

# Experiments of Boundary Stimulations of Social Influence Networks

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**Scripts and simulations at:**

<https://github.com/mboudour/SocInfluenceSims>

- **The IBV Problem of Diffusion**

$$\begin{aligned}\frac{\partial u}{\partial t} &= \alpha \Delta u, x \in \Omega, t > 0, \\ u(x, t) &= f(x, t), x \in \partial\Omega, t > 0, \\ u(x, 0) &= g(x), x \in \Omega \text{ (where } f(x, 0) = g(x), x \in \partial\Omega\text{)}.\end{aligned}$$

- **Discretization in Time and Space**

$$\begin{aligned}u(x, t + \delta t) &= u(x, t) + u_t(x, t)\delta t + O(\delta t^2), \\ u(x + \delta x, t) &= u(x, t) + u_x(x, t)\delta x + \frac{1}{2}u_{xx}(x, t)\delta x^2 + O(\delta x^3), \\ u(x - \delta x, t) &= u(x, t) - u_x(x, t)\delta x + \frac{1}{2}u_{xx}(x, t)\delta x^2 + O(\delta x^3), \\ u_t(x, t) &= \frac{u(x, t + \delta t) - u(x, t)}{\delta t} + O(\delta t), \\ u_{xx}(x, t) &= \frac{u(x - \delta x, t) - 2u(x, t) + u(x + \delta x, t)}{\delta x^2} + O(\delta x^2).\end{aligned}$$

- The Case of One-Dimensional Lattice

$$\begin{aligned}\Omega &= \{x_j : j = 1, \dots, n\}, \\ \partial\Omega &= \{x_1, x_n\}, \\ t &\in \{t_m : m = 0, 1, \dots\}.\end{aligned}$$

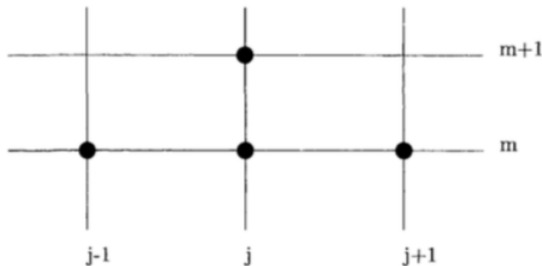


FIGURE 4.1. *The computational molecule of the explicit scheme.*

- **Notation**

$$\begin{aligned}x_j &= j, \quad j = 1, \dots, n, \\t_m &= m, \quad m = 0, 1, \dots, \\u(x_j, t_m) &= u_j^m, \\f(x_j, t_m) &= f_j^m, \\g(x_j) &= g_j.\end{aligned}$$

- **The Parameter  $r$**

$$r = \alpha \frac{\delta t}{\delta x^2}.$$

# The One-Dimensional Lattice $\Omega$ as a Graph

- **The Discrete Equation of Diffusion**

$$\begin{aligned}u_j^{k+1} - u_j^k &= r(u_{j-1}^k + u_{j+1}^k) - 2ru_j^k, j \in \Omega, k = 0, 1, \dots, \\u_j^k &= f_j, j \in \partial\Omega, k = 0, 1, \dots, \\u_j^0 &= g_j, j \in \Omega.\end{aligned}$$

- **Stability Condition**

$$r < \frac{1}{2}$$

- Adjacency between nodes of  $\Omega$

$$i \sim j \Leftrightarrow |i - j| = 1.$$

- **Adjacency Matrix**

$$A = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 & 0 & 0 \\ 1 & 0 & 1 & \dots & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & 1 & 0 & 1 \\ 0 & 0 & 0 & \dots & 0 & 1 & 0 \end{bmatrix}.$$

- **Degrees of nodes of  $\Omega$**

$$\deg_i = \sum_{j=1}^n A_{ij} = \begin{cases} 2, & \text{for } i = 2, \dots, n-1, \\ 1, & \text{for } i = 1, n. \end{cases}$$

- **Notation**

- Let  $D$  denote the diagonal **degree matrix** such that  $D_{ii} = \deg_i$ .
- Let  $I$  denote the diagonal **unit matrix** such that  $I_{ii} = 1$ .

- Thus, the discrete equation of diffusion in vector form is written as:

$$\begin{aligned}u^{k+1} - u^k &= r(A - D)u^k = \\&= -rLu^k, \text{ in } \Omega, k = 0, 1, \dots, \\u^k &= f, \text{ on } \partial\Omega, k = 0, 1, \dots, \\u^0 &= g, \text{ in } \Omega.\end{aligned}$$

where the  $n \times n$  matrix  $L = D - A$  denotes the  
(**combinatorial**) **Laplacian matrix**  $L = L_{ij}$  of graph  $\Omega$ :

$$L_{ij} = \begin{cases} \deg_i, & \text{whenever } i = j, \\ -1, & \text{whenever } i \sim j, \\ 0, & \text{otherwise.} \end{cases}$$

# The Discrete Equation of Diffusion on a Graph $G$

- Let  $G = (V, E)$  be a (general) graph, such that the set of vertices  $V$  is partitioned in two subsets:

$$V = \Omega \cup \partial\Omega,$$

assuming that

$$\partial\Omega = \{x \notin \Omega : \exists y \in \Omega \text{ such that } y \sim x\}.$$

- The **discrete equation of diffusion on  $G$  with Dirichlet boundary conditions** is (in vector form):

$$\begin{aligned} u^{k+1} - u^k &= -rLu^k, \text{ in } \Omega, k = 0, 1, \dots, \\ u^k &= 0, \text{ on } \partial\Omega, k = 0, 1, \dots, \\ u^0 &= g, \text{ in } \Omega. \end{aligned}$$

where  $L$  is the Laplacian matrix of  $G$ .



# The DeGroot–Friedkin–Johnsen Model of Social Influence

- Let  $G = (V, E)$  be a **graph** of  $n$  persons, i.e., we assume from now on that **vertices**  $\equiv$  **persons**.
- For each person  $i \in V$  and time  $k = 0, 1, 2, \dots$  (in the discrete case considered here), let  $v_i^k \in \mathbb{R}$  denote  $i$ 's **opinion** at time  $k$ .
- Person's  $i$  opinion at time  $t$  is updated at next instance  $t + 1$  according to the following equation of the  
**DeGroot–Friedkin–Johnsen Model of Social Influence:**

$$v_i^{k+1} = s_i N v_i^k + (1 - s_i) v_i^0,$$

- where  $N v_i^t$  is the average opinion of  $i$ 's neighbors
- and  $s_i$  is person's  $i$  **susceptibility coefficient**, a scalar parameter in the interval  $(0, 1]$ .

① Remarks on the definition of the **susceptibility coefficient**:

- If  $s_i = 0$ , then  $i$ 's opinion does not change ( $v_i^k = v_i^0$ , for each time  $k = 1, 2, \dots$ ). Such a person is called **stubborn** or **persistent** in her opinion.
- If  $s_i = 1$ , then  $i$  adopts the average opinion of her neighbors  $N v_i^t$ . Such a person is called **malleable** or **fully compliant** in adopting her neighbor's influence.
- If  $0 < s_i < 1$ , then  $i$ 's opinion is inserted in-between  $N v_i^t$  and  $v_i^0$ , where the exact inserted position is weighted by  $s_i$ .

② Remarks on the definition of matrix  $N$ , which is called **walk matrix** on graph  $G$ :

- Denoting by  $A, D$  the adjacency and the degree matrix of  $G$ , we have

$$N = D^{-1}A.$$

- Moreover, denoting by  $L, I$  the Laplacian and the unit matrix of  $G$ , we have

$$N = I - D^{-1}L.$$

# Reduction of Social Influence to a Diffusion Process

- Denoting by  $S$  the  $n \times n$  diagonal matrix with its diagonal entries equal to the  $s_i$ 's, if

$$S = I$$

(i.e., if all persons are fully malleable), then the DeGroot–Friedkin–Johnsen model of social influence becomes (in vector form):

$$v^{k+1} = N v^k,$$

i.e., since  $N = I - D^{-1}L$ ,

$$v^{k+1} - v^k = -D^{-1} L v^k,$$

which is a diffusion equation with a **variable diffusion coefficient** equal to  $D^{-1}$ .

# Reduction of Diffusion Process to Social Influence

- If  $G$  is a  $d$ -regular graph, for a sufficiently large positive integer  $d$ , and

$$r = \frac{1}{d},$$

then the equation of a diffusion process (in vector form) becomes:

$$u^{k+1} - u^k = -D^{-1}Lu^k = Nu^k - u^k,$$

i.e.,

$$u^{k+1} = Nu^k,$$

which is a process of social influence (in the DeGroot–Friedkin–Johnsen model) for  $S = I$  (i.e., when all persons are fully malleable).

# The Boundary of a Social Influence Process

- Suppose that we have a process of social influence on a graph  $G = (V, E)$  of  $n$  persons indexed by  $i = 1, 2, \dots, n$  such that each person has a coefficient of susceptibility  $s_i \in [0, 1]$ .
- Assumptions and Notation:
  - $G$  is connected.
  - $V = \Omega \cup \partial\Omega$ , where  $\Omega \cap \partial\Omega = \emptyset$ .
  - $\Omega = \{i \in V : s_i > 0\}$ .
  - $\partial\Omega = \{i \in V : s_i = 0\}$ .
  - $|\partial\Omega| < n$ .
  - $\forall j \in \partial\Omega, \exists i \in \Omega, i \sim j$ .

# Sources of Boundary Stimulations to Social Influence

- Given a process of social influence on a graph  $G = (\Omega \cup \partial\Omega, E)$  with regards to a vector of susceptibility coefficients  $s = \{s_i\}_{i \in \Omega}$ , person  $j_b \in \partial\Omega$  is said to be a **source of boundary stimulation**, or a **persistent source**, if  $j_s$  fires up the process governed by the following IBVP:

$$\begin{aligned}v_i^{k+1} &= s_i N v_i^k + (1 - s_i)\psi_i, \text{ for } i \in \Omega, k = 1, 2, \dots, \\v_{j_b}^k &= 1, \text{ for } k = 0, 1, \dots, \\v_i^1 &= \psi_i, \text{ for } i \in \Omega.\end{aligned}$$

where

$$\psi_i = \begin{cases} \frac{s_i}{\deg_i}, & \text{whenever } i \sim j_b, \\ 0, & \text{whenever } i \not\sim j_b. \end{cases}$$

# Transient and Steady Solutions to Boundary Stimulations

- Let  $v_i^k(j_b) = v_i^k(j_b; s)$  be the **transient** solution to the IBVP of social influence stimulated by source  $j_b$  with regards to the vector of susceptibility coefficients  $s = \{s_i\}_{i \in \Omega}$ .
- Then, for any  $i \in \Omega$ , the following limit exists:

$$\lim_{k \rightarrow \infty} v_i^k(j_b; s) = \bar{v}_i(j_b; s).$$

- $\bar{v}(j_b) = \bar{v}(j_b; s) = \{\bar{v}_i(j_b; s)\}_{i \in \Omega}$  is the **steady state** solution to the BV of social influence on  $G$ , which is stimulated by source  $j_s$  with regards to  $s$ :

$$\bar{v}(j_b) = (I - SN)^{-1} (I - S) \psi,$$

where  $\psi = \{\psi_i\}_{i \in \Omega}$ .

- If, for all  $i \in \Omega$ ,  $s_i = 1$ , then

$$\lim_{k \rightarrow \infty} v_i^k(j_b; \mathbf{1}) = 1, \text{ for any } i \in \Omega.$$

# Synchronization from One Source of Alternating Boundary Simulation to Social Influence

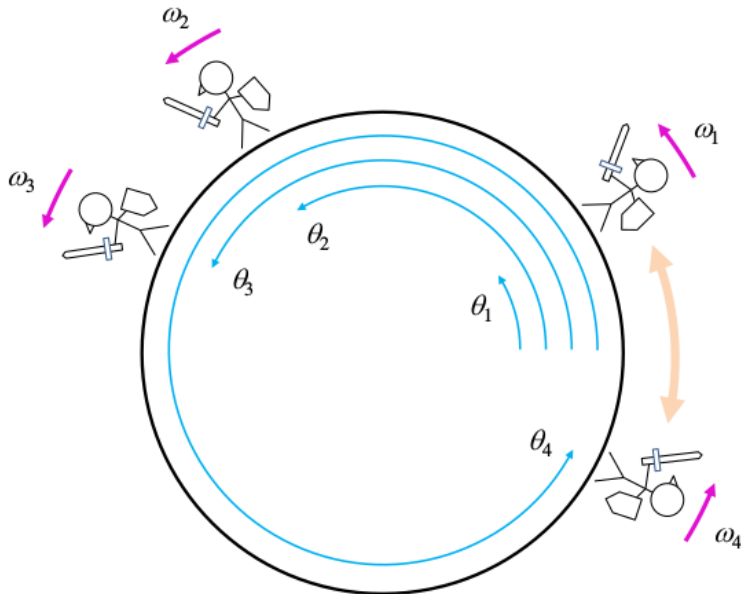
- A person  $j_a \in \partial\Omega$  is said to be an **alternating source of boundary stimulation**, or an **alternating source**, if  $j_a$  fires up a process of social influence on graph  $G$  that satisfies the following IBVP:

$$\begin{aligned}v_i^{k+1} &= s_i N v_i^k + (1 - s_i) v_i^0, \text{ in } \Omega, k = 0, 1, \dots, \\v_{j_a}^k &= 1, k = 0, 2, 4, \dots, \\v_{j_a}^k &= -1, k = 1, 3, 5, \dots, \\v_i^0 &= 0, \text{ in } \Omega.\end{aligned}$$

- Note that the opinion of the source  $j_a$  constantly oscillates between  $+1$  and  $-1$  without being influenced by her neighbors (although, as a source,  $j_a$  does influence her neighbors).



# Synchronization in the Kuramoto Model



- **The Kuramoto Model of Synchronization among Coupled Oscillators on a Graph**

**Oscillators** are located on the nodes of **graph**  $G = (V, E)$ . Each oscillator  $i \in V$  marches on a circular track with a preferred speed  $\omega_i$ . The position of oscillator  $i$  on the circle is given by an angle  $\theta_i$ . However, since oscillators are considered to be **coupled** with regards to their graph proximities,  $i$ 's angular position  $\theta_i$  depends on her neighbors' angular positions  $\theta_j$  according to the following equation:

$$\begin{aligned}\theta_i^{k+1} - \theta_i^k &= \omega_i + \alpha \frac{\sum_{j \sim i} \sin(\theta_j^k - \theta_i^k)}{\deg_i}, \text{ in } \Omega, k = 0, 1, \dots, \\ \theta_i^0 &= \phi_i, \text{ in } \Omega.\end{aligned}$$

# The Influentiability Matrix of a Graph Influence Process of Boundary Stimulations

- By a **graph influence process**, we mean a process of social influence on a graph  $G = (V, E)$  with regards to a vector of susceptibility coefficients  $s = \{s_i\}_{i \in V}$ .
- The **influentiability matrix** of a graph influence process stimulated on the boundary of graph  $G$  is defined as the following  $n \times n$  matrix  $\mathcal{U}^\infty = \{\mathcal{U}_{ij}^\infty(s)\}_{i,j \in V}$ :

$$\mathcal{U}_{ij}^\infty(s) = \begin{cases} 1, & \text{for } i = j, \\ \bar{v}_j(i; s), & \text{for } i \neq j. \end{cases}$$

# The Influence Degree Centrality Index of a Node in a Graph Influence Process

- Given a graph influence process on a graph  $G = (V, E)$  with regards to a vector of susceptibility coefficients  $s = \{s_i\}_{i \in V}$ , the **influence degree centrality index** of node  $i \in V$  is defined as follows:

$$\begin{aligned} c_{\text{influence-degree}}(i; s) &= \frac{1}{n-1} \sum_{j \sim i} \mathcal{U}_{ij}^{\infty}(s) \\ &= \frac{1}{n-1} \sum_{j \sim i} \bar{v}_j(i; s). \end{aligned}$$

- Denoting by  $\mathbf{1}$  the  $\mathbb{R}^{|V|}$  vector having all its components equal to 1, we get

$$c_{\text{influence-degree}}(i; \mathbf{1}) = \frac{1}{n-1} \deg_i = c_{\text{degree}}(i),$$

where  $c_{\text{degree}}(i)$  is the **degree centrality index** of node  $i \in V$ .

# Correlations between Influence Degree Centrality Index and Degree Centrality Index when $s_i = c_{degree}(i)$

For the Karate network:

