

# Opinion Dynamics over Cellular Sheaves

Enhancing the expressivity of time diffusion processes over graphs

*Social Networks and Online Markets*

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The purpose of my final project in the "*Social Networks and Online Markets*" course was to address opinion dynamics modeling from an algebraic topological perspective. The main reference is the following paper:

- *Opinion Dynamics on Discourse Sheaves*, Jacob Hansen, Robert Ghrist, 2020



# Table of Contents

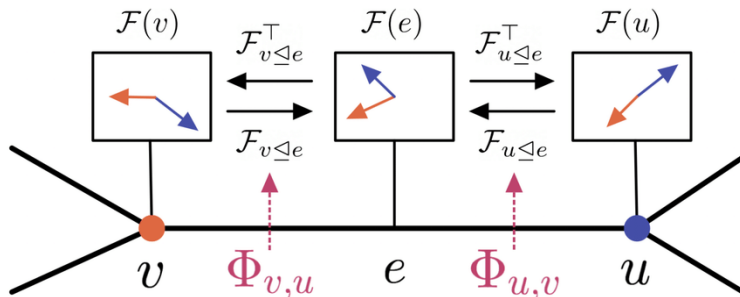
## 1 A short introduction to Graph Cellular Sheaves

- ▶ A short introduction to Graph Cellular Sheaves
- ▶ Opinion dynamics over discourse sheaves
- ▶ Advanced opinion dynamics modeling
- ▶ Beyond graph modeling: group opinion dynamics over simplicial complexes



# Why should we incorporate sheaves in networking?

## 1 A short introduction to Graph Cellular Sheaves



Cellular sheaves introduce **geometric structures** over a certain combinatorial topology, enhancing the expressivity and the possibilities of **learning and representation problems**



# The key concepts - The structure of a cellular sheaf

## 1 A short introduction to Graph Cellular Sheaves

### Definition (Graph cellular sheaf)

Considering a graph  $G(V, E)$ , we define a **graph cellular sheaf**  $\mathcal{F}$  specified by:

- A vector space  $\mathcal{F}(v)$  for each node  $v \in V$  (**stalk** over a vertex);
- A vector space  $\mathcal{F}(e)$  for each edge  $e \in E$  (**stalk** over a edge);
- A linear map  $\mathcal{F}_{v \triangleleft e} : \mathcal{F}(v) \rightarrow \mathcal{F}(e)$  for each incidence relation  $v \triangleleft e$  for each edge (**restriction maps**).

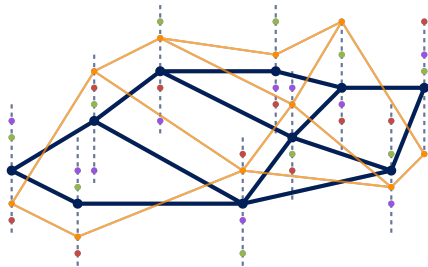


# The key concepts - Information encoding

## 1 A short introduction to Graph Cellular Sheaves

The direct sum of the stalks are called **spaces of cochains**. In particular:

- 0-cochains  $x$  belong to  $\mathcal{C}^0(G, \mathcal{F}) = \bigoplus_{v \in V} \mathcal{F}_v$ ;
- 1-cochains  $\xi$  belong to  $\mathcal{C}^1(G, \mathcal{F}) = \bigoplus_{e \in E} \mathcal{F}_e$ .





# The key concepts - The sheaf Laplacian

## 1 A short introduction to Graph Cellular Sheaves

The **coboundary map** is the linear operator  $\delta : \mathcal{C}^1(G, \mathcal{F}) \rightarrow \mathcal{C}^0(G, \mathcal{F})$  defined as follows, up to an arbitrary orientation of the restriction maps:

$$(\delta^T x)_e = \mathcal{F}_{v \triangleleft e} x_v - \mathcal{F}_{u \triangleleft e} x_u \quad (1)$$

The **sheaf Laplacian** is the operator defined as follows:

$$L_{\mathcal{F}} = \delta \delta^T : \mathcal{C}^0(G, \mathcal{F}) \rightarrow \mathcal{C}^0(G, \mathcal{F}) \quad (2)$$



# The key concepts - Dirichlet Energy over edges

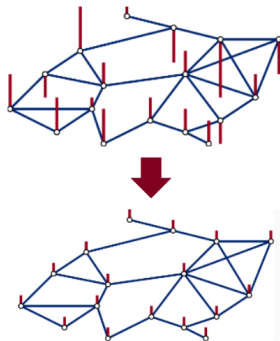
## 1 A short introduction to Graph Cellular Sheaves

A classic result in network science is that we can quantify the network "disagreement" using the quadratic form in the Laplacian:

$$x^T L x = \sum_{i,j \in V} A[i,j] (x_i - x_j)^2 \quad (3)$$

The Laplacian of a graph cellular sheaf supports a generalization of such a concept:

$$x^T L_{\mathcal{F}} x = \sum_{e \in E} \|\mathcal{F}_{v \triangleleft e} x_v - \mathcal{F}_{u \triangleleft e} x_u\|^2 \quad (4)$$







# Table of Contents

## 2 Opinion dynamics over discourse sheaves

- ▶ A short introduction to Graph Cellular Sheaves
- ▶ **Opinion dynamics over discourse sheaves**
- ▶ Advanced opinion dynamics modeling
- ▶ Beyond graph modeling: group opinion dynamics over simplicial complexes



## The reference model - $\mathbb{R}^n$ constant sheaf

### 2 Opinion dynamics over discourse sheaves

In a  $\mathbb{R}^n$  constant sheaf every stalk is  $\mathbb{R}^n$  and every restriction maps is the identity map. The related discourse sheaf is such that:

- Each agent residing on each node has a *private opinion* being a vector in  $\mathbb{R}^n$ : each component of this vector represents an opinion on a certain topic among  $n$  topics;
- Each edge represent a pairwise communication, leading to an *expressed opinion* over the same  $n$  topics;
- The restriction maps are all set to the identity maps  $\mathbb{I}_n$ : this reconnects to classic approaches where private opinions are expressed without modifications.



# Laplacian heat equation: from graphs to sheaves

## 2 Opinion dynamics over discourse sheaves

For a sheaf  $\mathcal{F}$  defined over a graph and a 0-cochain  $x$ , the dynamic

$$\frac{dx}{dt} = -\alpha L_{\mathcal{F}} x \quad (5)$$

has solutions equal to

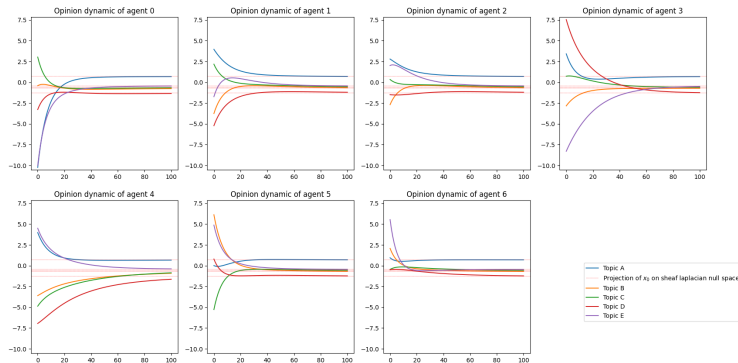
$$x(t) = e^{-\alpha L_{\mathcal{F}} t} x_0 \quad (6)$$

and converges to the projection of the initial state  $x_0 \in \mathcal{C}^0(G, \mathcal{F})$  over the null space of the Laplacian  $L$ , in a way that is completely analogous to the classic graph consensus dynamic.  $\ker(L_{\mathcal{F}})$  is called **0th-cohomology**, or **space of global sections**, i.e. 0-cochains satisfying  $\mathcal{F}_{v \triangleleft e} x_v = \mathcal{F}_{u \triangleleft e} x_u$  for each edge.



# Laplacian heat equation: from graphs to sheaves

## 2 Opinion dynamics over discourse sheaves

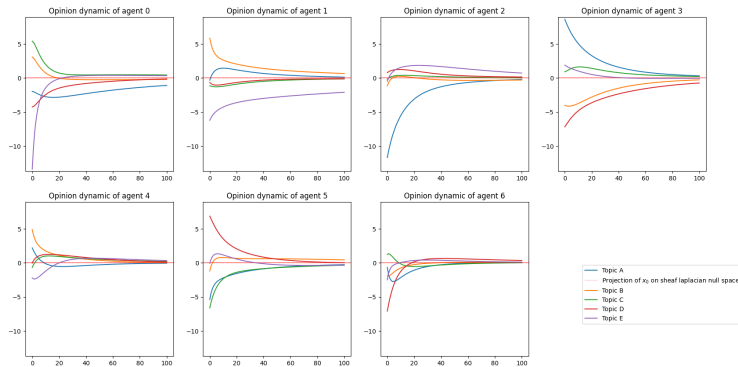


**Figure:** Opinion trajectories for a constant  $\mathbb{R}^n$  sheaf defined over a network of 7 agents and a  $n = 5$  topics basis and Gaussian distributed initial opinions



# Laplacian heat equation: from graphs to sheaves

## 2 Opinion dynamics over discourse sheaves



**Figure:** Opinion trajectories for a random sheaf defined over a network of 7 agents and a  $n = 5$  topics basis: the consensus is trivial even if the underlying graph is connected



# Table of Contents

## 3 Advanced opinion dynamics modeling

- ▶ A short introduction to Graph Cellular Sheaves
- ▶ Opinion dynamics over discourse sheaves
- ▶ Advanced opinion dynamics modeling**
- ▶ Beyond graph modeling: group opinion dynamics over simplicial complexes



# Modeling stubborn agents

## 3 Advanced opinion dynamics modeling

A first variation we can introduce is to define  $U$ -restricted dynamics, i.e. considering a set  $U$  of agents which do not modify their opinion:

$$\frac{dx}{dt} = -\alpha \Pi_U(L_{\mathcal{F}}x) \quad (7)$$

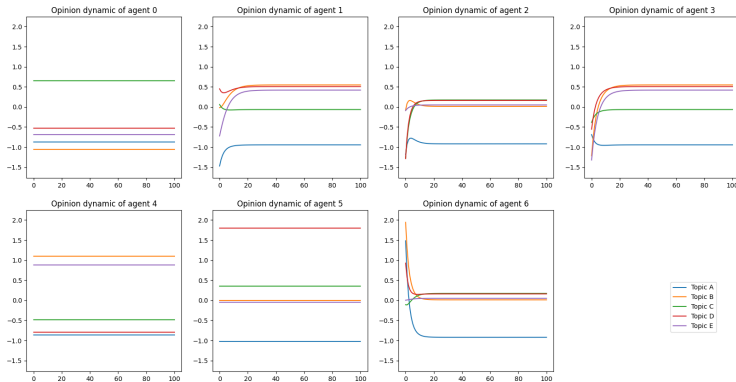
being  $\Pi_U$  a proper function setting to zero the dynamic of stubborn agents. In this case the dynamic is proved to converge to the *harmonic extension* of  $x_0|_U$  being the closest to  $x_0$ : the harmonic extension  $\tilde{x}_0|_U$  of  $x_0|_U$  satisfies:

$$\begin{cases} [\tilde{x}_0|_U]_u = [x_0|_U]_u & \forall u \in U \\ [L_{\mathcal{F}}\tilde{x}_0|_U]_v = 0 & \forall v \notin U \end{cases}$$



# Modeling stubborn agents

## 3 Advanced opinion dynamics modeling



**Figure:** Opinion trajectories for a constant  $\mathbb{R}^n$  sheaf defined over a network of 7 agents and a  $n = 5$  topics basis and Gaussian distributed initial opinions with  $|U| = 3$  stubborn agents





# Weighted reluctance

## 3 Advanced opinion dynamics modeling

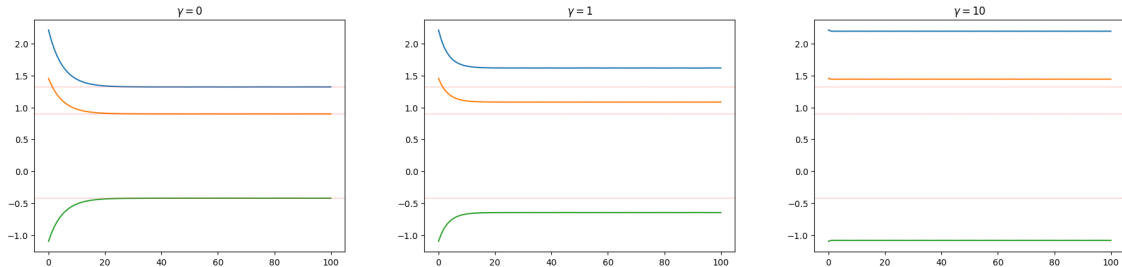
This model requires a specific augmentation construction:

1. Given  $G(V, E)$ , define a copy of  $V$  and attach each  $v \in V$  with its copy  $v' \in V'$  via  $e'$ ;
2. Extend the sheaf  $\mathcal{F}$  to  $\mathcal{F}'$  by letting for each  $e \sim (v, v')$ :
  - $\mathcal{F}'(v) = \mathcal{F}'(v') = \mathcal{F}'(e'), ;$
  - $\mathcal{F}'_{v \triangleleft e'} = \mathcal{F}'_{v' \triangleleft e'} = \sqrt{\gamma} \mathbb{I}$
3. Given an initial condition  $x(0) \in \mathcal{C}^0(G, \mathcal{F})$ , extend it to  $x'(0) \in \mathcal{C}^0(G', \mathcal{F}')$  by setting  $x_0(v) = x'_0(v')$  for each  $e \sim (v, v')$ ;
4. Apply the stubborn-agents model to  $\mathcal{F}'$  defining the set of stubborn agents as  $V'$ .



# Weighted reluctance

## 3 Advanced opinion dynamics modeling



Opinion trajectory of one agent on a constant  $\mathbb{R}^n$  ( $n = 3$ ) discourse sheaf over a graph with  $|V| = 4$  agents modifying the reluctance weight



# Learning to lie

## 3 Advanced opinion dynamics modeling

Opinion dynamics up to now are all based on private-opinions distributions, i.e. time functionals of the 0-cochains. Using sheaf formalism, we can also model dynamics of the restriction maps using the coboundary matrix:

$$\frac{d\delta^T}{dt} = -\beta \Pi_\delta (\delta^T \mathbf{x} \mathbf{x}^T) \quad (8)$$

where  $\mathbf{x}$  is a fixed distribution of private opinions and  $\Pi_\delta$  is a projector to preserve the block structure of  $\delta$ . This equation can be decomposed within each restriction map as:

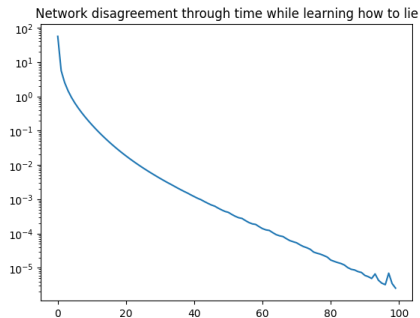
$$\frac{d\mathcal{F}_{v \triangleleft e}}{dt} = -\beta (\mathcal{F}_{v \triangleleft e} \mathbf{x}_v - \mathcal{F}_{u \triangleleft e} \mathbf{x}_u) \mathbf{x}_v^T \quad (9)$$



# Learning to lie

## 3 Advanced opinion dynamics modeling

The main result is that this dynamic converges to the nearest sheaf in terms of  $l_2$  distance between the respective coboundary maps such that the fixed private opinion distributions is a global section.





# Joint dynamic of private opinion and expression

## 3 Advanced opinion dynamics modeling

The culmination of this modeling approach is to combine in a single system the dynamics of the private opinions and the expression:

$$\begin{cases} \frac{dx}{dt} = -\alpha \Pi_U(\delta \delta^T \mathbf{x}) + Bu \\ \frac{d\delta}{dt} = -\beta \Pi_G(\delta^T \mathbf{x} \mathbf{x}^T) \\ \frac{dy}{dt} = Cx \\ x(0) = x_0 \\ \delta(0) = \delta_0 \end{cases}$$



# Table of Contents

## 4 Beyond graph modeling: group opinion dynamics over simplicial complexes

- ▶ A short introduction to Graph Cellular Sheaves
- ▶ Opinion dynamics over discourse sheaves
- ▶ Advanced opinion dynamics modeling
- ▶ Beyond graph modeling: group opinion dynamics over simplicial complexes

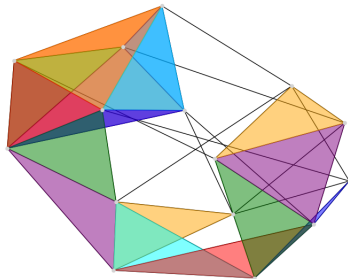


## Why simplicial complexes?

### 4 Beyond graph modeling: group opinion dynamics over simplicial complexes

Complex systems are well modeled by graphs and learning methods on them has gained fortune in exploring *pairwise relations*.

Unfortunately, this well established frameworks are not capable to catch higher order informations residing in more complex relations: further topological structures provides new perspective on this problem, and **simplicial complexes** represent the first step in this direction.





## Definitions and main properties

### 4 Beyond graph modeling: group opinion dynamics over simplicial complexes

Given a finite set of vertices  $\mathcal{V} = \{v_0, \dots, v_{N-1}\}$ , a  $k$ -simplex  $\sigma_i^k$  is an unordered set of  $k + 1$  points. An abstract simplicial complex  $\mathcal{X}$  is a finite set of simplices that is **closed under the inclusion of faces**.

The fundamental operator when defining such structures is the *boundary operator*, which is well represented by the set of adjacency matrices  $B_k$ , encoding for **inclusions of subsets** and **coherence of orientation**:

$$B_k(i, j) = \begin{cases} 0 & \text{if } \sigma_i^{k-1} \not\subset \sigma_j^k; \\ 1 & \text{if } \sigma_i^{k-1} \subset \sigma_j^k \text{ and } \sigma_i^{k-1} \sim \sigma_j^k; \\ -1 & \text{if } \sigma_i^{k-1} \subset \sigma_j^k \text{ and } \sigma_i^{k-1} \not\sim \sigma_j^k \end{cases} \quad (10)$$

The fundamental property for simplicial complexes is the relation between boundaries of subsequent order:

$$B_k B_{k+1} = 0 \quad (11)$$





# Simplicial cellular sheaves

4 Beyond graph modeling: group opinion dynamics over simplicial complexes

## Definition (2-Simplicial cellular sheaf)

Considering a simplicial complex  $\mathcal{S}(V, E, T)$ , we define a **simplicial cellular sheaf**  $\mathcal{F}$  specified by:

- A vector space  $\mathcal{F}(v)$  for each node  $v \in V$ ;
- A vector space  $\mathcal{F}(e)$  for each edge  $e \in E$ ;
- A vector space  $\mathcal{F}(t)$  for each triangle  $t \in T$ ;
- A linear map  $\mathcal{F}_{v \triangleleft e} : \mathcal{F}(v) \rightarrow \mathcal{F}(e)$  for each boundary relation  $v \triangleleft e$  for each edge;
- A linear map  $\mathcal{F}_{e \triangleleft t} : \mathcal{F}(e) \rightarrow \mathcal{F}(t)$  for each boundary relation  $e \triangleleft t$  for each triangle;



# Simplicial Laplacians and Simplicial Sheaf Laplacians

4 Beyond graph modeling: group opinion dynamics over simplicial complexes

The structure of a K-simplicial complex is well described by the set of its *combinatorial Laplacians*, encoding for the lower and the upper adjacencies:

$$L_k = B_k^T B_k + B_{k+1} B_{k+1}^T \quad (12)$$

Similarly, for a simplicial cellular sheaf we generalize this idea using the k-th order coboundary map  $\delta_k$ :

$$L_{\mathcal{F}_k} = \delta_k^T \delta_k + \delta_{k+1} \delta_{k+1}^T \quad (13)$$



## $\mathbb{R}^n$ constant simplicial discourse sheaf

### 4 Beyond graph modeling: group opinion dynamics over simplicial complexes

We shift our initial model to our brand new higher order topology. The related simplicial discourse sheaf is such that:

- Each agent resides on one edge and has a *private opinion* as a vector in  $\mathbb{R}^n$ ;
- Each triangle represent a triple-wise communication, leading to a *tri-expressed opinion* over the same  $n$  topics;

This implies that now:

- The stack of all *private opinions* is a 1-cochain  $\xi \in \mathcal{C}^1(\mathcal{S}, \mathcal{F})$ ;
- The stack of all *tri-shared opinions* is a 2-cochain  $\tau \in \mathcal{C}^2(\mathcal{S}, \mathcal{F})$ ;



# 1-cochains opinion dynamics

## 4 Beyond graph modeling: group opinion dynamics over simplicial complexes

We used the quadratic form  $\langle x, L_{\mathcal{F}}x \rangle$  when working with 0-cochains flow as a **measure of agreement** among agents: let's derive a similar expression for the 1-cochains opinion flow.

$$\begin{aligned} \xi^T L_{\mathcal{F}_1} \xi &= \xi^T (\delta_1^T \delta_1 + \delta_2 \delta_2^T) \xi = \xi^T \delta_1^T \delta_1 \xi + \xi^T \delta_2 \delta_2^T \xi = \\ &= \sum_{v \in V} \|(\delta_1)_v \xi\|_2^2 + \sum_{t \in T} \|\mathcal{F}_{e_1 \triangleleft t} \xi_{e_1} + \mathcal{F}_{e_2 \triangleleft t} \xi_{e_2} - \mathcal{F}_{e_3 \triangleleft t} \xi_{e_3}\|_2^2 \end{aligned} \quad (14)$$

This makes us question the tractability of lower order dynamics from higher order ones. If we just consider the entanglement of 1-cochains and 0-cochains to be  $\xi = \delta_0^T x$  (*agents on edges as **mediators** between agents on nodes*), this would yield:

$$\langle \xi, L_{\mathcal{F}_1} \xi \rangle = \langle \delta_1^T x, (\delta_1^T \delta_1 + \delta_2 \delta_2^T) x \rangle = x^T \delta_1 \delta_1^T \delta_1 \delta_1^T x + \cancel{x^T \delta_1 \delta_2 \delta_2^T \delta_1^T x} = \|L_{\mathcal{F}_0} x\|_2^2 \quad (15)$$



## 1-cochains opinion dynamics

### 4 Beyond graph modeling: group opinion dynamics over simplicial complexes

We'll consider the following dynamic over the 1-cochains:

$$\frac{d\xi}{dt} = -\gamma L_{\mathcal{F}_1} \xi = -\gamma(\delta_1^T \delta_1 + \delta_2 \delta_2^T) \xi \quad (16)$$

Given an initial state  $\xi(0)$ , the trajectory has the form:

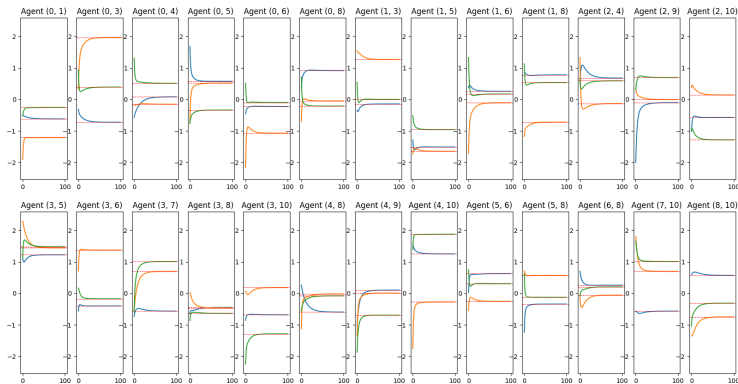
$$\xi(t) = e^{-\gamma(\delta_1^T \delta_1 + \delta_2 \delta_2^T)t} \xi(0) \quad (17)$$

which converges to the projection of the initial state over  $\ker(\delta_1^T \delta_1 + \delta_2 \delta_2^T)$ .



# Numerical experiments

## 4 Beyond graph modeling: group opinion dynamics over simplicial complexes



**Figure:** Opinion trajectories for a constant  $\mathbb{R}^n$  sheaf defined for a network of 26 agents placed over the edges of an ER graph of 11 nodes, expanded to a 2-SC with 3 triangles (2, 4, 10), (0, 5, 8), (1, 3, 6)



## Numerical experiments

### 4 Beyond graph modeling: group opinion dynamics over simplicial complexes

As we can see, the disagreement between agents is a decreasing function of time depending on the superposition of the effect of both the lower and the upper adjacencies. Further tools might be used to investigate the nature of  $\ker(L_{\mathcal{F}_1})$  in relation with  $\ker(\delta_1^T \delta_1)$  and  $\ker(\delta_2 \delta_2^T)$ .

