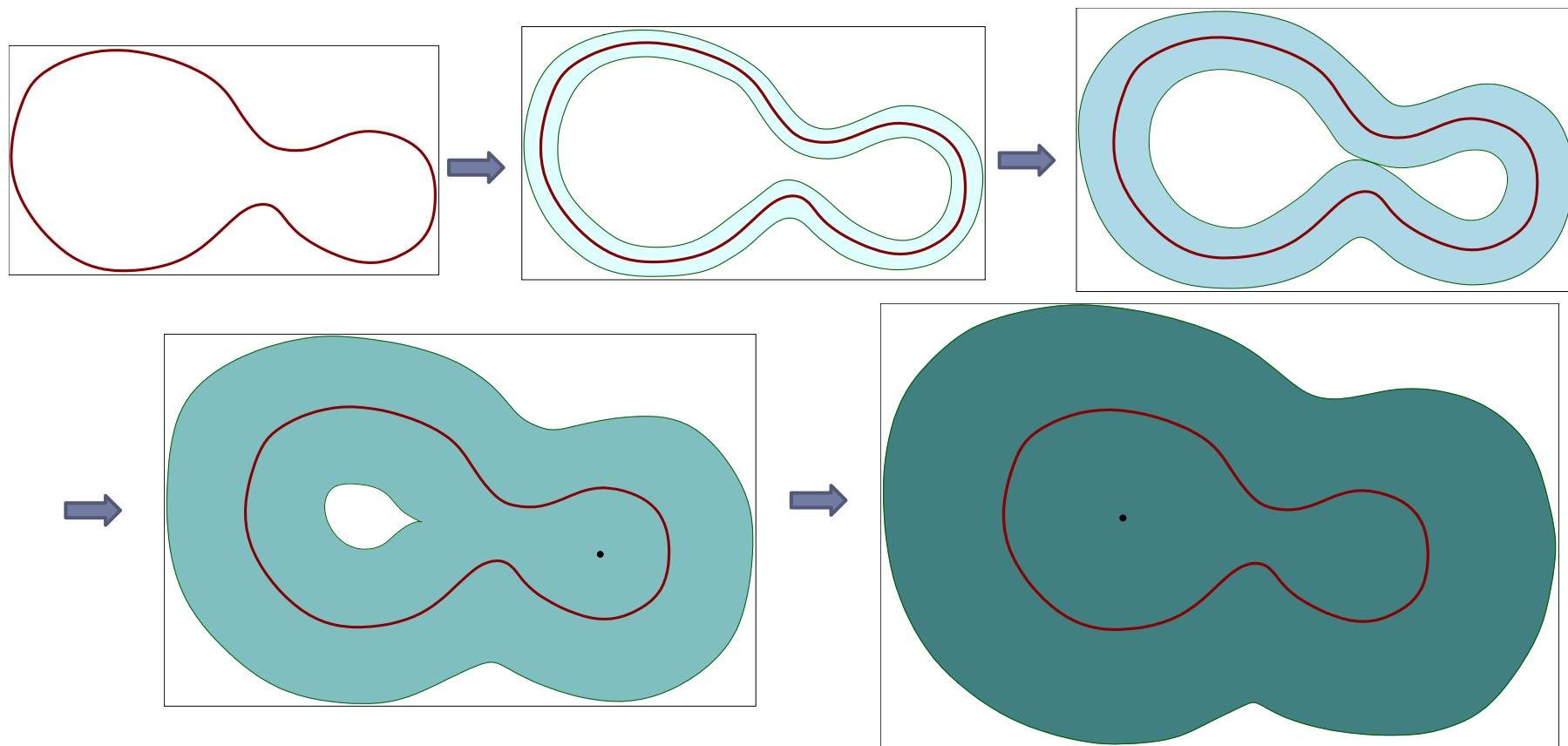
(c)



(d)

- ▶ Which scale to take?
- ▶ No single good scale!
- ▶ All scales?
- ▶ Some ``features'' persists longer than others

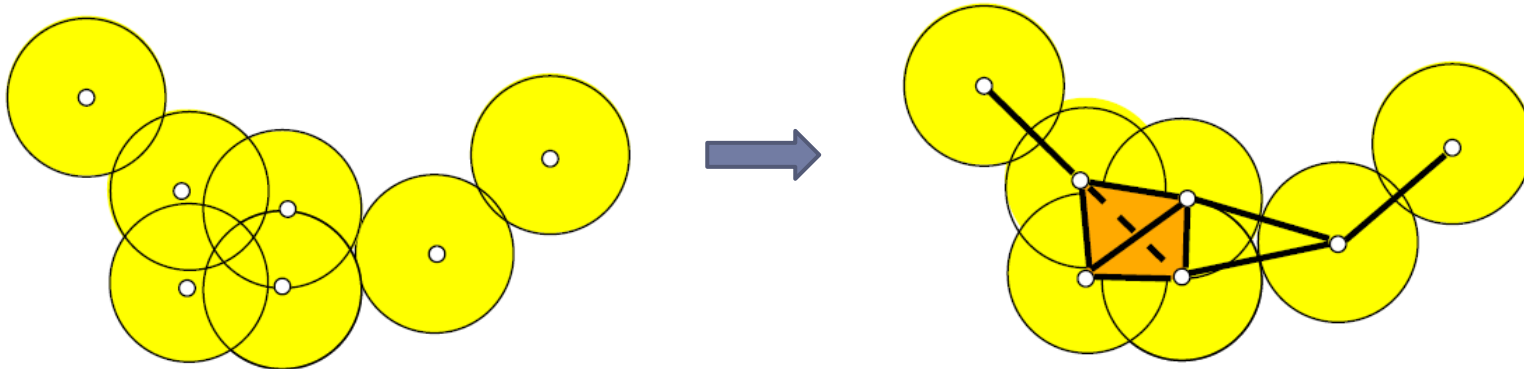
## Another Example



- ▶ Want to capture features of different “sizes”

# Recall: Čech Complex $C^r(P) := \text{Nrv} \left( \{ B(p_i, r) : p_i \in P \} \right)$

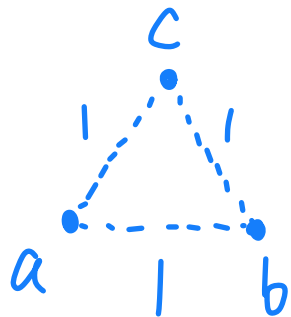
- ▶ Given a set of points  $P = \{p_1, \dots, p_n\} \subset \mathbb{R}^d$
- ▶ Given a real value  $r > 0$ , the **Čech complex**  $C^r(P)$  is the **nerve** of the set of closed balls  $\{B(p_i, r)\}_{i=1, \dots, n}$ , where  $B(p, r) = \{x \in \mathbb{R}^d \mid d(p, x) \leq r\}$ 
  - ▶ i.e,  $\sigma = \{p_{i_0}, \dots, p_{i_s}\} \in C^r(P)$  iff  $\bigcap_{j=0, \dots, s} B(p_{i_j}, r) \neq \emptyset$
- ▶ The definition can be extended to a finite sample  $P$  of a metric space  $(X, d)$ .



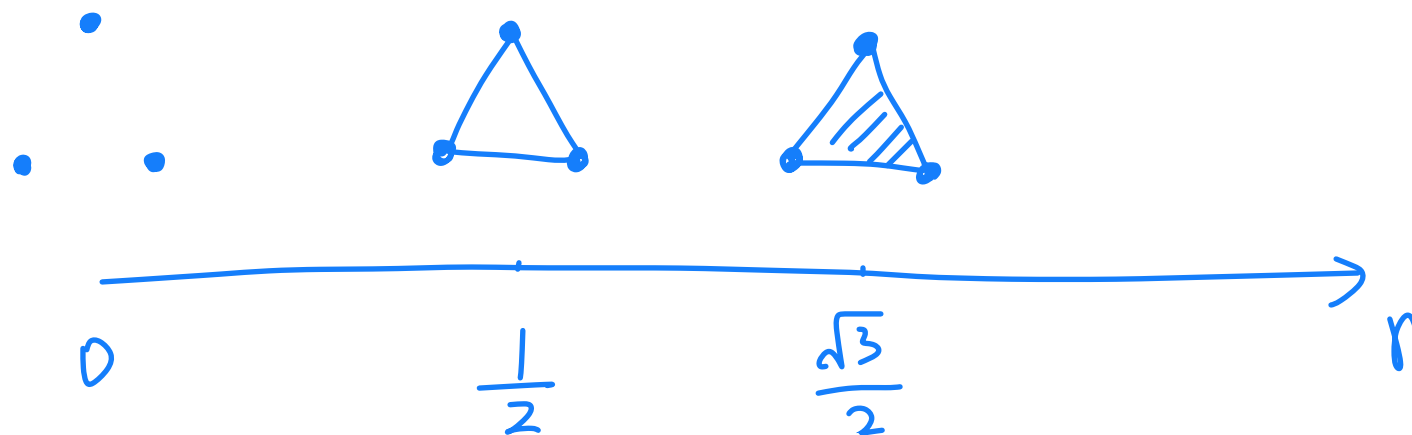
# Čech Filtration

$$C^r(P) = \{ p_{i_1} \cdots p_{i_k} \mid \bigcap_j B(p_{i_j}, r) \neq \emptyset \}$$

- ▶ Given a set of points  $P = \{p_1, \dots, p_n\} \subset \mathbb{R}^d$
- ▶  $(C^r(P))_{r \geq 0}$  is called the Čech filtration



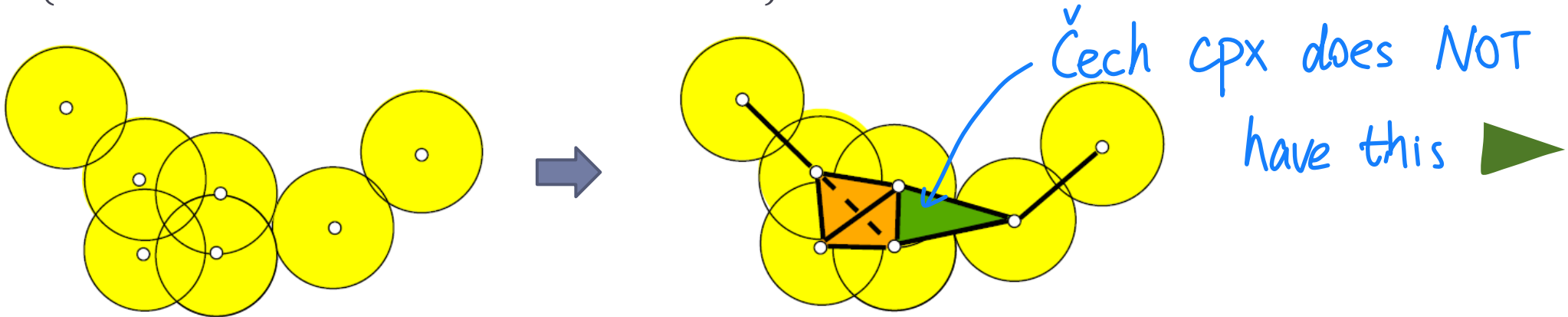
$C^r(\{a, b, c\})$ :



$$C^r(\{a, b, c\}) = \begin{cases} \text{three points} & \text{if } r \in [0, \frac{1}{2}) \\ \text{triangle} & \text{if } r \in [\frac{1}{2}, \frac{\sqrt{3}}{2}) \\ \text{filled triangle} & \text{if } r \geq \frac{\sqrt{3}}{2} \end{cases}$$

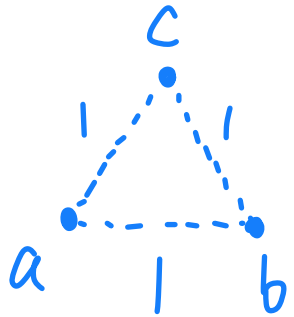
# Recall: Vietoris-Rips (Rips) Complex

- ▶ Given a set of points  $P = \{p_1, \dots, p_n\} \subset \mathbb{R}^d$
- ▶ Given a real value  $r > 0$ , the *Vietoris-Rips (Rips) complex*  $Rips^r(P)$  is:
  - ▶  $\left\{ (p_{i_0}, \dots, p_{i_k}) \mid B(p_{i_l}, r) \cap B(p_{i_j}, r) \neq \emptyset, \forall l, j = 0, \dots, k \right\}$ .
- ▶ More generally for  $P$  in a metric space  $(X, d)$ :
  - ▶  $Rips^r(P) = \left\{ (p_{i_0}, \dots, p_{i_k}) \mid d(p_{i_l}, p_{i_j}) \leq 2r, \forall l, j = 0, \dots, k \right\}$ .

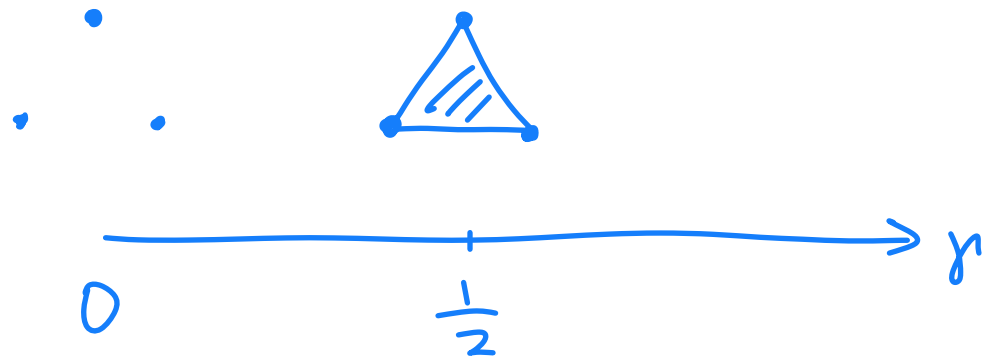


# Vietoris-Rips (Rips) Filtration $Rips^r(P) = \{p_{i_1} \cdots p_{i_k} \mid d(p_{i_j}, p_{i_l}) \leq \underline{2r}\}$

- ▶ Given a set of points  $P = \{p_1, \dots, p_n\} \subset \mathbb{R}^d$
  - ▶  $(Rips^r(P))_{r \geq 0}$  is called the Vietoris-Rips (Rips) Filtration
- Another convention is to use  $r$  instead.



$$Rips^r(\{a, b, c\})$$



$$Rips^r(\{a, b, c\}) = \begin{cases} \text{three points} & \text{if } r \in [0, \frac{1}{2}) \\ \text{filled triangle} & \text{if } r \in [\frac{1}{2}, \infty) \end{cases}$$

# Finitely presented filtration

- ▶ A filtration  $(K_t)_{t \in [0, \infty)}$  is called **finitely represented** if
  - ▶ There exist  $0 = t_0 < t_1 < \dots < t_n$  such that
  - ▶  $K_t = K_{t'}, \quad \forall t_i \leq t < t' < t_{i+1}$  and  $i = 0, \dots, n$  ( $t_{n+1} := \infty$ )
- ▶ So  $(K_t)_{t \in [0, \infty)}$  is essentially the same as (or can be reconstructed from)  
 $(K_{t_i})_{i=0, \dots, n}$
- ▶ Both Čech and Rips filtrations are finitely represented