

TDA - fall 2025

Vineyards +  
time-varying  
data



# Recap

- Find project proposals - due this week (slot coming soon)
- Office hours today, 2-3pm, if you have any questions
- Older assignments - please submit ASAP

## Stability of persistence

Previously, we saw several notions of stability, ie:

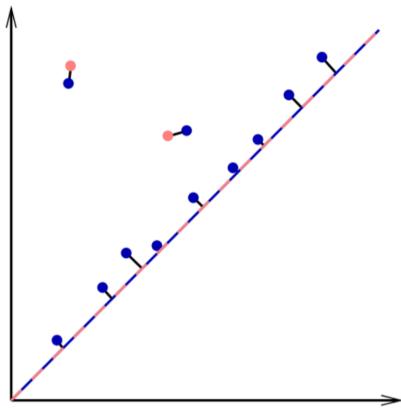
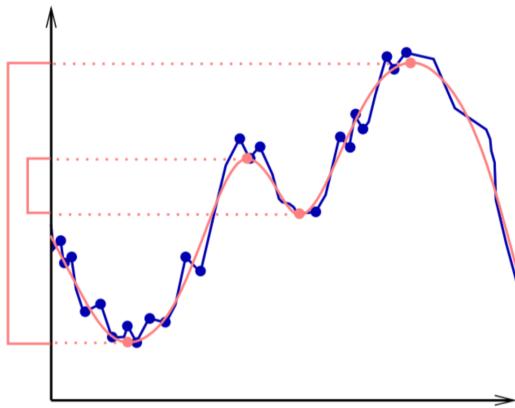
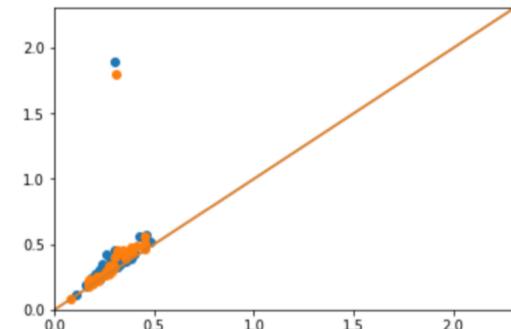
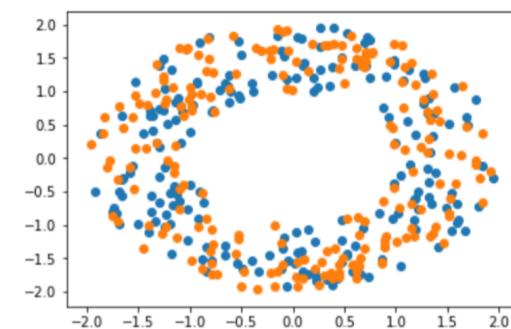
Fix a triangulatable space  $X$  with 2 continuous, tame functions  $f, g: X \rightarrow \mathbb{R}$ ,  
→ & let  $D_f$  &  $D_g$  be the resulting sublevel set persistence diagrams.

Then:  $B(D_f, D_g) \leq \|f - g\|_\infty$

(Similar statements for point clouds)

↳ Rips or Cech filtrations

# Pictures



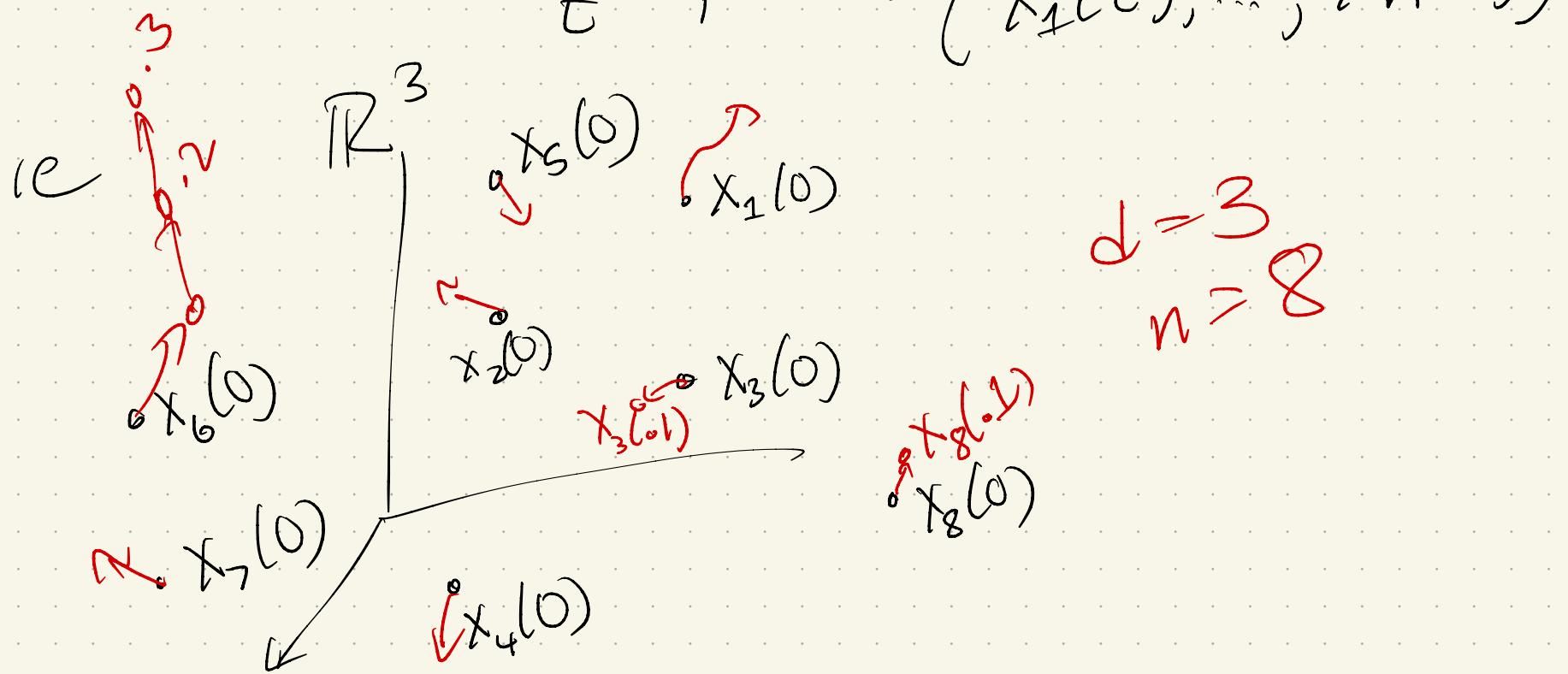
There are many notions of stability,  
but almost all result in an  
explicit map from  $D_g$  to  $D_f$   
(where points on the diagonal are also  
used).

# Time varying point clouds

A dynamic point cloud  $X(t) = (x_1(t), \dots, x_n(t))$  is a point cloud which is moving continuously, for  $t \in [0, 1]$ ,

i.e.  $[0, 1] \rightarrow (\mathbb{R}^d)^n$

$$t \mapsto (x_1(t), \dots, x_n(t))$$



# Vineyards

Cohen-Steiner, Edelsbrunner, Morozov  
2006

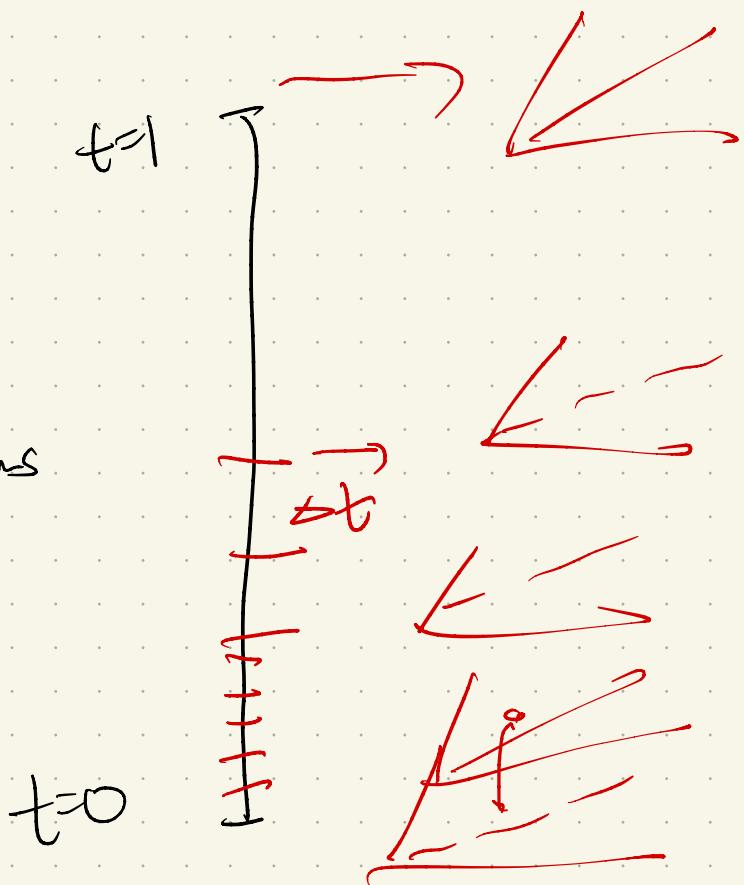
For each time  $t$ , we have a persistence diagram  $D(X(t))$ .

This one parameter family is called a vineyard:

$$V(X) = \{D(X(t)) \mid t \in [0, 1]\}$$

$$[0, 1] \rightarrow D \xleftarrow{\text{space of pers. diagrams}}$$

$$t \mapsto D(X(t))$$

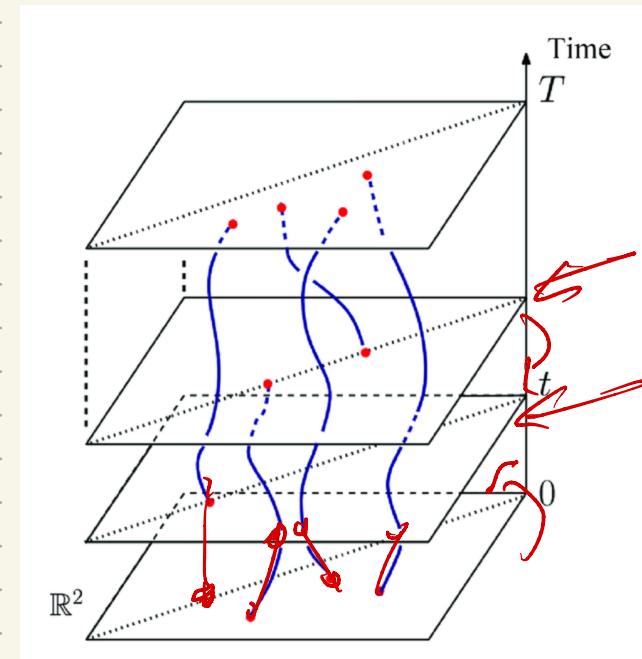


If the dynamic point set  $X(t)$  is continuous with respect to Hausdorff distance, the Vineyard is continuous with respect to bottleneck distance.

Why? Stability!

Vine: trace of some off-diagonal persistence point.

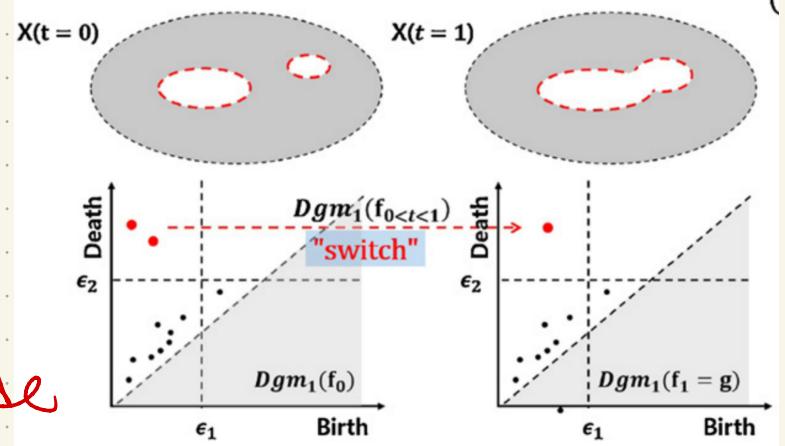
Note: things can appear from diagonal, disappear into it, & collide



Similarly, consider a simplicial complex  $|K|$  with 2 functions  $f, g: |K| \rightarrow \mathbb{R}$   
 & any homotopy  $f_t$  interpolating from  $f=f_0$  to  $g=f_1$  (with each  $f_t$  tame).  
 Vines are smoother except where critical pairs change

↳ called knees.

How? places where  
 birth/death times coincide



## Aside on vines

Not clear if arbitrary vineyards  
can decompose nicely into vines.

Munch 2013

Turner 2023

(mostly because of diagonal & knees)

Nonetheless, computation & statistics  
for vineyards are well-studied.

# Computation

Cohen-Steiner et al 2006

For triangulation  $K$  of space  $X$  with  
monotone function  $f$ , recall our matrix

algorithm:

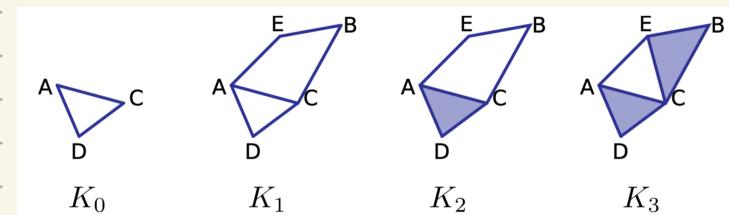
$$R = B$$

**for**  $j = 1 \dots m$  **do**

→ **while**  $\exists j' < j$  with  $low(j') = low(j)$  **do**  
add column  $j'$  to column  $j$   
**end while**  
**end for**

	A	C	D	AC	CD	AD	E	B	AE	BE	BC	ACD	CE	BCE
A														
C														
D														
AC														
CD														
AD														
E														
B														
AE														
BE														
BC														
ACD														
CE														
BCE														

$K_0$     $K_1$     $K_2$     $K_3$

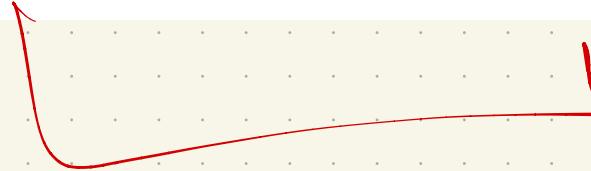
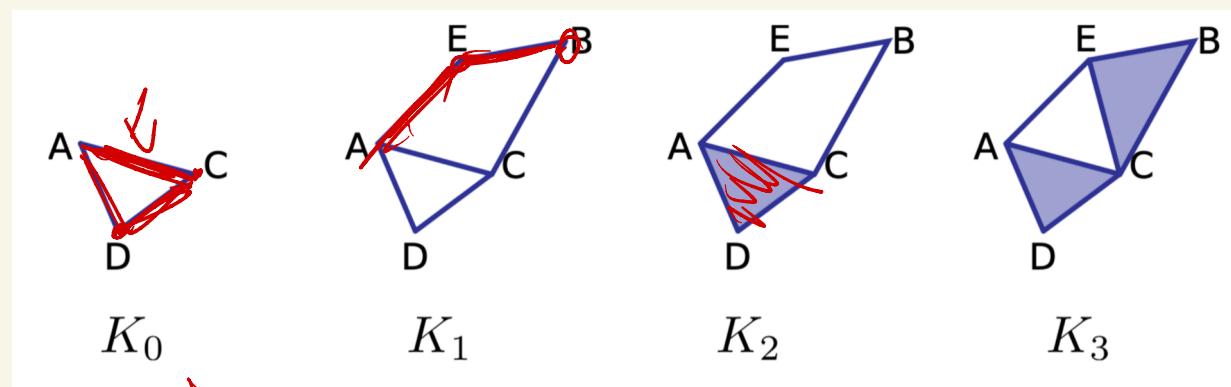


Result : pairs  $(\sigma_i, \sigma_j)$  with  
 $i = \text{lower}(j)$

	A	C	D	AC	CD	AD	E	B	AE	BE	BC	ACD	CE	BCE
A														
C	*													
D		*												
AC														
CD														
AD												*		
E							*							
B								*						
AE								*						
BE														
BC														
ACD														
CE													*	
BCE														

$\sigma_i$  positive  
 $\sigma_j$  negative

~~R~~  
 R



## Updating the pairing

Suppose order of 2 simplices flip:  
need to update persistence pairs  
(maybe).

Some linear algebra! can rewrite

$$R = \underline{DV}, \quad V \text{ invertible + upper-triangular}$$

$$\Rightarrow \underline{D} = R(V^{-1}) = RU$$

positive simplices  $\rightarrow$  zero columns  $\leftarrow$   
negative  $\rightarrow$  non-zero ↗

## Matrix cases

Sweeping  $i$  &  $i+1$  in filtration  
can cause change in birth/death pairs

	$k$	$l$		$k$	$l$		$i$	$i+1$		$i$	$i+1$	
$i$			$R$			$PRP$			$U$			$PUP$
$i+1$	1	1		1	1		1	1		1	1	

Figure 2: The transposition renders this particular  $R$  non-reduced and this particular  $U$  non-upper-triangular.

Worst case update cost:

$O(n)$  (at most the number of 1's in affected rows & columns)

## Initial application

Proteins : curve  $b: [0, 1] \rightarrow \mathbb{R}^3$

BBA5 :  $N=23$  amino acids

$m+1 = 201$  frames, where  $0 \leq i \leq m$ ,  
the snapshot is a curve  $b_i$

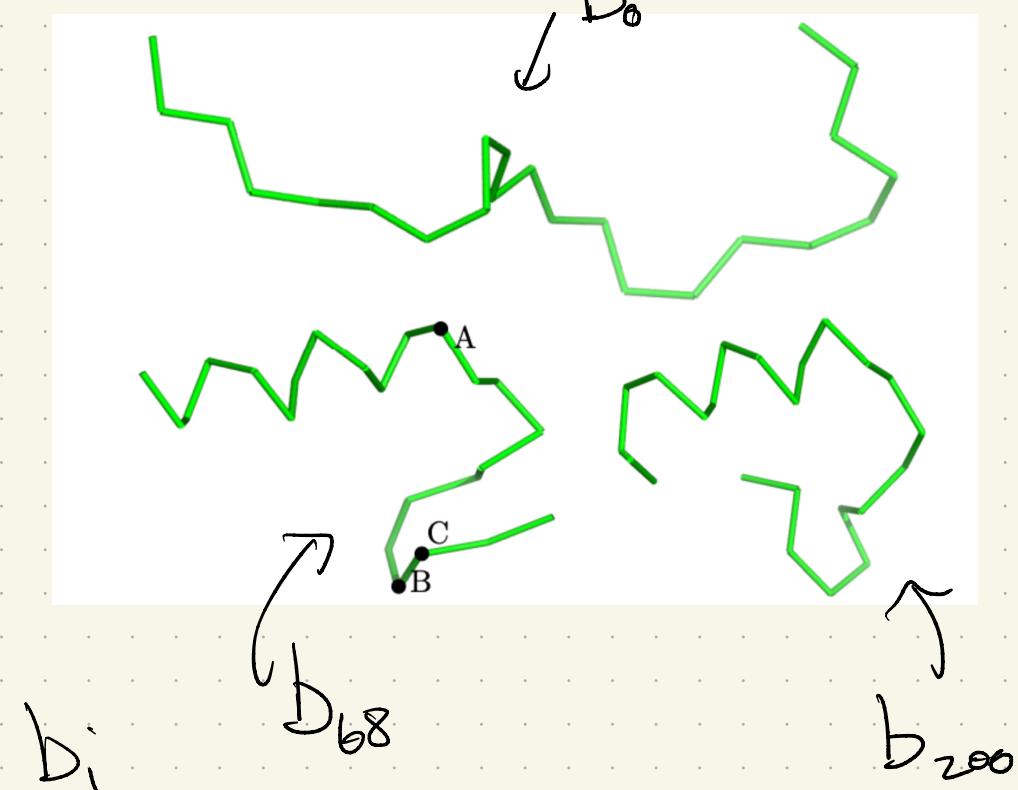
They define a  
pair wise distance

$[0, 1]^2 \rightarrow \mathbb{R}$  with

$$(r, s) \mapsto \|b(r) - b(s)\|$$

on each  $b_i$

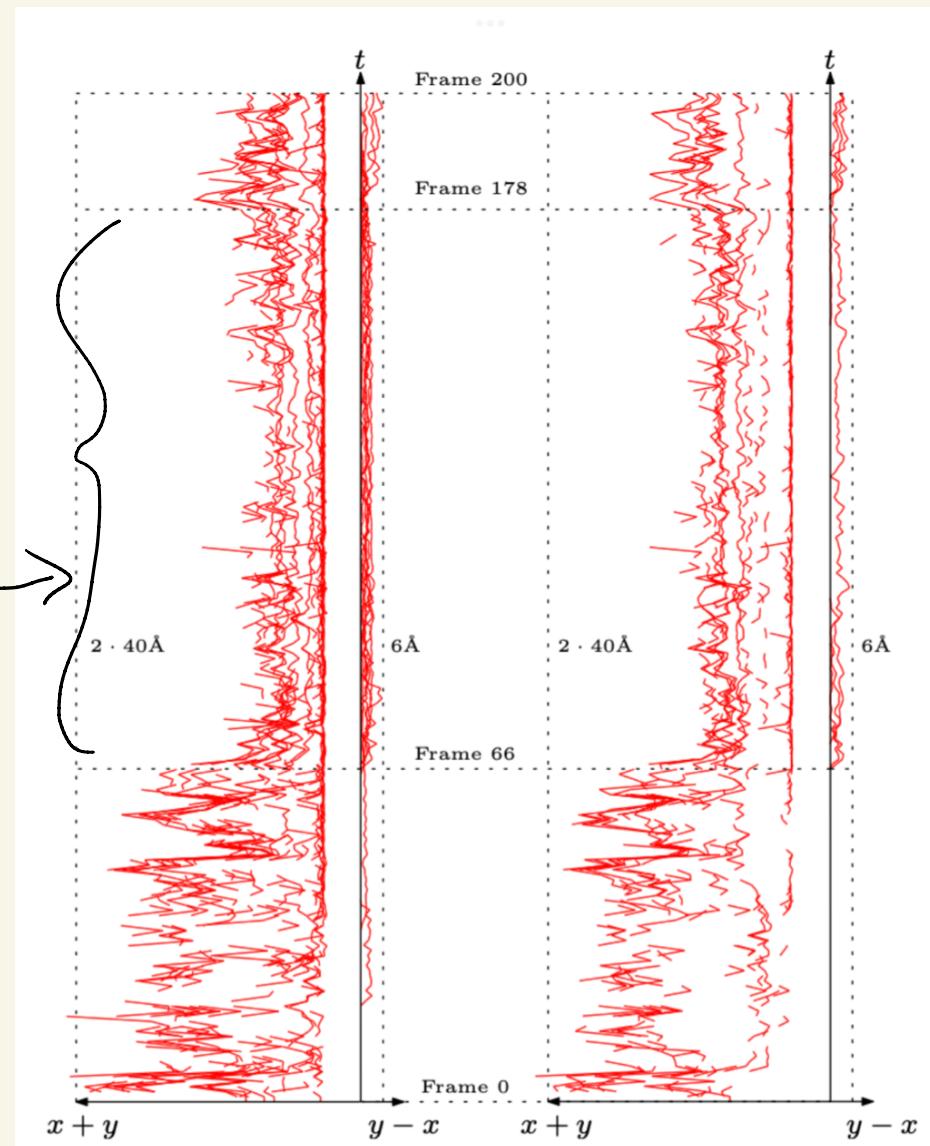
→ build filtration for  $b_i$



Then, build vineyards:

Some what hard to  
parse, but gives  
evidence of stages

2<sup>nd</sup> stage  
 $\alpha$ -helix forms  
& backbone is  
an S-shape  
[59 knees here]



**Figure 8:** The front view ( $x+y, t$ ) and the side view ( $y-x, t$ ) of the dimension 0 vineyard on the left and the dimension 1 vineyard on the right. The side views are simplified by removing vines with lifetime less than 20 frames.

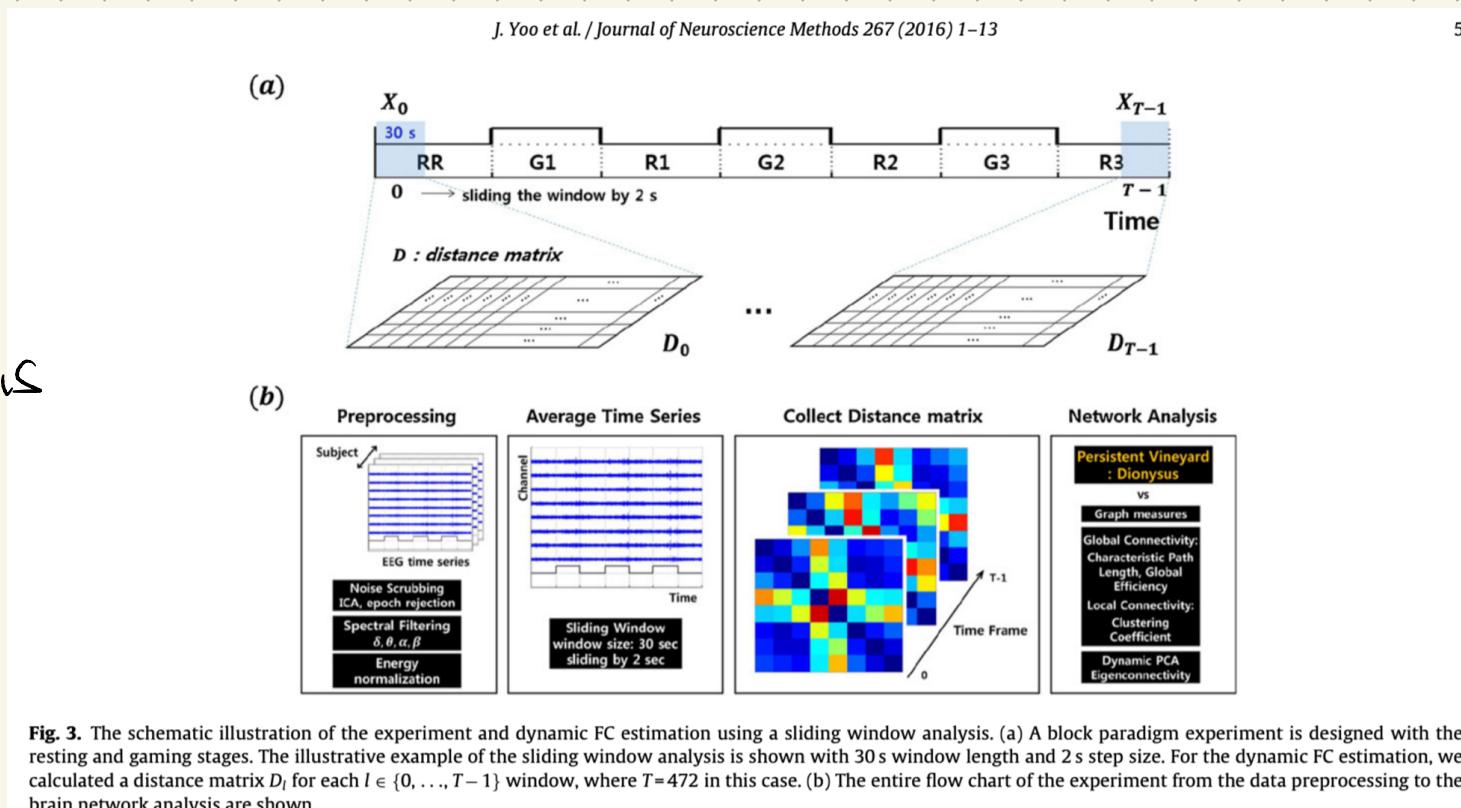
Another application

Brain connectivity

Yoo, Kim, Ahn, Ye 2016

Dynamic functional connectivity in  
resting and gaming stages  
EEG data of 54 students

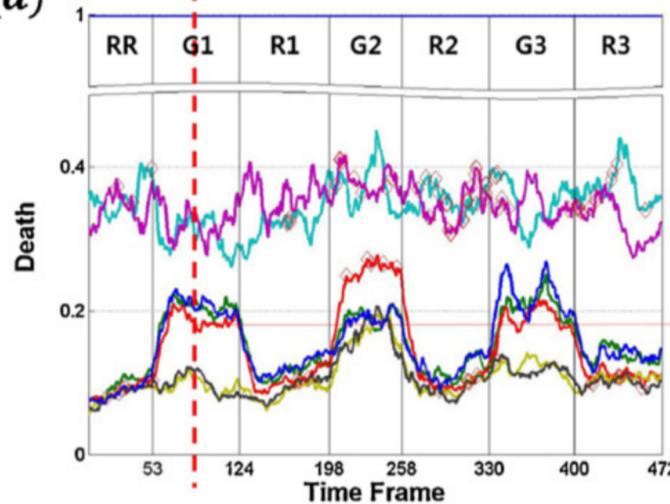
Using  
Dionysus  
tool



**Fig. 3.** The schematic illustration of the experiment and dynamic FC estimation using a sliding window analysis. (a) A block paradigm experiment is designed with the resting and gaming stages. The illustrative example of the sliding window analysis is shown with 30 s window length and 2 s step size. For the dynamic FC estimation, we calculated a distance matrix  $D_l$  for each  $l \in \{0, \dots, T-1\}$  window, where  $T=472$  in this case. (b) The entire flow chart of the experiment from the data preprocessing to the brain network analysis are shown.

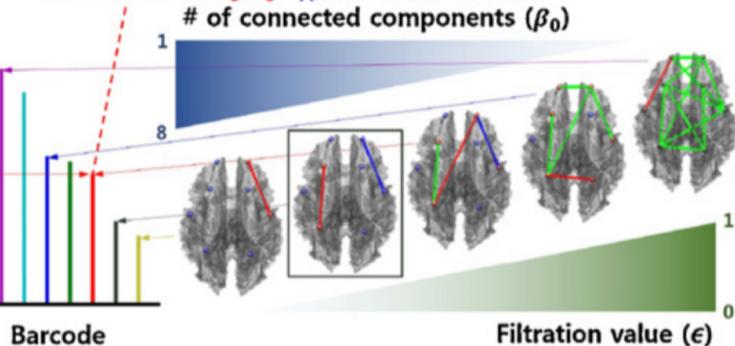
**Results:** Worked better than PCA + eigenconnectivity

(a)

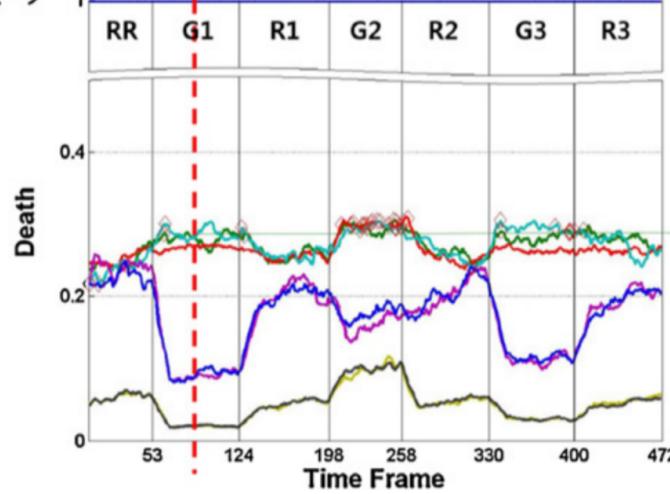


Persistent Homology Analysis at  $\dots$   
(Alpha band)

Two independent components in the box merge together to a single cluster as the connecting edge appears at this filtration value.

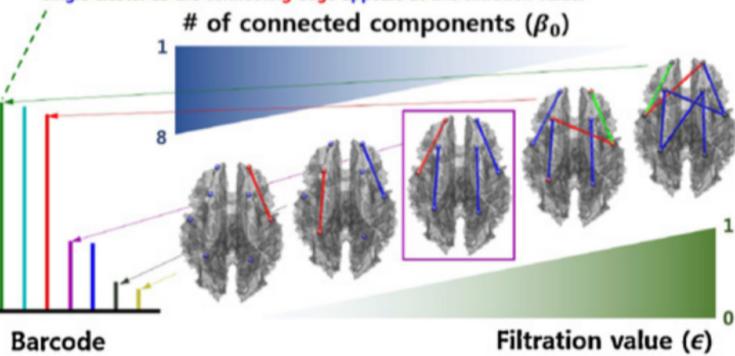


(b)



Persistent Homology Analysis at  $\dots$   
(Beta band)

Four independent components in the box finally merge together to a single cluster as the connecting edge appears at this filtration value.



**Crocker plots**: combining scale & time

Consider  $k^{\text{th}}$  Betti number  $B_k(\epsilon, t)$   
as a function of both time  $t$   
 $\epsilon$  in filtration bifurcations

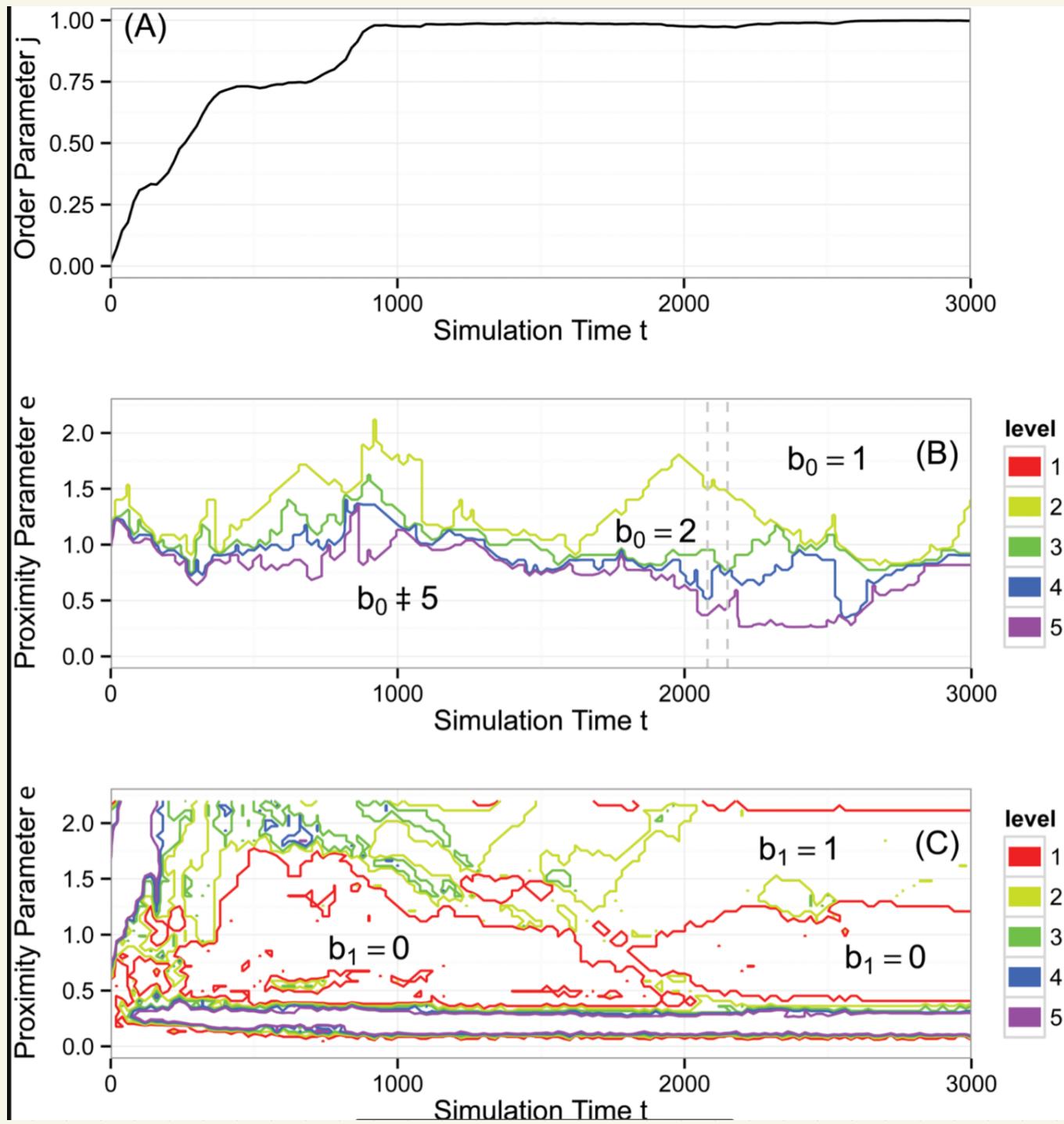
↳ build contour diagram

"Contour Realization of Computed  
K-dim hole Evolution in Rips complex"

Topaz, Ziegelmeier, Halverson 2015

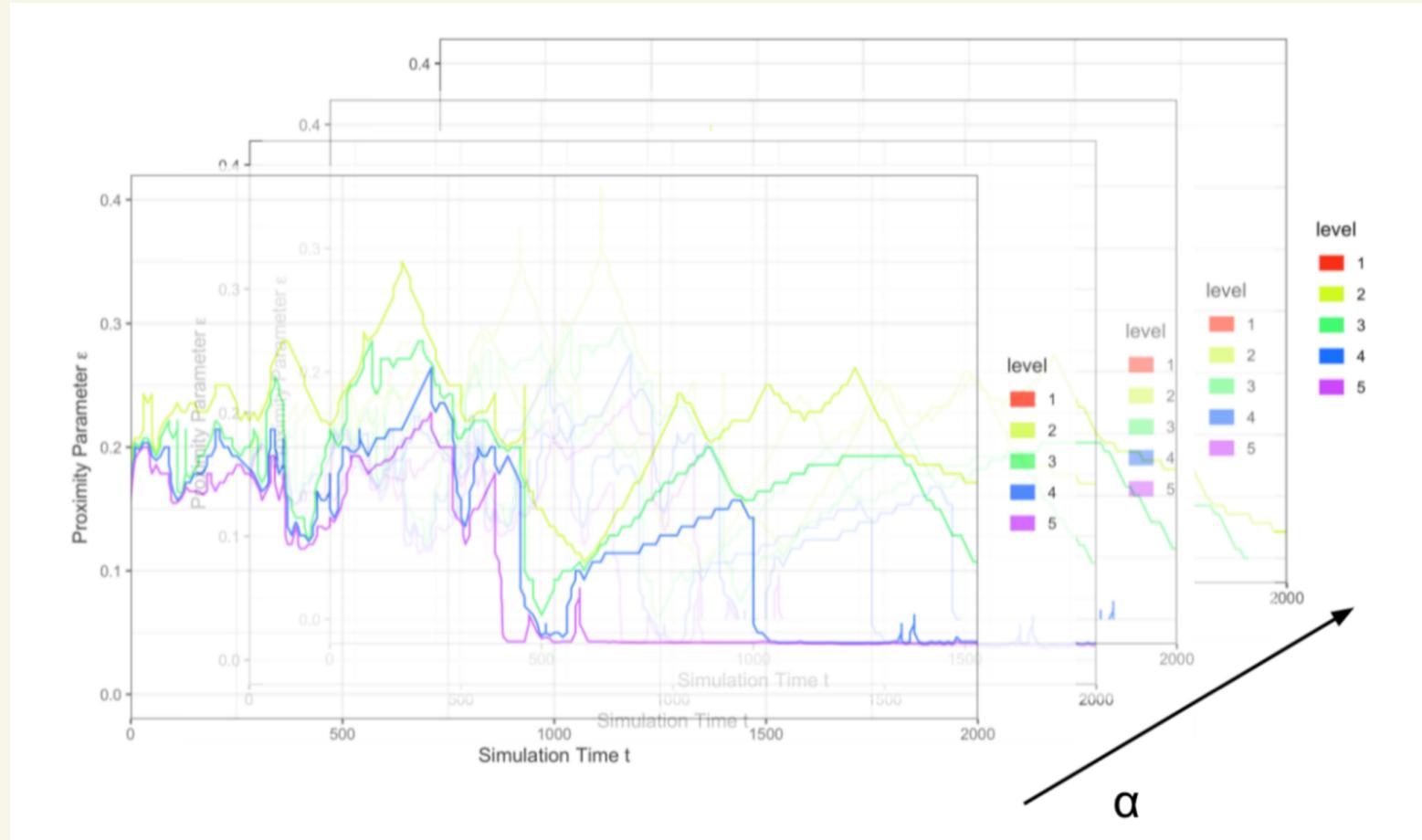


Breathing  
this  
down.



# Cracker Stacks

Xian, Adams, Topciz,  
et Ziegelmeier 2020



$\alpha$ : Smoothing parameter  
↳ can vectorize for ML models