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# Opinion Dynamics over Signed Social Networks 

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## Social Networks



## Opinions

iPhone, Blackberry, or Samsung? Republican or Democrat? Sell or buy AAPL? The rate of economic growth this year? Social cost of carbon?


French 1956; Benerjee 1992; Galam 1996

## Dynamics of Opinions

$$
x_{i}(k+1)=f_{i}\left(x_{i}(k) ; x_{j}(k), j \in \mathcal{N}_{i}\right)
$$



## De Groot Social Interactions

$$
\mathbf{x}(k+1)=\mathbf{P} \mathbf{x}(k)
$$

- $\mathbf{P}$ is a stochastic matrix


## De Groot Social Interactions



國 DeGroot 1974

## De Groot Social Interactions

$$
\mathbf{x}(k+1)=\mathbf{P} \mathbf{x}(k)
$$

An agreement is achieved if $\mathbf{P}$ is ergodic in the sense that

$$
\lim _{k \rightarrow \infty} x_{i}(k)=\mathbf{v}^{\top} \mathbf{x}(0)
$$

Trust and cooperation lead to social consensus!

## Wisdom of Crowds (with Trust)

Golub and Jackson 2010


## Disagreement Models

- Memory of initial values

Friedkin and Johnsen (1999)

- Bounded confidence

國 Krause (1997); Hegselmann-Krause (2002); Blondel et al. (2011); Li et al. (2013)

- Stubborn agents

Acemoglu et al. (2013)

- Homophily

Dandekar et al. (2013)

## This Talk

A model and theory for opinion dynamics over social networks with friendly and adversarial interpersonal relations coexisting.

## Friends and Adversaries



## Structural Balance Theory

- Strongly balanced if the node set can be divided into two disjoint subsets such that negative links can only exist between them;
- Weakly balanced if such a partition contains maybe more than two subsets.

Heider (1947), Harary (1953), Cartwright and Harary (1956), Davis (1963)

## Structural Balance



## When do nodes interact?

## Underlying World

$$
\mathrm{G}=\mathrm{G}^{+} \cup \mathrm{G}^{-}
$$

- Fixed
- Undirected
- Deterministic
- Connected

$$
\mathcal{N}_{i}=\mathcal{N}_{i}^{+} \cup \mathcal{N}_{i}^{-}
$$



## Gossip Model



## Gossip Model

Independent with other time and node states, at time $k$, (i) A node $i$ is drawn with probability $1 / N$;
(ii) Node $i$ selects one of its neighbor $j$ with probability $1 /\left|\mathcal{N}_{i}\right|$.

DOI:10.1145/2184319.2184338 A few hubs with many connections share with many individuals with few connections.

BY BENJAMIN DOERR, MAHMOUD FOUZ, AND TOBIAS FRIEDRICH

## Why Rumors

 Spread So Quickly in Social Networks

This visualization by Miguel Rios at Twitte shows the volume of @replies traveling int and out of Japan and worldwide retweet in the one-hour period just before and after the Töhoku earthquake on March 11, 2011 For an animated version visit http://blog twitter.com/2011/06/global-pulse.htm
studied network topologies need at least logarithmic time. Surprisingly, nodes with few neighbors are crucial
for mint dicoominotion

How a pair of nodes interacts with each other when they meet?

## Positive and Negative Interactions

A pair ( $\mathrm{i}, \mathrm{j}$ ) is randomly selected. The two selected nodes update.


$$
x_{i}(k+1)=x_{i}(k)+\alpha\left(x_{j}(k)-x_{i}(k)\right) \quad 0<\alpha<1
$$



$$
x_{i}(k+1)=x_{i}(k)+\beta\left(x_{i}(k)-x_{j}(k)\right)
$$

$$
\beta>0
$$

Altafini 2013

## Positive and Negative Interactions

Before:

$$
x_{i}(k+1)=x_{i}(k)+\alpha\left(x_{j}(k)-x_{i}(k)\right)
$$

After:

$$
x_{i}(k+1)=x_{i}(k)-\beta\left(x_{j}(k)-x_{i}(k)\right)
$$

After:


Mean/Mean-Square Evolution

## Relative-State-Flipping Model

$$
\begin{aligned}
& \mathbf{x}(k+1)=W(k) \mathbf{x}(k) \\
& \mathbb{E}\{W(k)\}=I-\alpha L_{\mathrm{pst}}^{\dagger}+\beta L_{\mathrm{neg}}^{\dagger} .
\end{aligned}
$$

- It's an eigenvalue perturbation problem!


## Phase Transition

Theorem. Suppose $\mathrm{G}^{+}$is connected and $\mathrm{G}^{-}$is non-empty. Then there exists $\beta_{*}$ such that
(i) $\lim _{k \rightarrow \infty} \mathbb{E}\left\{x_{i}(k)\right\}=\sum_{i=1}^{N} x_{i}(0) / N$ for all $i=1, \ldots, N$ if $\beta<\beta_{*}$;
(ii) $\lim _{k \rightarrow \infty} \max _{i, j}\left\|\mathbb{E}\left\{x_{i}(k)\right\}-\mathbb{E}\left\{x_{j}(k)\right\}\right\|=\infty$ if $\beta>\beta_{*}$.

- It's possible to prove that the expectation of the state transition matrix is eventually positive.


## Phase Transition

Let G be the complete graph. Let $\mathrm{G}^{-}$be the Erdos-Renyi random graph with link appearance probability $p$.
(i) If $p<\frac{\alpha}{\alpha+\beta}$, then

$$
\mathbf{P}(\text { Consensus in expectation }) \rightarrow 1
$$

as the number of nodes $N$ tends to infinity;
(ii) If $p>\frac{\alpha}{\alpha+\beta}$, then

$$
\mathbf{P}(\text { Divergence in expectation }) \rightarrow 1
$$

as the number of nodes $N$ tends to infinity.

## Sample Path Behavior

## Live-or-Die Lemma

Introduce
$\mathscr{C}_{x^{0}} \doteq\left\{\limsup _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=0\right\}, \quad \mathscr{D}_{x^{0}} \doteq\left\{\limsup _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=\infty\right\}$
$\mathscr{C}_{x^{0}}^{*} \doteq\left\{\liminf _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=0\right\}, \quad \mathscr{D}_{x^{0}}^{*} \doteq\left\{\liminf _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=\infty\right\}$

## Lemma.

Suppose $\mathrm{G}^{+}$is connected. Then (i) $\mathbb{P}\left(\mathscr{C}_{x^{0}}\right)+\mathbb{P}\left(\mathscr{D}_{x^{0}}\right)=1 ;$ (ii) $\mathbb{P}\left(\mathscr{C}_{x^{0}}^{*}\right)+\mathbb{P}\left(\mathscr{D}_{x^{0}}^{*}\right)=1$. As a consequence, almost surely, one of the following events happens:
$\left\{\lim _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=0\right\} ;$
$\left\{\lim _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=\infty\right\} ;$
$\left\{\liminf _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=0 ; \limsup _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=\infty\right\}$.

## Zero-One Law

$\mathscr{C} \doteq\left\{\limsup _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=0\right.$ for all $\left.x^{0} \in \mathbb{R}^{n}\right\}$,
$\mathscr{D} \doteq\left\{\exists\right.$ (deterministic) $x^{0} \in \mathbb{R}^{n}$, s.t. $\left.\limsup _{k \rightarrow \infty} \max _{i, j}\left|x_{i}(k)-x_{j}(k)\right|=\infty\right\}$

Theorem. Both $\mathscr{C}$ and $\mathscr{D}$ are trivial events (i.e., each of them occurs with probability equal to either 1 or 0 ) and $\mathbb{P}(\mathscr{C})+\mathbb{P}(\mathscr{D})=1$.

## No-Survivor Theorem

Theorem. There always holds

$$
\mathbb{P}\left(\liminf _{k \rightarrow \infty}\left|x_{i}(k)-x_{j}(k)\right|=\infty\left|\liminf _{k \rightarrow \infty} \max _{i, j}\right| x_{i}(k)-x_{j}(k) \mid=\infty\right)=1
$$

for all $i \neq j$.

## Phase Transition

## Theorem.

(i) Suppose $\mathrm{G}^{+}$is connected. Then there is a $\beta_{*}>0$ such that

$$
\mathbb{P}\left(\lim _{k \rightarrow \infty} x_{i}(k)=\sum_{i=1}^{N} x_{i}(0) / N\right)=1
$$

for all $i$ if $\beta<\beta_{*}$.
(ii) There is $\beta^{*}>0$ such that

$$
\mathbb{P}\left(\liminf _{k \rightarrow \infty} \max _{i, j}\left\|x_{i}(k)-x_{j}(k)\right\|=\infty\right)=1
$$

for all $i$ if $\beta>\beta^{*}$.

## Bounded State Model

## Bounded States

- Let $A>0$ be a constant and define $\mathscr{P}_{A}(\cdot)$ by $\mathscr{P}_{A}(z)=-A, z<-A, \mathscr{P}_{A}(z)=$ $z, z \in[-A, A]$, and $\mathscr{P}_{A}(z)=A, z>A$.
- Define the function $\theta: \mathrm{E} \rightarrow \mathbb{R}$ so that $\theta(\{i, j\})=\alpha$ if $\{i, j\} \in \mathrm{E}^{+}$and $\theta(\{i, j\})=-\beta$ if $\{i, j\} \in \mathrm{E}^{-}$.

Consider the following node interaction under relative-state flipping rule:

$$
x_{s}(t+1)=\mathscr{P}_{A}\left((1-\theta) x_{s}(t)+\theta x_{-s}(t)\right), s \in\{i, j\} .
$$

## Clustering of Opinions

Theorem. Let $\alpha \in(0,1 / 2)$. Assume that G is a weakly structurally balanced complete graph under the partition $\mathrm{V}=\mathrm{V}_{1} \cup \mathrm{~V}_{2} \cdots \cup \mathrm{~V}_{m}$ with $m \geq 2$. Let $\alpha \in(0,1 / 2)$. When $\beta$ is sufficiently large, almost sure boundary clustering is achieved in the sense that for almost all initial value $\mathbf{x}(0)$ w.r.t. Lebesgue measure, there are there are $m$ random variables, $l_{1}(\mathbf{x}(0)), \ldots, l_{m}(\mathbf{x}(0))$, each of which taking values in $\{-A, A\}$, such that:

$$
\mathbb{P}\left(\lim _{t \rightarrow \infty} x_{i}(t)=l_{j}(\mathbf{x}(0)), i \in \mathrm{~V}_{j}, j=1, \ldots, m\right)=1 .
$$

## Separation Events



- The power of minority groups.


## Numerical Example



## Oscillation of Opinions

Theorem. Let $\alpha \in(0,1 / 2)$. Assume that G is a complete graph and the positive graph $\mathrm{G}^{+}$is connected. When $\beta$ is sufficiently large, for almost all initial value $\mathbf{x}(0)$ w.r.t. Lebesgue measure, there holds for all $i \in \mathrm{~V}$ that

$$
\mathbb{P}\left(\liminf _{t \rightarrow \infty} x_{i}(t)=-A, \limsup _{t \rightarrow \infty} x_{i}(t)=A\right)=1 .
$$

## Numerical Example




## Related Publications

Shi et al. 2013 IEEE Journal on Selected Areas in Communications; Shi et al. 2015, 2016 IEEE Transactions on Control of Network Systems; Shi et al. 2016 Operations Research.

## Thank you!

