

# Stable Signatures for Dynamic Graphs and Dynamic Metric Spaces via Zigzag Persistence

Facundo Mémoli  
joint with Woojin Kim and Zane Smith

The Ohio State University  
<https://research.math.osu.edu/networks/formigrams>.

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Woojin Kim and Facundo Memoli.

Stable signatures for dynamic graphs and dynamic metric spaces via zigzag persistence.

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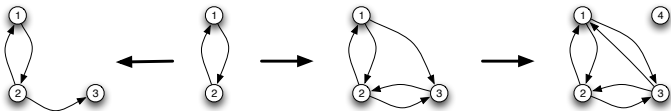


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# Motivations

- Study dynamic networks (social, financial, biological, ..)

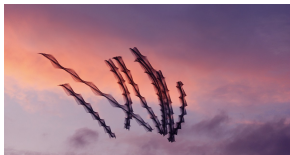


- Study/characterize flocking/swarming behavior in animals.



(wikipedia)

# Ornitographies, by Xavi Bau



# Characterization of flocking/herding behavior

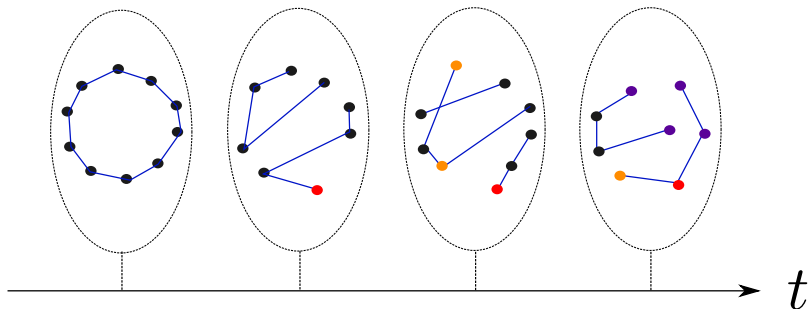
Video: Murmuration of Starlings in England

<https://www.youtube.com/watch?v=NREmtGhIHew>

# Dynamic graphs

**Proto-definition:** A *dynamic graph (DG)*  $\mathcal{G}_X$  over  $X$  is a graph that is subjected to a sequence of updates such as addition/deletion of vertices/edges.

$\mathcal{G}_X$



# Questions

- Q1. How do we **characterize** evolution of connected components of a dynamic graph (or dynamic simplicial complex)?

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- Q2. How do we **quantify** the **degree** of **difference** between **two** different dynamic graphs?

## Related work: TDA of dynamic data

- H. Edelsbrunner, J. Harer, A. Mascarenhas, V. Pascucci; (2004). *Time-varying Reeb graphs for continuous space–time data.*
- D. Cohen-Steiner, H. Edelsbrunner, D. Morozov; (2006). *Vines and vineyards by updating persistence in linear time.*
- E. Munch; (2013). *Applications of Persistent Homology to Time Varying Systems.*
- M. Hajij, B. Wang, C. Scheidegger, P. Rosen; (2017) *Visual Detection of Structural Changes in Time-Varying Graphs Using Persistent Homology.*

# Dynamic Graphs

**Definition:** A **dynamic graph**  $\mathcal{G}_X$  over the finite set  $X$  consists of a pair of maps as below satisfying four conditions:

$$V_X(\cdot) : \mathbb{R} \rightarrow \text{pow}(X) \quad \text{and} \quad E_X(\cdot) : \mathbb{R} \rightarrow \text{pow}(\text{pow}_2(V_X(\cdot))),$$

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<sup>1</sup> $\text{crit}(\mathcal{G}_X)$  is the union of the sets of discontinuity points of  $V_X(\cdot)$  and  $E_X(\cdot)$ .

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- 1 (Self-loops) For all  $t \in \mathbb{R}$ ,  $E_X(t) \supseteq \text{diag}(V_X(t) \times V_X(t))$ .

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- 4 (Comparability) for every  $t \in \mathbb{R}$ , it holds that

$$\mathcal{G}_X(t - \varepsilon) \subseteq \mathcal{G}_X(t) \supseteq \mathcal{G}_X(t + \varepsilon)$$

for all sufficiently small  $\varepsilon > 0$ .

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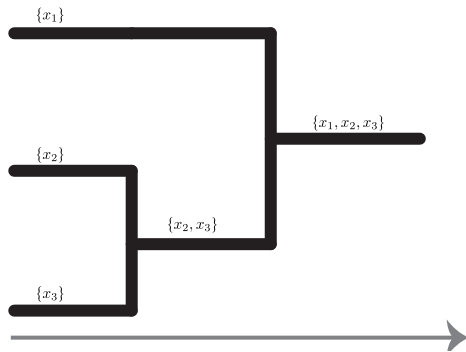
<sup>1</sup> $\text{crit}(\mathcal{G}_X)$  is the union of the sets of discontinuity points of  $V_X(\cdot)$  and  $E_X(\cdot)$ .

# Encoding evolution of connected components of a dynamic graph.

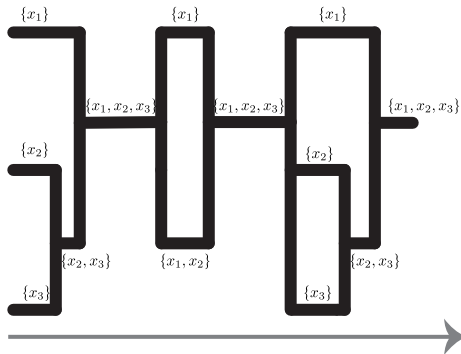
Want to be able to encode both **merging** and **splitting** of connected components of dynamic graphs.

# Dendrograms

Merging **only**



# Formigrams (Formicarium + Diagram)



Merging and **splitting**



(↑ Wikipedia)

(Formica = ant in latin.)

# Formigrams

A *formigram* over a finite set  $X$  is any function

$$\theta_X : \mathbb{R} \rightarrow \mathcal{P}^{\text{sub}}(X)$$

such that:

- 1 (Tameness) the set  $\text{crit}(\theta_X)$  of points of discontinuity of  $\theta_X$  is locally finite. We call the elements of  $\text{crit}(\theta_X)$  *the critical points of  $\theta_X$* .
- 2 (Interval lifespan) for every  $x \in X$ , the set  $I_x := \{t \in \mathbb{R} : x \in B \in \theta_X(t)\}$ , said to be the *lifespan of  $x$* , is a non-empty closed interval,
- 3 (Comparability) for every point  $c \in \mathbb{R}$  it holds that  $\theta_X(c - \varepsilon) \leq \theta_X(c) \geq \theta_X(c + \varepsilon)$  for all sufficiently small  $\varepsilon > 0$ .<sup>2</sup>

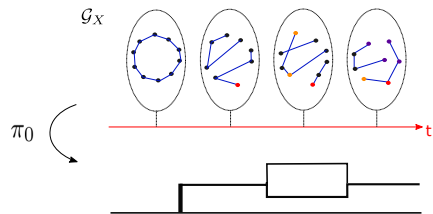
<sup>2</sup>If  $\theta_X$  is discontinuous at  $c$ , then at least one of the relations of  $\theta_X(c - \varepsilon) \leq \theta_X(c) \geq \theta_X(c + \varepsilon)$  will be strict for small  $\varepsilon > 0$ . Otherwise,  $\theta_X(c - \varepsilon) = \theta_X(c) = \theta_X(c + \varepsilon)$  for small  $\varepsilon > 0$ .

# Dynamic Graph $\rightarrow$ Formigram

Recall the  $\pi_0$  functor: If  $G_X = (X, E_X)$  is a graph on the vertex set  $X$ , then  $\pi_0(G_X) :=$  Partition of  $X$  into connected components.

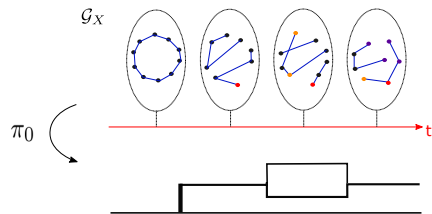
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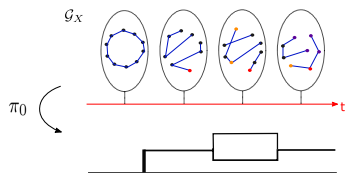
**Proposition.** For any  $\mathcal{G}_X$ ,

$$\pi_0(\mathcal{G}_X) : \mathbb{R} \rightarrow \mathcal{P}^{sub}(X)$$

is a formigram.

Here  $\mathcal{P}^{sub}(X)$  is the set of all sub-partitions of  $X$ .

# Remarks about formigrams.



## Remarks.

- Formigrams generalize dendrograms:

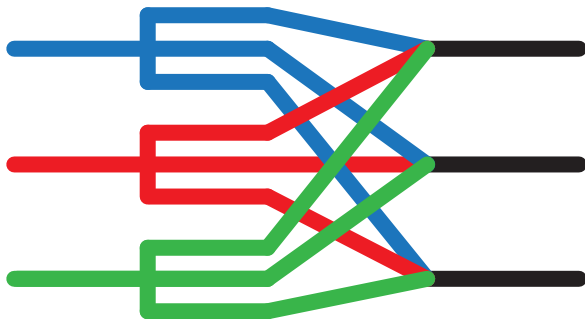
Standard simplicial filtration  $\rightarrow$  Dendrogram

**Zigzag** simplicial filtration  $\rightarrow$  Formigram<sup>3</sup>

- Any formigram has an underlying (Reeb) graph.
- Underlying graph is **not** always planar.

<sup>3</sup>A dynamic graph is an 1-dim zigzag simplicial filtration.

Non-planar formigram:<sup>4</sup>



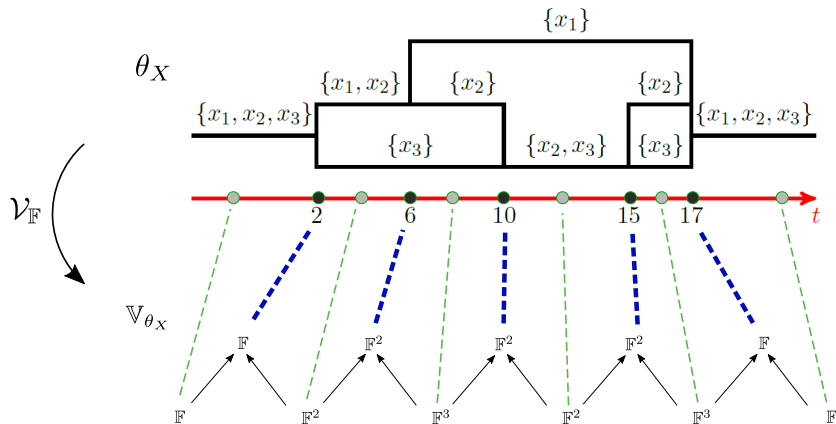
⇒ We want further simplification step to facilitate:

1. **simpler visualization**, and
2. **reduced computational cost of comparison**.

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<sup>4</sup>By Z. Smith.

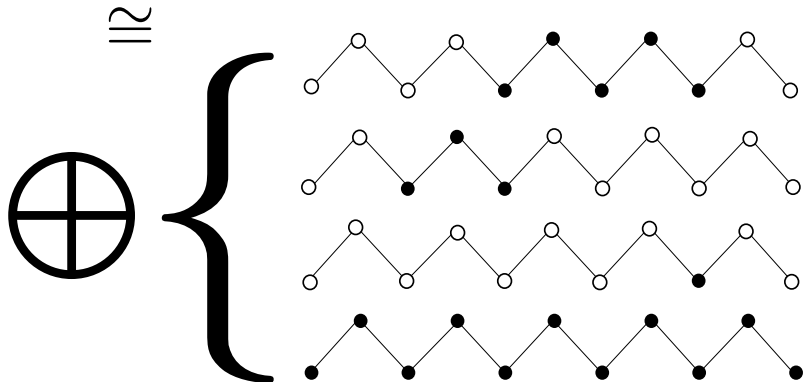
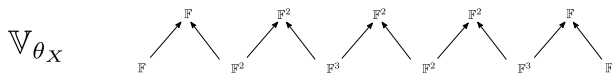
# Simplification: Formigrams $\Rightarrow$ Zigzag modules

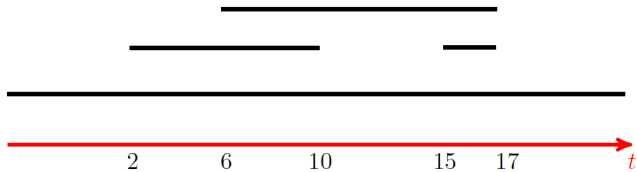
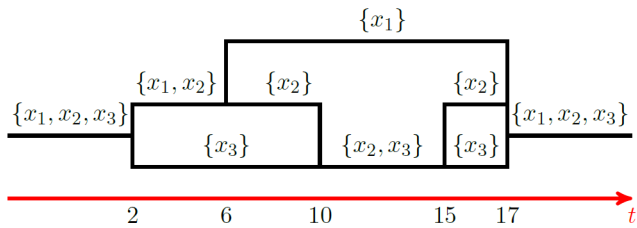


$\mathcal{V}_F$  is the free functor.

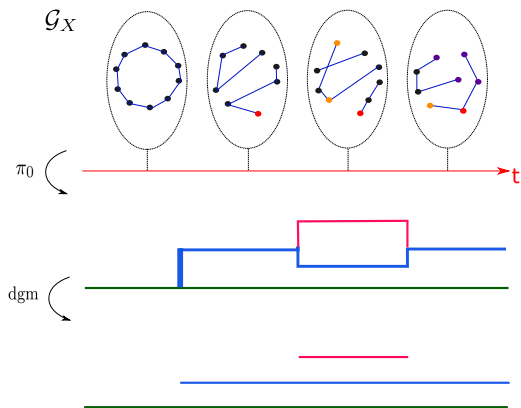
# Simplification: Zigzag modules $\Rightarrow$ Barcodes

Gabriel 1972, de Silva-Carlsson 2010

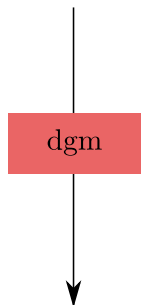




# Summary



Dynamic Graph



Barcode

# Questions (again)

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Q1+. Is the process  $\text{Dynamic Graph} \xrightarrow{\text{dgm}} \text{Barcode}$  **stable**?

Q2. How do we **quantify** the **degree** of **difference** between **two** different dynamic graphs?

# Stability Theorem (Kim-M 2018) [KM17]

We construct  $d_I^{\text{dynG}}$ , a metric on DGs and prove:

**Theorem.** For any two dynamic graphs  $\mathcal{G}_X$  and  $\mathcal{G}_Y$ ,

$$2 d_I^{\text{dynG}}(\mathcal{G}_X, \mathcal{G}_Y) \geq d_B(\text{dgm}(\mathcal{G}_X), \text{dgm}(\mathcal{G}_Y)).$$

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Structure of  $d_I^{\text{dynG}}$  is some sort of hybrid of:

- the interleaving distance between Reeb graphs (de Silva, Munch, Patel 2015), and
- the Gromov-Hausdorff distance.

# Interconnecting dynamic graphs via correspondences

Let  $\mathcal{G}_X = (V_X(\cdot), E_X(\cdot))$  and  $\mathcal{G}_Y = (V_Y(\cdot), E_Y(\cdot))$  be two dynamic graphs, and let  $R$  be a correspondence between  $X$  and  $Y$ .

By  $\mathcal{G}_X \xrightarrow{R} \mathcal{G}_Y$ , we mean:

For any  $(x, y), (x', y') \in R$  and any  $t \in \mathbb{R}$ ,

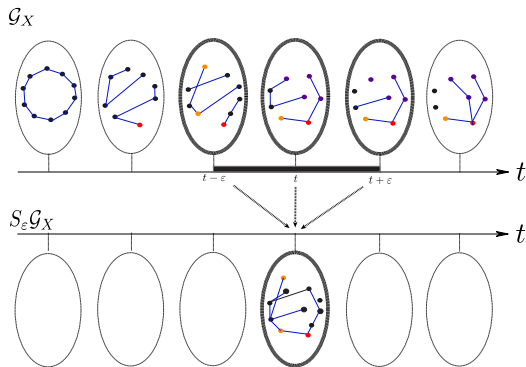
$$(x, x') \in E_X(t) \Rightarrow (y, y') \in E_Y(t).$$

(This implies that if  $x \in V_X(t)$ , then  $y \in V_Y(t)$ .)

# $S_\varepsilon$ : $\varepsilon$ -smoothing of dynamic graphs

Let  $\varepsilon \geq 0$ . For  $t \in \mathbb{R}$ ,

$$S_\varepsilon \mathcal{G}_X(t) := \left( \bigcup_{s \in [t]^\varepsilon} V_X(s), \bigcup_{s \in [t]^\varepsilon} E_X(s) \right).$$

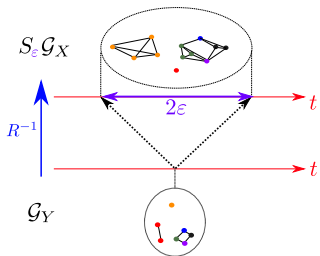
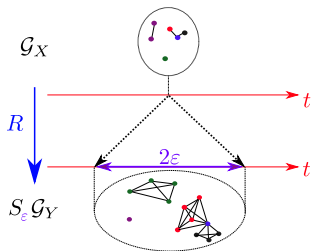


**Proposition.**  $S_\varepsilon$  maps DGs to DGs.

# $d_I^{\text{dynG}}$ for dynamic graphs

**Definition.** Given  $\mathcal{G}_X$  and  $\mathcal{G}_Y$  be dynamic graphs, define:

$$d_I^{\text{dynG}}(\mathcal{G}_X, \mathcal{G}_Y) := \inf \left\{ \varepsilon \geq 0 : \exists R, \mathcal{G}_X \xrightarrow{R} \mathcal{S}_\varepsilon \mathcal{G}_Y \text{ and } \mathcal{G}_Y \xrightarrow{R^{-1}} \mathcal{S}_\varepsilon \mathcal{G}_X \right\}.$$



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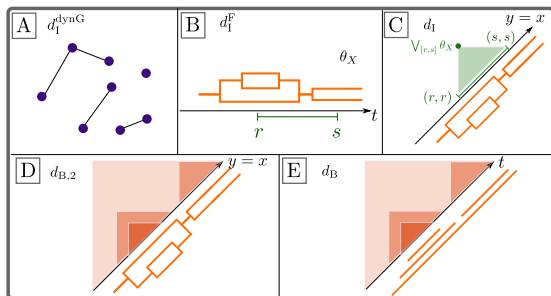
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The proof uses ideas and results from the recent papers by Botnan-Lesnicks, and by Bjerkevik.



# Flocking



(Xavi Bau)

# Flocking: Questions

- Q1. How do we **characterize** flocking behavior?
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- Very large literature in biology.
  - C.Topaz, L. Ziegelmeier, T. Halverson; (2015). *Topological Data Analysis of Biological Aggregation Models*.
  - P. Corcoran and C. B. Jones; (2017). *Modeling Topological Features of Swarm Behavior in Space and Time With Persistence Landscapes*.

# Dynamic Metric Spaces

A **dynamic metric space** is a pair  $(X, d_X(\cdot))$  where  $X$  is finite set and  $d_X(\cdot) : \mathbb{R} \times X \times X \rightarrow \mathbb{R}_{\geq 0}$  is such that

- $\forall t \in \mathbb{R}, d_X(t) : X \times X \rightarrow \mathbb{R}_{\geq 0}$  is a pseudo-metric on  $X$ .
- $\forall x, x' \in X, t \mapsto d_X(t)(x, x')$  is continuous and *tame*.

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**Notation:**  $\gamma_X = (X, d_X(\cdot))$ ;  $\mathcal{M}^{\text{dyn}}$  is the collection of all DMSs.

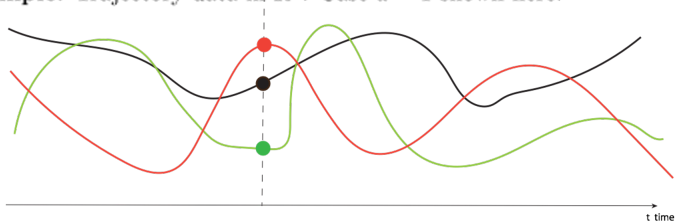
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**Example:** Trajectory data in  $\mathbb{R}^d$ . Case  $d = 1$  shown here:

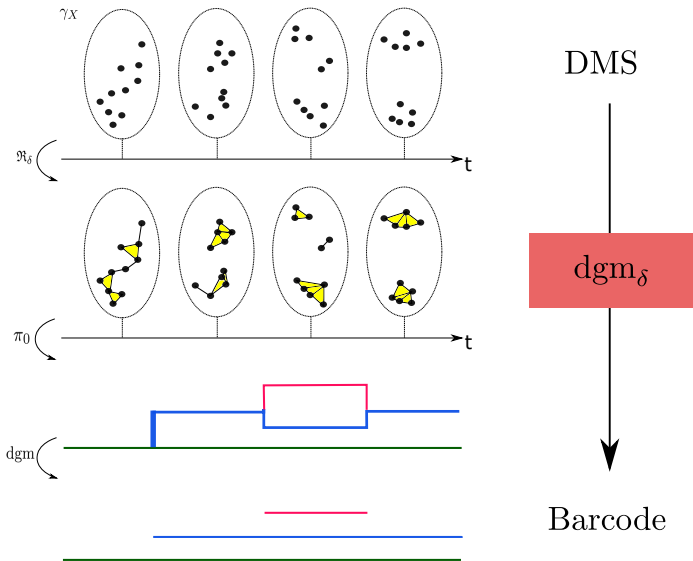


# Encoding clustering behavior of DMSs.

Fix  $\delta \geq 0$ . Given a metric space  $(X, d_X)$ , let  $\mathcal{R}_\delta^1(X, d_X)$  be the 1-skeleton of the Rips complex  $\mathcal{R}_\delta(X, d_X)$ .

**Proposition.** Let  $\gamma_X = (X, d_X(\cdot))$  be a DMS. Then,  $\mathcal{R}_\delta^1(\gamma_X)$  is a dynamic graph.

The map  $\text{dgm}_\delta : \mathcal{M}^{\text{dyn}} \rightarrow \text{Barcodes}$ .



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We introduce a family  $\{d_{I,\lambda}^{\text{dynM}}\}_{\lambda \geq 0}$  of metrics on DMSs. Here  $\lambda$  is a slack parameter.

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**Corollary** (Formigram barcodes are stable). For DMSs  $\gamma_X$  and  $\gamma_Y$ ,

$$2 d_I^{\text{dynM}}(\gamma_X, \gamma_Y) \geq \sup_{\delta \geq 0} d_B(\text{dgm}_\delta(\gamma_X), \text{dgm}_\delta(\gamma_Y)),$$

where  $\text{dgm}_\delta(\gamma_X) = \text{dgm} \circ \mathcal{R}_\delta^1(\gamma_X)$ .

What about  $(\delta, t)$ -persistence? SLHC of DMSs

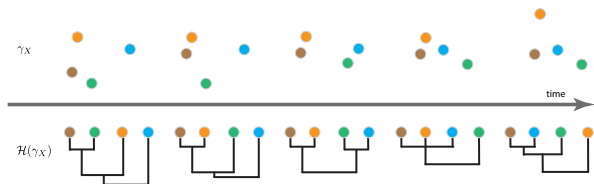
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Given a DMS  $\gamma_X$ , one obtains an ultrametric DMS  $\mathcal{H}(\gamma_X)$ .

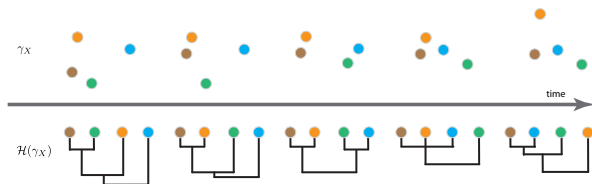


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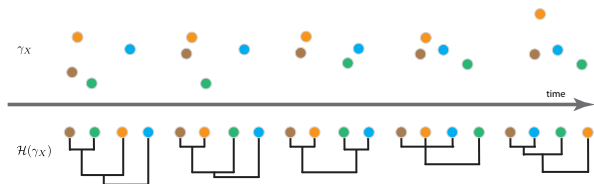
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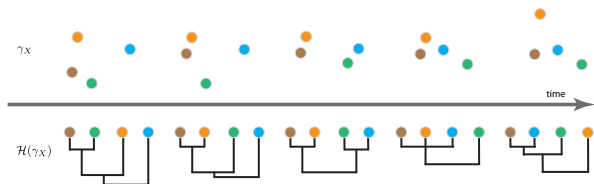
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**Theorem** (Kim-M, 2018). For  $\ell > 0$ , let  $\gamma_X$  and  $\gamma_Y$  be any  $\ell$ -Lipschitz DMSs. Then for any  $\lambda > 0$ ,

$$\lambda + \ell \cdot d_{\text{dyn}}(\gamma_X, \gamma_Y) \geq d_{\text{dyn}}(\mathcal{H}(\gamma_X), \mathcal{H}(\gamma_Y))$$

## $\varepsilon$ -smoothing of DMSs

For  $\varepsilon \geq 0$ , consider  $\varepsilon$ -smoothing  $S_\varepsilon : \mathcal{M}^{\text{dyn}} \rightarrow \mathcal{M}^{\text{dyn}}$  s.t.

$$S_\varepsilon : \quad \gamma_X = (X, d_X(\cdot)) \longmapsto \gamma_X^\varepsilon = (X, d_X^\varepsilon(\cdot))$$

where

$$d_X^\varepsilon(t)(x, x') = \min_{s \in [t]^\varepsilon} d_X(s)(x, x')$$

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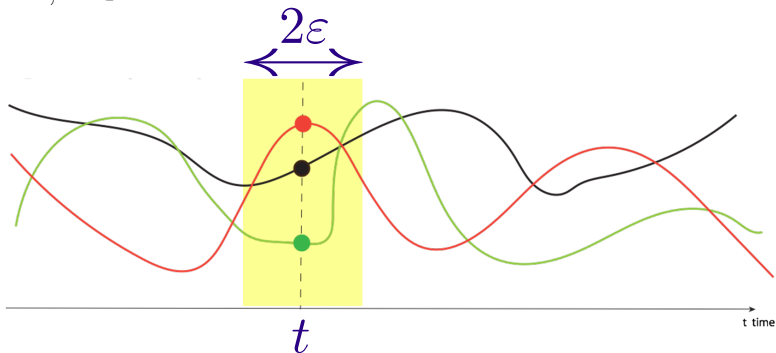
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## Definition

Two DMSs  $\gamma_X, \gamma_Y$  are  $(\lambda, \varepsilon)$ -interleaved when there is a correspondence  $R \subseteq X \times Y$  s.t.  $\forall t \in \mathbb{R}, \forall (x, y), (x', y') \in R$

- $d_X^\varepsilon(t)(x, x') \leq d_Y(t)(y, y') + \lambda \cdot \varepsilon$
- $d_Y^\varepsilon(t)(y, y') \leq d_X(t)(x, x') + \lambda \cdot \varepsilon$

Whenever this happens we write  $\gamma_X \approx_{(\lambda, \varepsilon)} \gamma_Y$ . Define:

$$d_{I,\lambda}^{\text{dyn}}(\gamma_X, \gamma_Y) := \inf \{ \varepsilon \geq 0 : \gamma_X \approx_{(\lambda, \varepsilon)} \gamma_Y \}.$$

**Remark.** For  $\lambda \in [0, +\infty)$ ,  $d_{I,\lambda}^{\text{dyn}}$  can be  $+\infty$ .

**Remark.**  $\lambda \leq \lambda' \quad \Rightarrow \quad d_{I,\lambda}^{\text{dyn}} \geq d_{I,\lambda'}^{\text{dyn}}$ .

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**Theorem.** For  $\lambda \geq 0$ ,  $d_{I,\lambda}^{\text{dyn}}$  is an **extended metric** on  $\mathcal{M}^{\text{dyn}}$ .

**Theorems.** For  $\lambda > 0$ ,

- $d_{I,\lambda}^{\text{dyn}}$  is a **metric** on **bounded** DMSs.
- $d_{I,\lambda}^{\text{dyn}}$  recovers the Gromov-Hausdorff distance between constant DMSs (up to a constant factor).<sup>5</sup>
- The metrics  $d_{I,\lambda}^{\text{dyn}}$  for different  $\lambda > 0$  are bilipschitz equivalent.

---

<sup>5</sup>This implies that Stability Theorem for SLHC of DMS is proper generalization of SLHC for standard metric spaces.

**Theorem.** There is an **isometric embedding**  $\Delta : \mathcal{M} \hookrightarrow \mathcal{M}^{\text{dyn}}$ , i.e.

$$d_{\text{GH}}(X, Y) = d_{\text{I}}^{\text{dyn}}(\Delta(X), \Delta(Y)).$$

Recall

**Corollary.** For any DMSs  $\gamma_X$  and  $\gamma_Y$ ,

$$2 d_{\text{I}}^{\text{dyn}}(\gamma_X, \gamma_Y) \geq \sup_{\delta \geq 0} d_{\text{B}}(\text{dgm}_{\delta}(\gamma_X), \text{dgm}_{\delta}(\gamma_Y))$$

LHS is **NP-hard** whereas the RHS is **poly time**.

## Summary: Stability theorems for DGs and DMSs

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**Q. What about higher-dimensional homological features?**

## Q. Extension to high-dimensional PH?

Any DG  $\mathcal{G}_X$  induces a ZZ simplicial filtration (and its barcode) for **each** dimension  $k \in \mathbb{Z}_+$ .

For every  $t$ , just consider the associated *clique complex*  $\mathcal{C}(\mathcal{G}_X(t))$ .

# Bad news..

**Theorem (Kim-M).** For each integer  $k \geq 1$ , there exists a finite set  $X_k$  and a pair of DGs  $\mathcal{G}_{X_k}$  and  $\mathcal{G}'_{X_k}$  over  $X_k$  such that

$$d_I^{\text{dynG}}(\mathcal{G}_{X_k}, \mathcal{G}'_{X_k}) < \infty,$$

but

$$d_B\left(\text{dgm}(H_k(\mathcal{C}(\mathcal{G}_{X_k}))), \text{dgm}(H_k(\mathcal{C}(\mathcal{G}'_{X_k})))\right) = \infty.$$

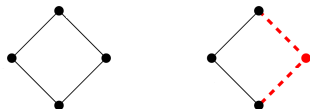
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These two DGs are finitely interleaved but the bottleneck distance between their associated  $H_1$  barcodes is  $\infty$ .

# Application: classifying flocking behaviors [KMS]

- a variation of Boids simulation (parameters: Cohesion, Alignment, Separation, etc.)
- Consider 4 behaviors by tuning parameters.
- Repeat simulation 20 times for each behavior with random initial configurations.
- Compute clustering barcodes (Morozov's Dionysus).
- Construct  $80 \times 80$ -matrix of bottleneck distances. Carry out hierarchical clustering and MDS plot of it.

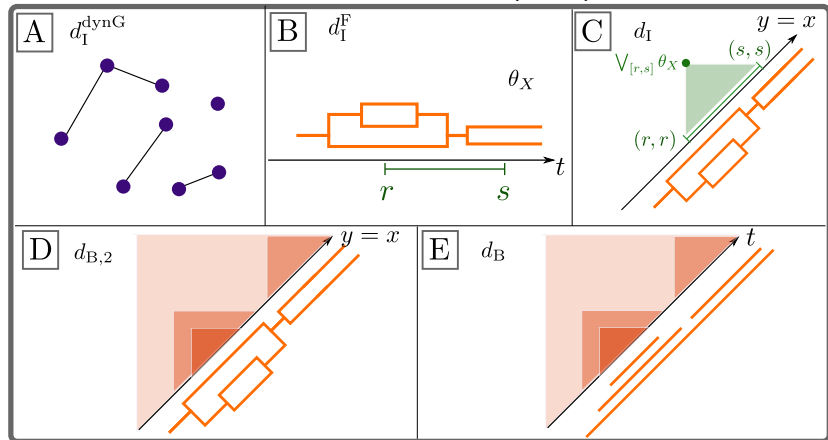
<https://research.math.osu.edu/networks/formigrams>

# Discussion

- Extensions: dynamic networks.
- Extensions: dynamical systems?
- $\gamma_X \mapsto \text{PH}_{k \geq 1}(\gamma_X)$  stably?

# Proof of Stability Theorem

Inspired by M. Botnan and M. Lesnick (2016)



# Tameness

A continuous map  $f : \mathbb{R} \rightarrow \mathbb{R}$  is **tame** if for all  $r \in \mathbb{R}$ , the closed set  $f^{-1}(r)$  has only finitely many connected components.

A DMS  $\gamma_X = (X, d_X(\cdot))$  is **tame** if for every  $x, x' \in X$ ,

$$d_X(\cdot)(x, x') : \mathbb{R} \rightarrow \mathbb{R}$$

is tame.

# Example of $d_I^{\text{dyn}} < \infty$ .

- $X := \{x, x'\}$ .  $\gamma_X := (X, d_X(\cdot))$ .  $\overline{\gamma}_X := (X, \overline{d}_X(\cdot))$ .

$$d_X(t) = 1 + \cos(t)$$

$$\overline{d}_X(t) = 1 + \cos(t + \tau) \text{ for some } \tau \in (0, 2\pi).$$

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$$d_I^{\text{dyn}}(\gamma_X, \overline{\gamma}_X) = \min(\tau, 2\pi - \tau) \leq \tau$$