

The Effects of Homophily and Influence On Voting Games

Thesis Defense in Requirements of Thesis Track for Master of Science in
Computer Science at Columbia University

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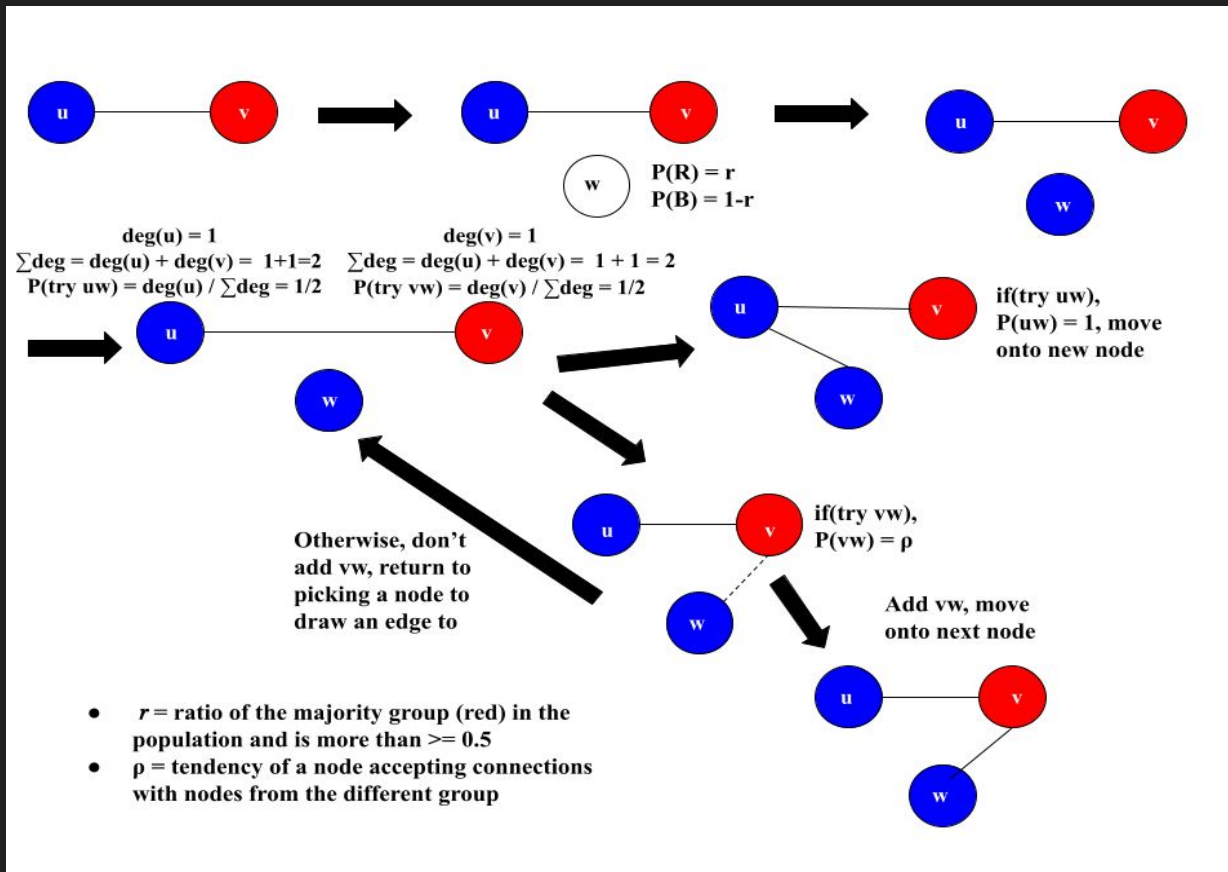
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Other Models....

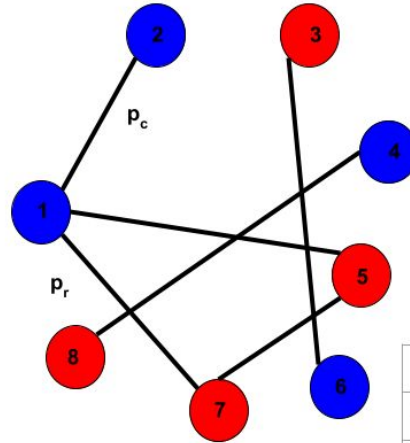
- Voter Model
 - Voters choose opinion proportional to neighbors
 - Must converge at full consensus
- Independent Cascade Model
 - Nodes start as inactive except for starting set, can only be activated
 - Nodes get one shot at activating neighbors
- Linear Threshold Model
 - Nodes start as inactive except for starting set, can only be activated
 - Node is activated when sum of indegree weight from activated set matches threshold

My Model

Biased Preferential Attachment (BPA)

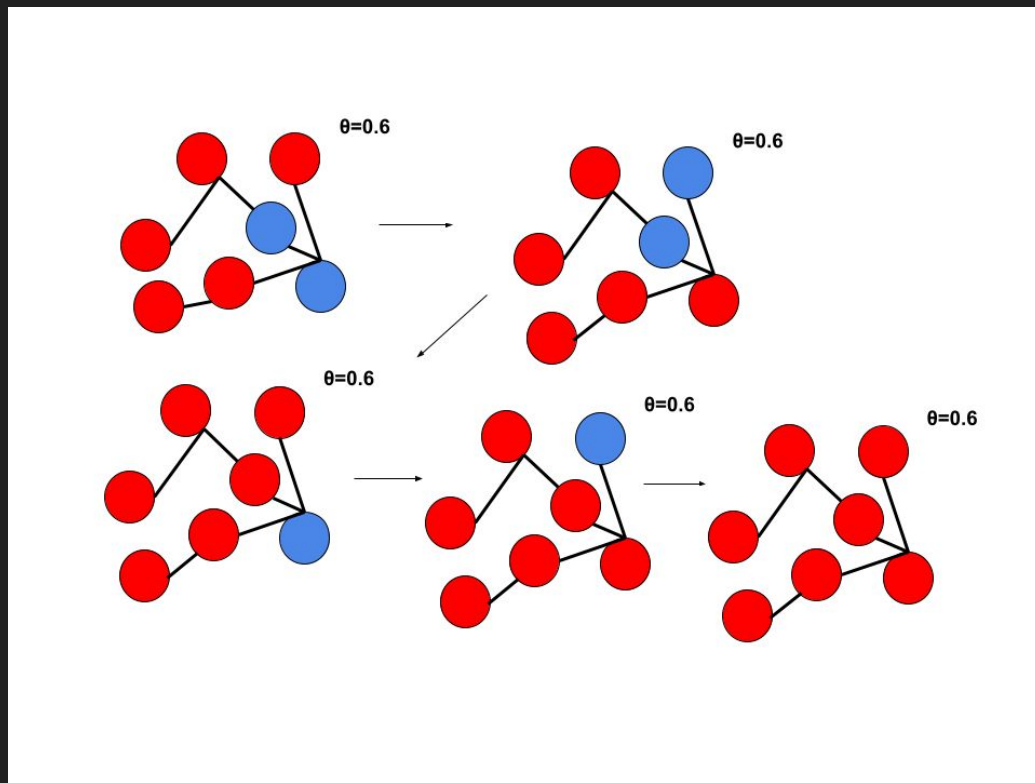


Stochastic Block (SB)



	1	2	3	4	5	6	7	8
1	0	p_c	p_r	p_c	p_r	p_c	p_r	p_r
2	p_c	0	p_r	p_c	p_r	p_c	p_r	p_r
3	p_r	p_r	0	p_r	p_c	p_r	p_c	p_c
4	p_c	p_c	p_r	0	p_r	p_c	p_r	p_r
5	p_r	p_r	p_c	p_r	0	p_r	p_c	p_c
6	p_c	p_c	p_r	p_c	p_r	0	p_r	p_r
7	p_r	p_r	p_c	p_r	p_c	p_r	0	p_c
8	p_r	p_r	p_c	p_r	p_c	p_r	p_c	0

Threshold Influence



Justification and Novelty

- Influence parameter allows for an individual to consider how much they value their own independent thought
- Realistic analogue of convergence process a bit difficult but can think of as agents influencing each other over the course of something such as an election cycle, where discrete time intervals are like polls
- Model does not appear to be discussed or analyzed elsewhere
 - Similar to Linear Threshold but allows nodes to return to “deactivated” state
- Entire Network behavior can be thought of as Markovian process with $O(2^n)$ states

Theoretical Results

Graph Properties

- **BPA**
 - Node Degree Distribution follows a **Power Law / Scale-Free**
 - **Acyclic**
 - **Sparse**
 - **Rich-get-richer / Preferential Attachment**
 - **Homophilic**
- **Stochastic Block**
 - **Generally dense**
 - **Many cycles**
 - **Homophilic**

We can follow through with modelling each process to understand distribution of red neighbors

$$\mathbb{E}[\alpha_{t+1}|\alpha_t] = \alpha_t + \frac{F(\alpha_t) - \alpha_t}{t+1}, \quad \text{where}$$

$$F(x) = \left(1 - (1-r)\frac{(1-x)}{1-x(1-\rho)} + r\frac{x}{1-(1-x)(1-\rho)}\right)/2.$$

Theorem 3.8. The red fraction distribution $\alpha(u)$ for $u \in V$ of a network of n nodes formed by SB with a red node proportion of r , intra-homophily factor of p_c and inter-homophily factor p_r has a cumulative distribution function as such:

$$P(\alpha(u) \geq \alpha^*) = \sum_{i=1}^{n-1} \frac{\sum_{k=\lceil \alpha^* i \rceil}^i F_{ki}}{\sum_{j=0}^i F_{ji}}$$

where

$$F_{xi} = \binom{nr-1}{x} p_c (1-p_c)^{nr-1-x} \binom{n-nr}{i-x} p_r^{i-x} (1-p_r)^{n-nr-i+x}$$

Theorem 3.7. The red fraction distribution $\alpha(u)$ for a red node $u \in V$ of a network of n nodes formed by BPA with a red node proportion of r , homophily factor of p has a cumulative distribution function as such:

$$P(\alpha(u) \geq \alpha^*) = \sum_{i=1}^n \sum_{j=1}^{|2^M|} \sum_{k=1}^{|2^{Q_j}|} \frac{\prod_{a=1}^{|T_k|} F_{ka} \prod_{b=1}^{|S_k|} 1 - F_{kb}}{\prod_{c=1}^{|Q_j|} W_{jc} \prod_{d=1}^{|R_j|} 1 - W_{jd}}$$

where

$$M = \{i, \dots, n\}$$

$$Q = \{q \mid q \in 2^M\} \text{ for } 2^M = \text{power set of } M$$

$$T = \{t \mid t \in 2^{Q_j} \wedge |t| \geq \lceil \alpha^* |Q_j| \rceil\} \text{ for } 2^{Q_j} = \text{power set of } Q_j$$

$$R = \{\beta \mid \beta \in \mathbf{Z} \wedge i \leq \beta \leq n \wedge \beta \notin Q_j\}$$

$$S = \{\omega \mid \omega \in \mathbf{Z} \wedge i \leq \omega \leq n \wedge \omega \notin T_k\}$$

$$W_{jx} = \frac{r(x+1)}{2(Q_{jx}+1) - (2(Q_{jx}+1) - (x+1))(1-p)} + \frac{(1-r)(x+1)p}{2(Q_{jx}+1) - (x+1)(1-p)}$$

$$F_{kx} = \frac{r(x+1)}{2(T_{kx}+1) - (2(T_{kx}+1) - (x+1))(1-p)}$$

Influence Convergence

Forbidden Subgraphs

Influence Process ***fails*** to converge if we have a forbidden subgraph such that

- Every node within the subgraph fails to meet the threshold condition for its own current color
- Number of neighbors node has within subgraph (constantly flipping) must exceed number of neighbors outside subgraph (constant)

Lemma 3. *If $G_{\theta,i}$ is such that $\forall u_i \in G_{\theta,i} \mid \frac{|N(u_i)_{C(u_i)}|}{|N(u_i)|} < \theta \wedge \frac{|N(u_i) - C(u_i)|}{|N(u_i)|} \geq \theta$ then $\mathbf{I}_{G,\theta}$ will not converge.*

Lemma 4. *If $G_{\theta,i}$ is colored to match a bipartite assignment of its nodes, then $\mathbf{I}_{G,\theta}$ will not converge.*

Influence Maximization

	1.0	r	0
RR	(b,1]	[0,b]	\emptyset
RB	(b,1]	[0,r]	(r,b]
BR	(b,r]	[0,b]	(r,1]
BB	\emptyset	[0,r]	(r,1]
$R \frac{1}{2}$	(0.5,1]	[0,0.5]	\emptyset
$B \frac{1}{2}$	\emptyset	[0,0.5]	[0.5,1)

Table 1: Where UV corresponds to node u of color U and neighbor majority of color V, we look at the intervals required to have a weight of 1, the red neighbor fraction r , or 0

Proposed Algorithms

- Optimize for threshold that maximizes sets
- Optimize for degree of the above

Theorem 3.4. The optimal interval $\theta_{OPT} \in [\theta_{OPT,min}, \theta_{OPT,max}]$ that maximizes $|G_{R,\theta,k}|$ given that $\mathbf{I}_{\{G,\theta,k\}}$ converges at $i \leq k$ will be such that $\theta = \operatorname{argmax}_{\theta} \sum_{u \in BR_{\theta}} \psi(u) - \sum_{v \in RB_{\theta}} \psi(v)$ where $BR_{\theta} = \{u_0 \in G_0 \wedge C(u_0) = B \wedge C(u_1) = R\}$ and $RB_{\theta} = \{u_0 \in G_0 \wedge C(u_0) = R \wedge C(u_1) = B\}$

Key claim:

If a node at time i u_i 's red fraction $rf(u_i)$ increased from timestep $i - 1$, the probability of node u_i itself being red at time $i + 1$ is non-decreasing (increasing for non- $R \frac{1}{2}$ - $B \frac{1}{2}$ cases).

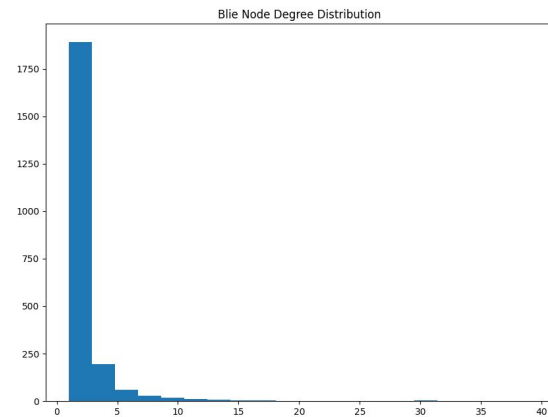
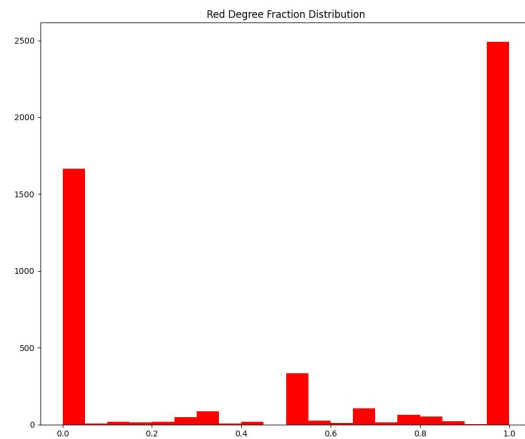
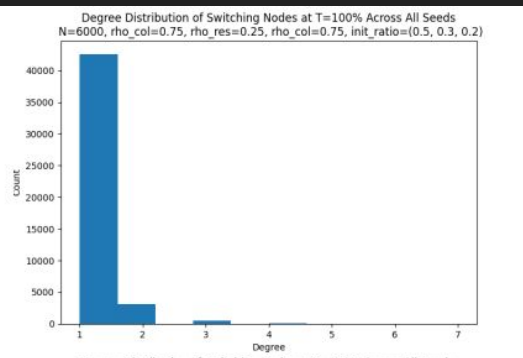
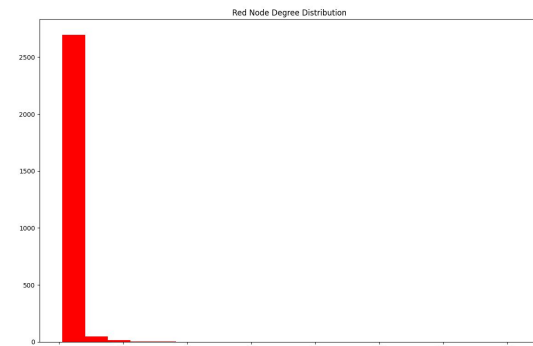
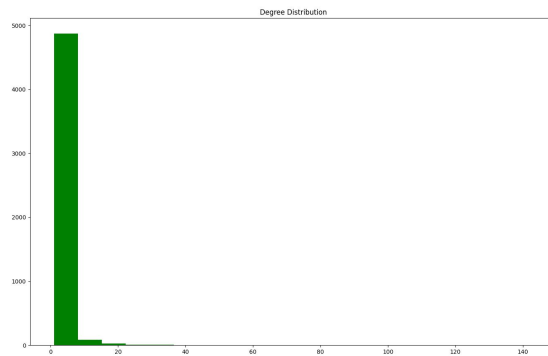
Algorithm to Maximize Red Votes at $t=t+1$

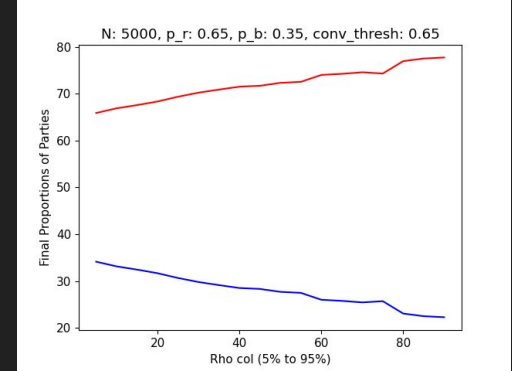
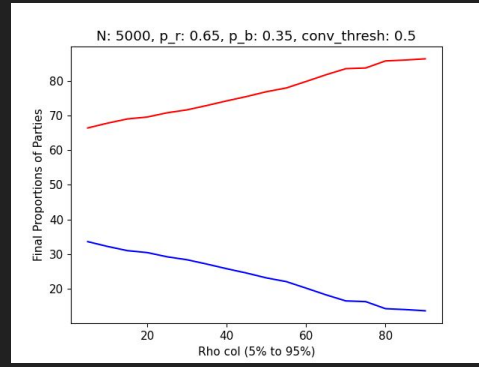
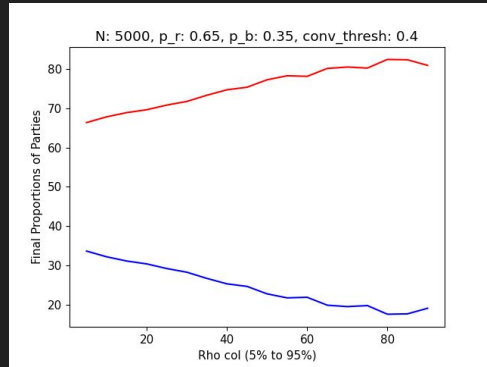
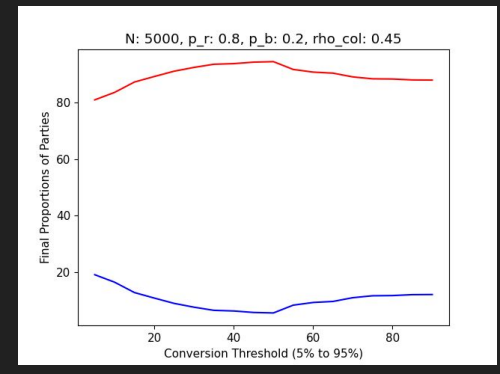
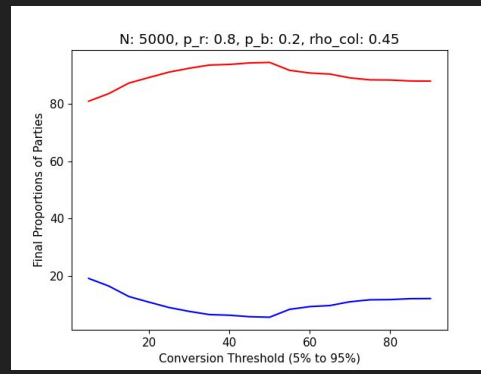
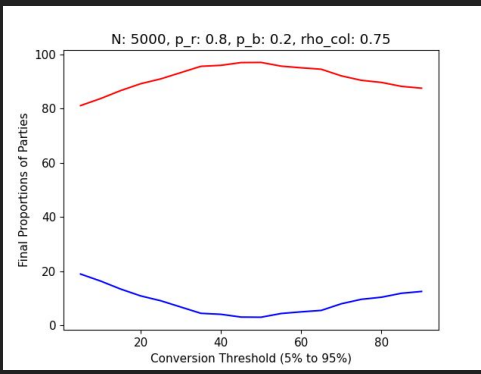
1. For all nodes u , if red fraction q is less than 0.5, generate interval $[0,q)$ with weight q . If u itself happens to be red also add interval $(1-q,1]$ with weight 1. If red fraction is exactly 0.5, generate interval $[0,0.5]$ with weight 0.5, if u itself happens to be red, add $(0.5,1]$ with weight 1. If $q > 0.5$, add interval $(q,1]$ with weight 1 and instead $(1-q,1]$ with weight 1 if u happens to be red, and $[0,1-q]$ with weight q .
2. Mark each interval tuple with (Start,End) tags, then sort all the endpoints of the interval
3. Linearly traverse your sorted list from beginning to end, adding the weight w_i if the current endpoint was the start of interval i , and subtract weight w_i if the current endpoint is the end of interval i . Keep track of your sum.
4. Return the interval in which the sum was maximal as the optimal threshold

*Putting It All Together: Homophily,
Population Ratio, Influence Threshold*

- Homophily + Minority Ratio Affect BPA => BPA Defines $G_{0,\theta}$
=> Influence defines $G_{\infty,\theta}$
- Homophily increases interactions between majority and minority group
- Greater impact on increase in majority fraction on minority group than decrease in majority fraction on majority group
- Influence:
 - Threshold very high => Too high for many nodes to change
 - Threshold very low => Approaches voter model, change proportional directly to majority fraction, may encourage perturbations for minority
 - Middle Threshold => Large enough for majority to be valid but minority to miss the cut, has greatest encouragement of minority switches

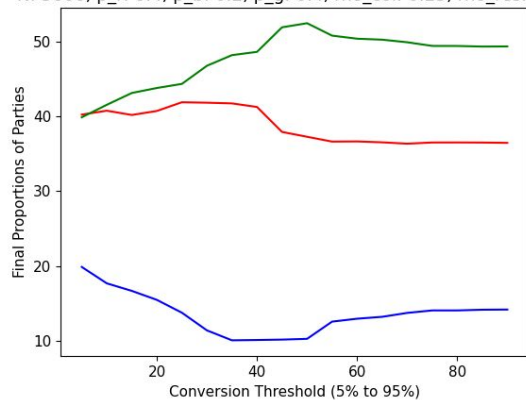
Simulation



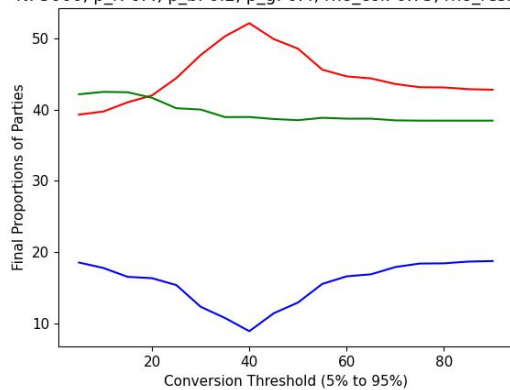


Extension: Three-Party Simulation

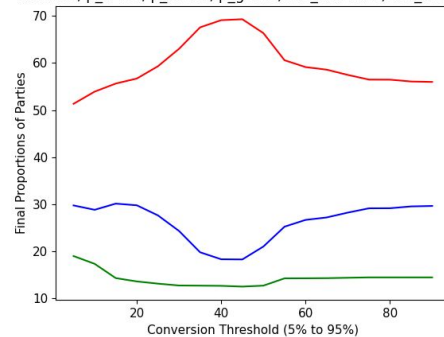
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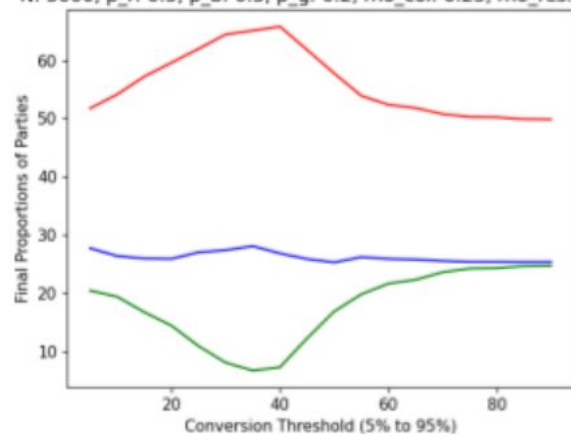
N: 5000, p_r : 0.4, p_b : 0.2, p_g : 0.4, ρ_{col} : 0.75, ρ_{res} : 0.25



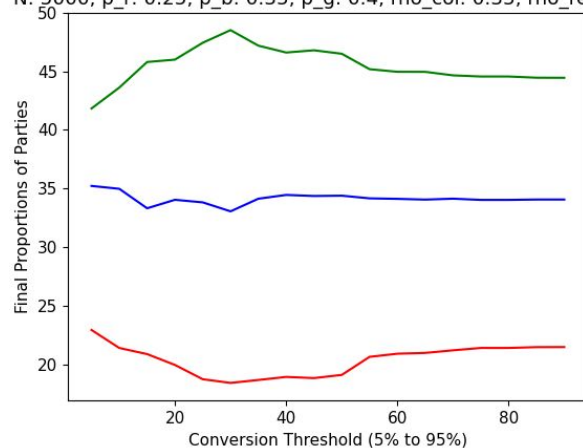
N: 5000, p_r : 0.5, p_b : 0.3, p_g : 0.2, ρ_{col} : 0.75, ρ_{res} : 0.25



N: 5000, p_r : 0.5, p_b : 0.3, p_g : 0.2, ρ_{col} : 0.25, ρ_{res} : 0.75



N: 5000, p_r : 0.25, p_b : 0.35, p_g : 0.4, ρ_{col} : 0.35, ρ_{res} : 0.5



Observations

- **The general 2-party observations seem to generalize to 3 parties**
- Dominant (> 0.5) party still cannot lose
- We can manipulate the parameters to allow the 3rd place party to come in 2nd by the end, or the #2 to overtake 1st place

Further Work and Concluding Remarks

Takeaways

- Developed new model for influence propagation / evolutionary behavior
- Increasing homophily reduces the minority party's power
- Optimal region for influence threshold that minimizes the minority party's power
- Influence method converges under lack of forbidden substructure, several possible ways to optimize it

Further work

- Optimality of Influence Optimization? Average-case rate of convergence?
- **How does the theory generalize to 3+ parties?**
- Can we explore other types of voting systems, ex: Ranked Choice, Condorcet

Questions?