

Recovering from Viral Narratives

Delayed Interventions and Strategic Seeding in Polarised Networks

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PROJECT REPOSITORY

<https://github.com/alexandrugaloiu/RPpolarisation>



Motivation

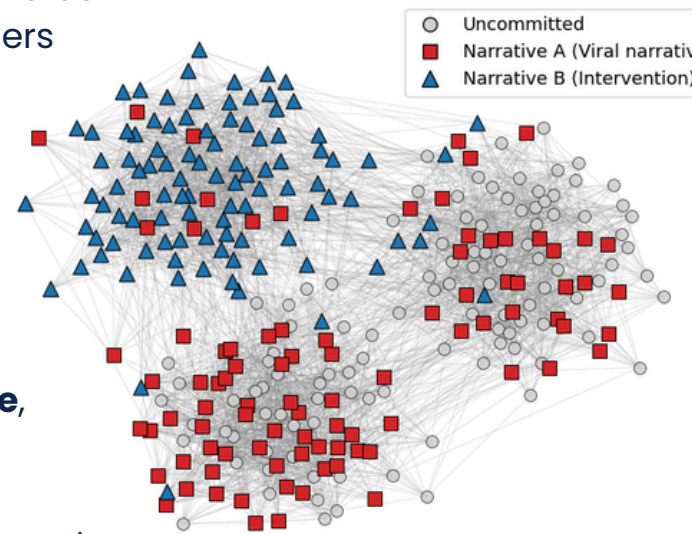


According to a recent medical review, the treatment has shown moderate benefits under specific conditions. Summary of evidence from a peer-reviewed analysis published in the Journal of Clinical Medicine.

- **Online social networks** enable spread of **narratives**.
- **Different** spreading mechanisms [1]
 - **viral**, limited exposure
 - **social reinforcement**, repeated exposure
- **Objective: interventions** against harmful online narratives

Competing Narrative Diffusion Problem

- **One social network** partitioned in disjunct **communities**
- **Two** narratives compete to be adopted by the same users
- Narrative **A = viral information**
- Narrative **B = corrective intervention**
- Factors investigated:
 - **community structure**, segregation
 - **intervention timing**
 - **strategic seed** placement



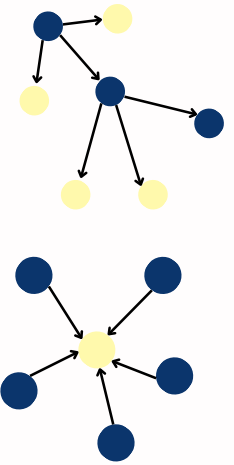
Research question

How do
• **network structure**,
• **intervention timing**
• **intervention strategy**
influence competition between **viral** and **reinforcement-based** diffusion processes?

Approach

Narrative **A** → **Independent Cascade (IC)** [2]
 $P(\text{activation}) = p$

Narrative **B** → **Linear Threshold (LT)** [2]
 $\frac{\text{active neighbors}}{\text{total neighbors}} \geq \theta$



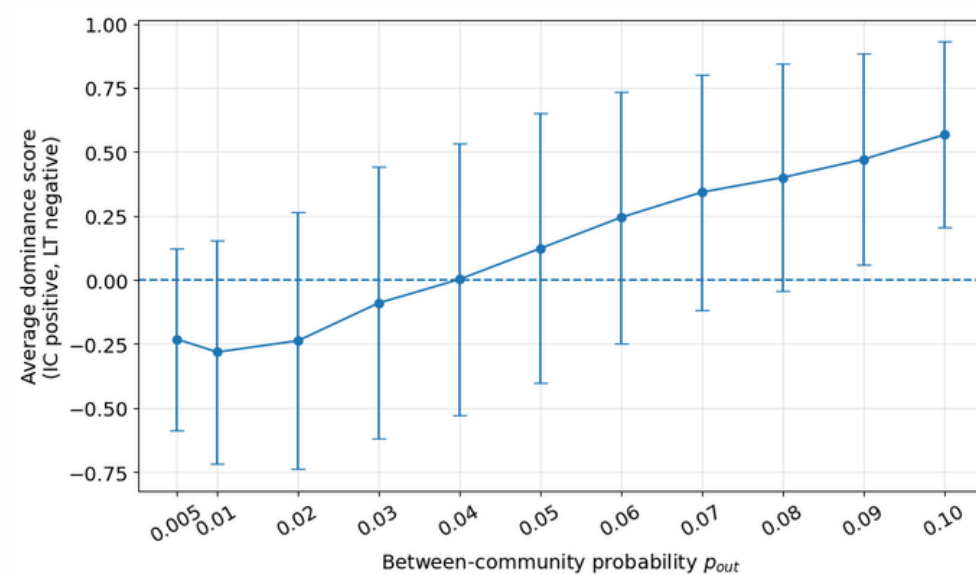
Network generation:

- Stochastic Block Model (SBM) [3]
- Lancichinetti-Fortunato-Radicchi (LFR) [4]

Key metrics:

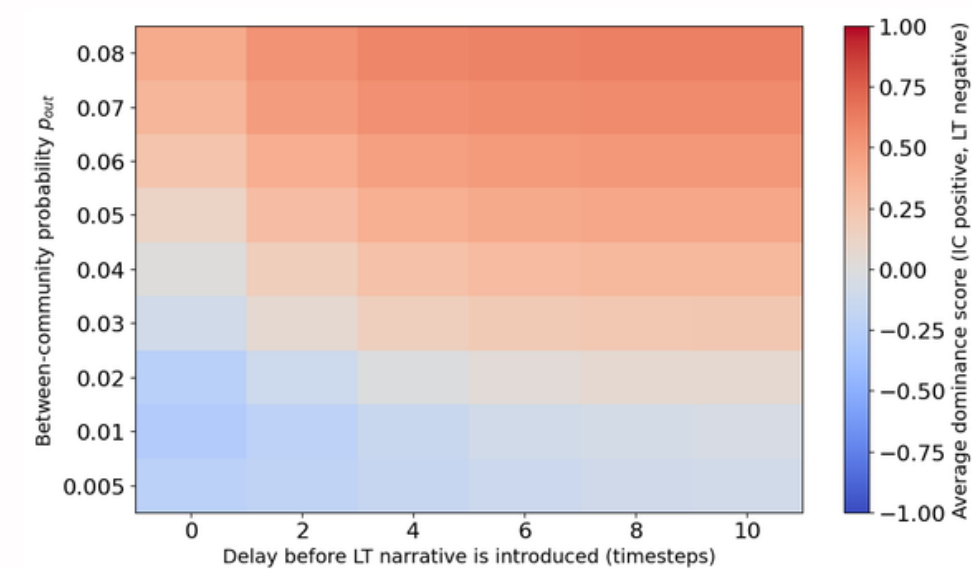
- Adoption counts: $|A|, |B|$
- Dominance score:
 $D = \frac{|A| - |B|}{N}$

Experiment 1: Community structure



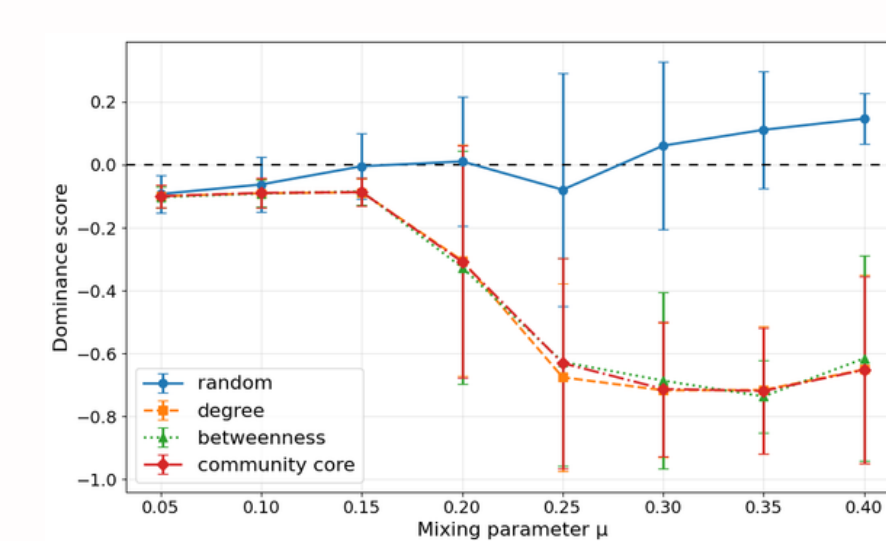
- Key findings:**
- **More segregation** → **corrective intervention (LT)** wins
 - **Less segregation** → **viral narrative (IC)** wins

Experiment 2: Delayed interventions



- Key findings:**
- delays reduce intervention effectiveness
 - interventions **still competitive** in **highly segregated** networks

Experiment 3: Strategic seed placement



- Key findings:**
- strategic seedings **outperform** random seeding
 - strategic seeding **changes the relationship** between segregation and intervention success

Conclusion

- **Network structure** strongly affects **intervention success**
- **Delayed** interventions can still **recover**
- **Strategic seeding** strongly improves outcomes

Effective interventions depend on **when, where and how** the network is structured, **not just the message**.

Future work

- Real social-media networks
- More realistic diffusion models
- Dynamic opinions

End Goal: Better public communication tools

References

- [1] Centola & Macy (2007), Complex Contagions and the Weakness of Long Ties.
- [2] Kempe et al. (2003), Maximizing the Spread of Influence through a Social Network.
- [3] Holland et al. (1983), Stochastic Blockmodels: First Steps.
- [4] Lancichinetti et al. (2008), Benchmark Graphs for Testing Community Detection Algorithms.
- [5] Budak et al. (2011), Limiting the Spread of Misinformation in Social Networks.
- [6] Stein et al. (2023), Network Segregation and the Propagation of Misinformation.