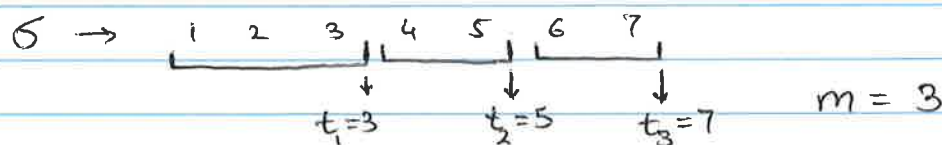
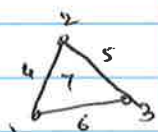


Mar 2

### # Spectral Sequences Algorithm

→ Consider filtration  $\phi = K_0 \subseteq K_1 \subseteq \dots \subseteq K_n$   
 with total compatible ordering of simplices  $\sigma_1, \sigma_2, \dots, \sigma_n$   
 Let  $M$  be the  $n \times n$  matrix  $\rightarrow$  boundary matrix.  
 $i$ th column of  $M$  represents  $\sigma_i$ , the  $i$ th simplex in ordering.

→ Consider sequence of indices  $[0 < t_1 < t_2 < \dots < t_m = n]$   
 partition the sequence of  $n$  simplices in the ordering into  
 $m$  non-overlapping parts s.t.  $j$ th part consists of  
 simplices indexed  $t_{j-1} + 1$  to  $t_j$



→ This converts the  $n \times n$  boundary matrix  $M$  into an  
 $m \times m$  block matrix  $D$

$$M = \begin{bmatrix} & 1 & 2 & 3 & 4 & 5 & 6 & 7 \\ 1 & & & & 1 & & 1 & \\ 2 & & & 1 & 1 & & & \\ 3 & & & & 1 & 1 & & \\ 4 & & & & & & & 1 \\ 5 & & & & & & & 1 \\ 6 & & & & & & & 1 \\ 7 & & & & & & & \end{bmatrix}$$

$$D = \begin{bmatrix} 1 & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} \\ 2 & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} \\ 3 & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} \\ 4 & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} \\ 5 & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} \\ 6 & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} \\ 7 & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} & \boxed{\phantom{000}} \end{bmatrix}$$

→ This is equivalent to having a filtration with multiple  
 simplices added at each step.

block  $D_j^i \Rightarrow$  simplices added in  $i$ th step that have  
 faces in  $j$ th step.

Notation: superscript: column chunk / block  
 subscript: row chunk / block.

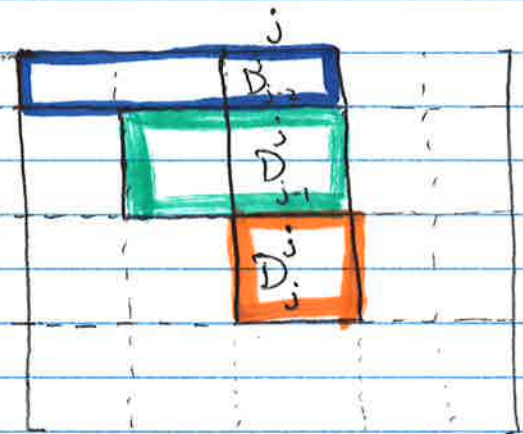
### # Algorithm: Given $m \times m$ block matrix $D$

- For  $r = 1$  to  $m$
- For  $j = r$  to  $m$
- For columns  $l$  in  $D_j^j$   
(from left to right)
- while ( $low(l) \in D_{j-r+1}^{j-r+1}$  and  
 $\exists l' \text{ in } D_{j-r+1}^{j-r+1} \text{ s.t.}$   
 $low(l') = low(l)$ )
- add  $l'$  to  $l$  (columns)
- end while
- end for
- end for
- end for

$r$  indicates "phases" of the algorithm.

**In phase 1** we consider for each column chunk, only columns that have pivot in the diagonal block.

Only reduce column  $l$  if there is a column  $l'$  in the same column chunk with its pivot in diagonal block.



processing  $j^{\text{th}}$  column chunk.

in phase 1,  $r=1$  only consider columns that have pivot in  $D_j^j$   
 $l' \rightarrow$  columns in  $D_j^j$  with pivot in  $D_j^j$  (on left of col  $l$ )

in phase 2,  $r=2$ . Consider columns  $l$  in  $D_j^j$  that have pivot in  $D_{j-1}^{j-1}$   
 $l' \rightarrow$  columns in  $D_{j-1}^{j-1}$  that have pivot in  $D_{j-1}^{j-1}$  and  $D_{j-1}^{j-1}$

- $l'$  belongs to
- in phase 1
  - in phase 2
  - in phase 3

- $\rightarrow$  In the standard reduction algorithm we sweep columns from left to right
- $\rightarrow$  spectral sequences algorithm  $\rightarrow$  sweeping diagonally.
- $\rightarrow$  persistence pairs are found in increasing order of persistence

## Persistent homology Computation: Key ideas

- ① We want to find pairs of simplices  $(i, j)$  [persistence pairs] where
- $i \rightarrow$  positive simplex, generator of PH class
  - $j \rightarrow$  negative simplex, destroyer of PH class.

$\therefore$  Given total compatible ordering of simplices in filtration

- ① Construct boundary matrix, ② apply reduction ③ infer pairs

$\rightarrow$  Pairs of simplices correspond to pairs of columns in boundary  $M$ .

- ② A simplex  $\sigma_i$  ( $i^{\text{th}}$  column in BM) can either be +ve or -ve
- $\rightarrow$  +ve columns reduce to 0 after reduction
  - $\rightarrow$  -ve columns have non-zero pivots

## # Strategies to reduce number of computations

- ① Clearing: Each column  $M^j$  either reduces to 0 or gives a persistence pair  $(i, j)$  where  $i = \text{pivot}(j)$  then  $i$  is +ve column and will eventually reduce to 0
- $\Rightarrow$  if found pair  $(i, j)$  set column  $i$  to 0.
  - $\Rightarrow$  Only works if reduction is applied from right to left.

- ② Compression: if found persistence pair  $(i, j) \Rightarrow j$  is -ve col.  $\therefore j$  can not be a pivot (+ve column) for any other col.
- $\Rightarrow$  set row  $j$  to 0 because there is no pivot in the  $j^{\text{th}}$  row.

i.e. if found column  $M^j$  with  $i = \text{pivot}(j)$  [pair  $(i, j)$ ]

then clearing  $\Rightarrow$  set col  $i$  to 0 [requires right to left]

Compression  $\Rightarrow$  set row  $j$  to 0 [requires left to right]

## Reduction in chunks

→ Construct the  $m \times m$  block matrix  $D$  from boundary matrix  $M$ .

Idea! (a) Perform few steps of spectral sequence algorithm (2-phases)  
 → right to left reduction within blocks/chunks  
 so that clearing optimization can be applied.

→ Only finds local pairs  $(i, j)$  where  $i$  and  $j$  are from the same or ~~one~~ adjacent chunks.

→ if col  $j$  has pivot  $i$  s.t.  $i$  belongs to the same chunk as  $j$  or to chunk adjacent to  $j$  then  $j$  is called local column.  
 all other columns are global columns.

(b) Next, independently compress all columns globally.  
 (Given local pairs  $(i, j)$  from (a) set rows  $j$  to 0)

(c) After (a) & (b), we are left with a submatrix of global rows and columns → reduce using clearing.

### Algorithms!

Compress  $(R, k)$

- for non-zero entry index  $l$  of  $R_k$
- | if  $l$  is paired
- | | if  $l$  is inactive
- | | |  $R_k^l \leftarrow 0$
- | | else
- | | |  $j \leftarrow L[l]$
- | | |  $R^k \leftarrow R^k + R^j$
- | | end if
- | end if
- end for.

Mask\_Active\_Entries  $(R)$

- For each unpaired col  $R^k$
- | Mark\_Column  $(R, k)$

Mask\_Column  $(R, k)$

- For each non-zero entry index  $i$  of  $R^k$
- | if  $i$  is unpaired
- | | mark  $k$  as active
- | | if  $i$  is +ve
- | | |  $j \leftarrow L[i]$
- | | | if  $j \neq k$  and Mask\_Column  $(R, j)$
- | | | | mark  $k$  as active
- | | end if
- | end if
- mark  $k$  as inactive

Local-Reduction ( $M, t_1, t_2, \dots, t_m$ )

$R \leftarrow M; L \leftarrow [0, 0, \dots, 0]$

• for  $\delta = d, d-1, \dots, 0$

• for  $r = 1, 2$

• for  $b = r, \dots, m$

• for  $j = t_{b-1} + 1, \dots, t_b$   
with  $\dim \sigma_j = \delta$

• if  $j$  is not paired

• while ( $R^j \neq 0$  and  $L[\text{pivot}(R^j)] \neq 0$

and  $\text{pivot}(R^j) > t_{b-r}$ )

•  $R^j \leftarrow R^j + R^{L[\text{pivot}(R^j)]} \xrightarrow{\text{reduction step}}$

• if  $R^j \neq 0$

•  $i \leftarrow \text{pivot}(R^j)$

• if  $i > t_{b-r}$

•  $L[i] \leftarrow j$

•  $R^i \leftarrow 0 \xrightarrow{\text{clearing step}}$

• mark  $(i, j)$  paired

[Right to left in decreasing order  
of simplex dimension]

this loop can be parallelized

[local reduction of column chunks  
in spectral sequence phases  
is independent]

$\Rightarrow M$  is the boundary matrix.

$t_i$  define block boundaries in block matrix  $D$  [implicit]

$\Rightarrow$  Right to left reduction within each ~~to~~ column chunk is applied in decreasing order of simplex dimensions

why?  $\Rightarrow$  (a) Reduce simplex before its faces

(b) Reduce all  $d$ -dimensional simplices before  $(d-1)$ -dimensional simplices within the block.

Persistence in chunks : Given boundary matrix  $M$  and block boundary indices  $t_0, t_1, \dots, t_m$

- $(R, L) \leftarrow \text{Local\_Reduction}(M, t_0, t_1, \dots, t_m)$  } (Can be parallelized)
  - $\text{Mask\_Active\_Entries}(R)$
  - for  $\delta = d, d-1, \dots, 0$ 
    - for  $j = 1, \dots, n$  and  $\dim(\sigma_j) = \delta$ 
      - if col  $j$  is not paired } Global column Compression
      - Compress  $(R, j)$  } (Can be parallelized)
    - for  $j = 1, \dots, n$  and  $\dim(\sigma_j) = \delta$ 
      - while  $R^j \neq 0$  and  $L[\text{pivot}(R^j)] \neq 0$ 
        - $R^j \leftarrow R^j + R^{L[\text{pivot}(R^j)]}$  } Reducing Column
      - if  $R^j \neq 0$ 
        - $i \leftarrow \text{pivot}(R^j)$
        - $L[i] = j$
        - $R^i = 0$  } Clearing
        - pair  $(i, j)$  marking
- Reducing submatrix of global columns and rows. [Should have size much smaller than  $M$ ]

Run time: Let  $g$  be the number of global columns after step 1

$\rightarrow$  with  $\left(\frac{n}{\log n}\right)^m$  chunks, each of size  $\log n$   $g \ll n$   
ideally.

run time :  $O(n \log^2 n + gn \log n + g^3)$

$\rightarrow$  with  $(\sqrt{n})^m$  chunks, each of size  $\sqrt{n}$

runtime :  $O(n^2 + gn\sqrt{n} + g^3)$

$\Rightarrow$  much better than the  $O(n^3)$  runtime of standard algorithm.