

# On the design of public debate in social networks

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  - social influence on individuals decay over time (→ intensity of social debate, or speed at which opinion crystallize) (similar to DeMarzo, Vayanos and Zwiebel (2003))

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  - **trade-off between fostering the convergence of opinions and the risk of polarization** from the view-point of the social planner

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- $N_{\mathcal{G}}(i)$ : neighborhood of  $i$  in the network  $\mathcal{G}$

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- 2  $\sigma = \tau$ ,  $\lambda = 1$ : **Hegselmann-Krause model**

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  - These intervals are ordered and never nested: For all  $i, j$

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- $(\sigma, \tau)$ -*compatible implies compatible.*

## Lemma

Let  $(x(t), \mathcal{G}(t))$  be a compatible network-opinion pair. Then for any network formation process, one has:

- 1 If  $x_i(t) = x_j(t)$  then  $x_i(t') = x_j(t')$  for all  $t' \geq t$ .
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In the following, we assume that agents are initially numbered so that

$$x_1(0) < x_2(0) < \dots < x_n(0)$$

Due to the above Lemma, they will always remain ordered in this way.

## Link fragility

- A link  $\{i, j\}$  is *maximal* if  $i' \leq i, j' \geq j$  implies that  $\{i', j'\}$  is not a link or is equal to  $\{i, j\}$ .

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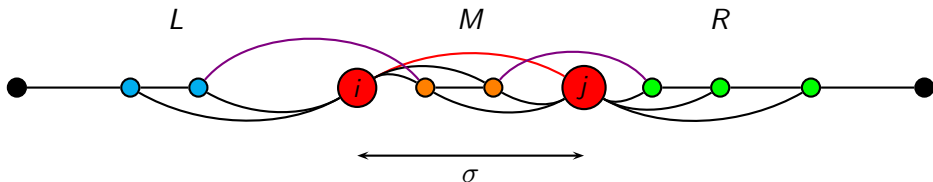
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- An answer is possible, knowing the *local structure of the link*  $\{i, j\}$ :  
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## Definition

Using the above notation, the fragility of a (maximal) link  $\{i, j\}$  in  $\mathcal{G}$  is defined by:

$$\phi_{i,j}(\mathcal{G}) = \max_{m \in \{0, \dots, M\}} \frac{(L + M + 2)(2R - r_m) + (R + M + 2)(L - \ell_{m+1}) + (L - R)(M - m + 1)}{(L + M + 2)(R + M + 2)} \quad (1)$$

letting  $r_0 = 0$  and  $\ell_{M+1} = 0$ .



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## Proposition

*Let  $\sigma, \tau, \lambda$  be given and  $\{i, j\}$  be a maximal link in the network  $\mathcal{G}(t)$ . Then one has  $\{i, j\} \in \mathcal{G}(t + 1)$  for any  $x(t)$  that is  $(\sigma, \tau)$ -compatible with  $\mathcal{G}(t)$  if and only if  $\phi_{i,j}(\mathcal{G}(t)) \leq 1$ .*

We assume  $\tau = 0$  (no link creation) in all this part.

- A maximal link  $\{i, j\}$  is *fragile* if  $\phi_{i,j} > 1$ .

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*If the initial network  $\mathcal{G}(0)$  has no fragile link, then for all  $\lambda \in [0, 1]$ , one has  $\mathcal{G}(t) = \mathcal{G}(0)$  for all  $t \in \mathbb{N}$ , i.e., the network is stable for the dynamics independently of the choice of a compatible  $x(0)$ .*

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Answer: Supposing that the difference of opinions on a link is bounded above by  $\rho < \sigma$  and that  $\frac{\sigma}{\rho} \geq \bar{\phi}_{\mathcal{G}(0)}$ , the network is stable provided  $\lambda$  is smaller than a constant depending on  $\bar{\phi}_{\mathcal{G}(0)}$  and  $\sigma/\rho$ .

# Polarization

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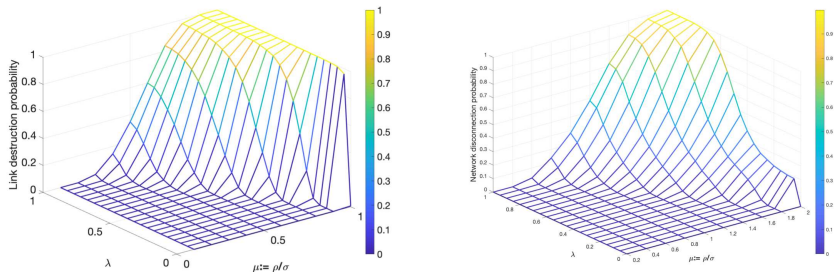
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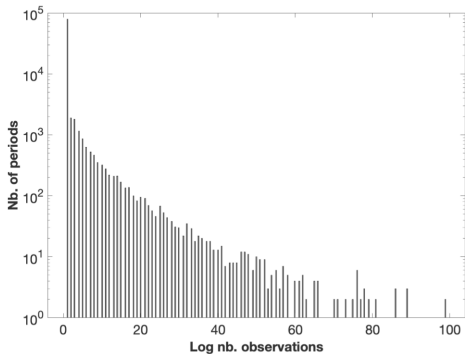
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- Similarly to network stability, in the case that  $\overline{\psi}_{\mathcal{G}(0)} > 1$ , and supposing that  $x_{i+1}(0) - x_i(0) \leq \rho < \sigma$  for  $i = 1, \dots, N - 1$ , and  $\frac{\sigma}{\rho} \geq \overline{\psi}$ , the network  $\mathcal{G}(t)$  remains connected for all  $t \in \mathbb{N}$  if  $\lambda$  is smaller than a constant depending on  $\overline{\psi}_{\mathcal{G}(0)}$  and  $\sigma/\rho$ .

# Simulations

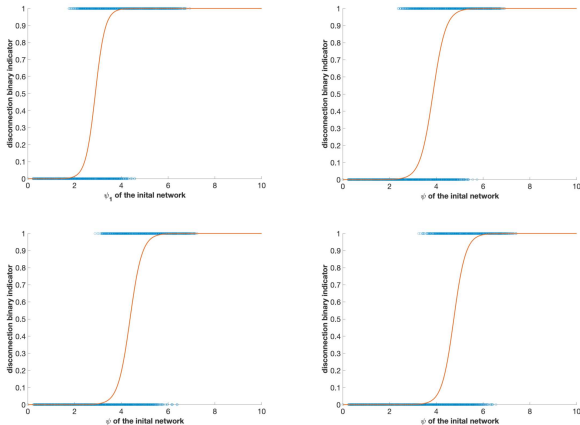


**Figure:** Empirical probability (computed over 50,000 graphs with various initial opinions and edge-connectedness) of the destruction of at least one link (left panel) and of the disconnection/polarization of the network (right panel) for various levels of  $\lambda$  and  $\mu = \rho/\sigma$ .



**Figure:** Histogram of the time to disconnection/polarization for networks in the sample that actually disconnects.

# Simulations



**Figure:** Scatter plot of a binary indicator of disconnection (equal to 1 if the network disconnects/polarizes in the course of the simulation and to 0 otherwise) as a function of the polarizability of the initial network  $\bar{\psi}$ . Upper left to lower right: edge-connectdness = 1, 2, 3, 4 resp. Logistic regression curve is displayed in red.

## Proposition

Assume that the network  $\mathcal{G}$  is stable and connected, with initial opinion vector  $x(0)$ , and denote by  $w$  the normalized left eigenvector of  $G$  associated to eigenvalue 1. Then the opinion vector  $x(t)$  can be asymptotically approximated by

$$\begin{aligned}x(\infty) &\approx \left[ \frac{\varphi(\lambda)}{\log(\lambda)} (\log(1-\lambda) + \Psi_\lambda(1)) G^2 + \varphi(\lambda) G \right. \\ &\quad \left. + \left( 1 - \frac{\varphi(\lambda)}{\log(\lambda)} (\log(1-\lambda) + \log(\lambda) + \Psi_\lambda(1)) \right) \mathbf{1} w^T \right] x(0) \\ &\approx \left[ (1 - \varphi(\lambda)) \mathbf{1} w^T + \varphi(\lambda) G \right] x(0).\end{aligned}$$

red: consensus part    blue: strong diversity part

$$\lim_{\lambda \rightarrow 1} \varphi(\lambda) = 0, \quad \lim_{\lambda \rightarrow 0} \varphi(\lambda) = 1$$

# Diameter of opinions

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- It can be shown that the diameter is decreasing with  $\lambda$ :

$$\lambda > \mu \Rightarrow \delta(x_\lambda(t)) \leq \delta(x_\mu(t))$$

(some restrictions are needed for  $<$ , e.g.,  $G$  connected and  $x(0)$  s.t.  $x_1(0) < x_2(0) \cdots < x_n(0)$ )

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- To ensure stability of the (links in the) society,  $\lambda$  must be smaller than a constant **increasing with**  $\bar{\phi}_{\mathcal{G}(0)}$  and  $\sigma/\rho$  (stronger condition).
- The result on convergence shows that, even if there is no polarization, there is no consensus in the society (contrarily to most models)(*strong diversity* vs. *weak diversity*).

# Campaign strategy and polarization

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- Consider two candidates  $\mathcal{L}$  and  $\mathcal{R}$  aiming at maximizing their electorate

$$u_{\mathcal{L}}(x) = \text{card}\{i \in \mathcal{N} \mid x(i) \leq 0\}$$

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- If  $\lambda$  increases, the diameter shrinks (and so  $\mathcal{R}$  is *a fortiori* winning), but at some point, the society disconnects: the voters in favor of  $\mathcal{L}$  are no more under the influence of  $\mathcal{R}$ : **the minority candidate has incentives to exacerbate the public debate so as to separate his/her electorate from the majority.**

# Campaign strategy and polarization

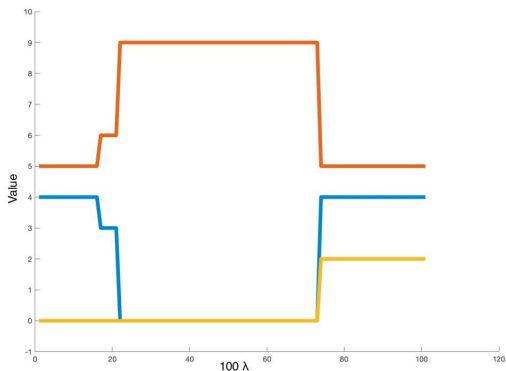


Figure: Values of  $u_L(x_\lambda(\infty))$  (blue), of  $u_R(x_\lambda(\infty))$  (red) and of the number of connected components (yellow) in  $G_\lambda$  for  $\lambda$  varying between 0 and 1.

Thank you for your attention!

# Convergence and asymptotic opinions

- We suppose that the network has reached or is initialized in a state  $(x(0), \mathcal{G}(0))$  such that no link is fragile, and link creation is discarded.

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- We suppose that the network has reached or is initialized in a state  $(x(0), \mathcal{G}(0))$  such that no link is fragile, and link creation is discarded.
- We suppose  $\lambda \in ]0, 1[$  and want to know the evolution of  $x(t)$ .
- We have:

$$\begin{aligned}x(t+1) &= [\lambda^t G + (1 - \lambda^t)I]x(t) \\ &= M(t)x(t) = M(t)M(t-1) \cdots M(1)x(1) = H(t)x(1)\end{aligned}$$

with  $x(1) = Gx(0)$ , and

$$H(t) = \prod_{k=1}^t (\lambda^k G + (1 - \lambda^k)I).$$

Hence,  $H(t)$  is a polynomial in  $G$  of degree  $t$ , i.e., of the form:

$$H(t) = h_{t,t}G^t + h_{t-1,t}G^{t-1} + \cdots + h_{1,t}G + h_{0,t}I.$$



## Proposition

The coefficients of the polynomial  $H(t)$  in  $G$  are given by

$$h_{t,t} = \lambda^{t(t+1)/2}$$

$$h_{0,t} = \prod_{i=1}^t (1 - \lambda^i)$$

$$h_{i,t} = \lambda^{t(t+1)/2} \sum_{1 \leq j_1 < j_2 < \dots < j_i \leq t} \prod_{\substack{\ell=1 \\ \ell \neq j_1, \dots, j_i}}^t \frac{1 - \lambda^\ell}{\lambda^\ell} \quad (0 < i \leq t)$$

with the convention that  $\prod_{\emptyset} = 1$ . The asymptotic behavior of the coefficients is given by  $\lim_{t \rightarrow \infty} h_{i,t} > 0$  if  $i \in \mathbb{N}$ , while  $\lim_{t \rightarrow \infty} h_{i(t),t} = 0$  if  $i(t) \rightarrow \infty$ . In particular, we have

$$\lim_{t \rightarrow \infty} h_{0,t} = \varphi(\lambda) \quad (\varphi : \text{Euler function}; \Psi_\lambda : \lambda\text{-Digamma function})$$

$$\lim_{t \rightarrow \infty} h_{1,t} = \varphi(\lambda) \left( 1 + \frac{\log(1 - \lambda) - \log \lambda}{\log \lambda} + \frac{1}{\log \lambda} \Psi_\lambda(1) \right).$$