

Introduction to the Game Theory of Information Sharing in Social Networks Report



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This my report presents a simplified analysis of a research paper that applies game theory to information dissemination in social networks. The original research, conducted by Zinoviev, Duong, and Zhang from Suffolk University, explores the mathematical and psychological factors that influence how information spreads through social networks.

In my analysis, I've focused on making this complex academic work accessible by using clear heading structures and simplified explanations. Rather than delving into the dense mathematical formulations and proofs found in the original paper, I've provided a high-level overview that captures the essential concepts while avoiding technical complexity.

The report breaks down how social networks are structured, how information is represented, and how different personality types within networks affect information flow. I've placed particular emphasis on explaining the three key components that determine audience orientation and engagement (knowledge, reputation, and popularity) and how these components interact within the game theory model.

Through this analysis, I show insights into why some online communities rapidly share information?? while others are more cautious!! without overwhelming the reader with the underlying mathematical details. The report concludes with potential directions for future research that could build upon this foundational work.

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Documentation Notice:

This document presents an original analysis of information dissemination in social networks using game theory. It offers a simplified explanation of the complex mathematical and psychological frameworks presented in Zinoviev, Duong, and Zhang's research. This analysis, including all explanations, interpretations, and practical implications, represents the original intellectual contribution of the author while making advanced concepts accessible to a wider audience.

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What is this research about?

This research investigates how information spreads in social networks using game theory. Researchers want to understand why people choose to share information or not, why they give feedback or don't, and how information spreads among people.

How a social network is built

The researchers modeled a social network using an undirected graph called $G = (Z, Y)$:

- Z represents the people in the network
- Y represents the connections between people
- Since the network is connected and undirected, meaning that everyone has access to everyone else

How information is represented

Here the researchers identified something in the network called "claims" - statements that can be true or false:

- There are a total of N possible claims in the network
- ϕ represents how likely a random claim is to be true
- ϕ is higher in serious communities (such as scientific groups)
- ϕ is lower in informal communities (such as chat rooms)

How people understand information

Everyone in the network knows some claims and has opinions about them, so we can display them in these categories:

- F_1 : Total claims a person knows
- F^+ : Claims they believe to be true
- F^- : Claims they believe to be false
- F° : Claims they are uncertain about (Rumors)

So from these categories we can calculate a person's knowledge as follows:

$$K_1 = F^+ + F^- + \lambda F^\circ$$

Where λ is the amount of weight they give to the rumor (between 0 and 1). This is normalized to $k_1 = K_1/N$ to represent what fraction of all possible assertions they know.

Components that determine audience orientation and engagement

Researchers explain that there are three main components that determine how an audience responds to actors sharing information:

1. **Knowledge (K):** This represents how much an actor knows and understands about what they are sharing (i.e. simply does the actor understand and comprehend what they are sharing or not?). When someone has recognized knowledge and is seen as a reliable source, this affects how others receive their information.
2. **Reputation (C):** People with a high reputation in the network have more influence (not fame here but someone of an important character, for example a well-known political figure or a leader in a certain position, so these have a high reputation). The researchers measure this using $c_1 = C_1/N$, which shows how much weight others give to that person's opinions. Reputation acts as a trust factor that affects whether others believe information.
3. **Popularity (P):** Celebrities have influence regardless of their level of knowledge. Their visibility itself affects the network. However, the researchers note that popularity declines over time unless it is maintained through continued engagement, an important factor that the researchers have noted. They measure this as $p_1 = P_1/N$.

These three components aren't motivators but rather factors that explain how information from different sources is evaluated and processed by the network.

Different types of people in networks

Based on how much people value each component (using coefficients κ , σ , and π), the researchers identified typical personality types:

- **Internet trolls:** $\kappa = \sigma = 0.1$, $\pi = 0.8$ (mostly care about popularity)
- **Scientific community:** $\kappa = \sigma = 0.5$, $\pi = 0$ (care about knowledge and reputation)
- **Internet experts:** $\kappa = 0.2$, $\sigma = 0.7$, $\pi = 0.1$ (mostly care about reputation)

These values show how much each type of actor values increasing their knowledge, reputation, and popularity.

The researchers use the formula: $U_1 = \kappa K_1 + \sigma C_1 + \pi P_1$

How information gets passed between people

When someone (sender) shares information with someone else (receiver):

1. The receiver evaluates the information based on:
 - Their own knowledge
 - The sender's reputation
 - The sender's opinion about the information
2. The receiver might:
 - Learn a new assertion
 - Change their opinion about an assertion they already knew
 - Ignore information they already agree with
3. The receiver might give feedback, which affects the sender's reputation
4. Both update their knowledge, reputation and popularity

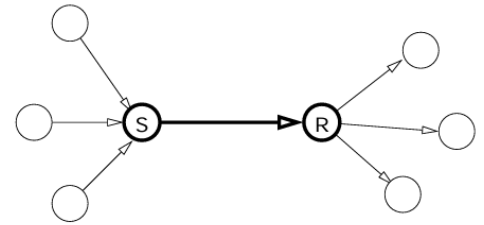


Figure 1: Communication between sender S and receiver R .

The game theory model

The researchers modeled this as a simple game with two players:

- The sender can either share information or not
- The receiver can either give feedback or not
- Both make decisions to maximize their own utility

The researchers proved this game has a "Nash equilibrium" - a stable situation where neither person would benefit from changing their strategy.

What the simulations showed

The researchers ran simulations with 1000 people and found:

- "Troll" networks (popularity-seeking):
 - Information spreads very quickly
 - Eventually everyone knows almost everything
 - The quality of knowledge matches the system average (80% true, 20% false)
- "Expert" networks (reputation-focused):
 - Information spreads much more slowly
 - Knowledge stays more distributed
 - People are more cautious about sharing unverified information
- In all networks:
 - People with low reputation tend to learn faster
 - People with high knowledge or high reputation have less motivation to learn new things

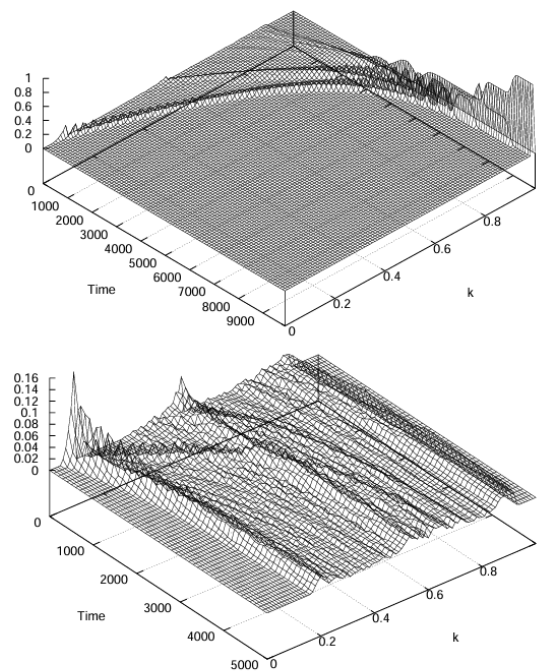


Figure 2: Distribution of actors by k as a function of simulation time. Top: "troll" community; bottom: "expert" community.

Future Research Ideas

Now that the researchers have shown us a good model for how information moves in social networks, there are some things they might want to look at next:

Firstly, the researchers themselves suggested they want to study networks with different kinds of people mixed together, and also look at strategies where people wait a bit before sharing information.

So that, this work opens many research directions:

1. How the structure of networks affects information spread. for example: when people are grouped in tight clusters versus when connections are more random
2. How people's sharing styles might change over time. For example: maybe someone who starts as a "troll" becomes more like an "expert" after certain experiences(this is critical and sensitive point)
3. Adding some kind of fact-checker to the model . As seeing what happens when people can verify information with different levels of effort, this make the traditional operations not useful.
4. Testing what happens when someone deliberately tries to spread false information through different types of networks
5. Checking if their model matches what actually happens on real social media platforms

Final Thoughts on the Research

What makes this research valuable is how it combines psychology with math to explain information sharing in social networks. By looking at how people's values (knowledge, reputation, popularity) affect their decisions to share information or give feedback, the researchers help us understand why information moves differently in different online communities.

Also, that research explains something we see all the time; why some online groups spread information (both true and false) very quickly?? while other groups are much more careful about what they share?? It shows that these differences aren't just random but come from the different types of people in these communities and what they value. This gives us a better understanding of how information spreads online and creates a foundation for future studies that might help design better social networks or find ways to slow down misinformation.