The Effect of Network Structure on Preference Formation

Samara Klar, University of Arizona Yotam Shmargad, University of Arizona

How does network structure influence opinion? Relying on theories of preference formation and social networks, we randomize a sample of adults into networks that vary in structure. In one (a clustered lattice), individuals' connections tend to be connected to each other; in another (a random network), individuals' connections tend not to be connected, instead providing access to different regions of the network. We seed messages that reflect competing sides of policy debates in each network: one underdog viewpoint is seeded less often, while a dominant viewpoint is seeded more often. We track their diffusion and find that the random network increases exposure to underdog views, compared with the clustered lattice. Individuals in the random network subsequently learn more about the policy debates and become more sympathetic toward the underdog perspective. This has implications for how less funded information campaigns can strategically target social networks to maximize exposure and change minds.

here may be two sides to every argument, but one perspective is often more prevalent than the other. As Stimson (2004) explains, "There are many matters where one side is dominant and is the only message the public hears" (18). In American political discourse, an asymmetric playing field is the norm rather than the exception. During the 2012 presidential campaign, for example, advertising expenditures by the two competing parties were split as equitably as 40:60 in only seven states. In the average state, 77% of campaign advertising expenses were spent by just one party,¹ creating competition between a dominant campaign and an underdog.

Exposure to campaign messages depends in large part on who people know, as political information is often communicated in social settings (Sinclair 2012). We investigate how the structure of the social network underlying information transfer can help to mitigate disparities in exposure to messages from asymmetrically resourced campaigns. Building on existing social network theories, we hypothesize that certain networks exacerbate the dominance of high-resourced campaigns, while others more readily expose people to messages from low-resourced underdogs. We then turn to literature from political psychology to formulate and test hypotheses about how such exposure can facilitate learning and influence political preferences.

The effects of network structure are difficult to ascertain with observational data (Manski 1993). People in certain types of networks may be different from those in other networks, and so the effects of structure are easily conflated with personal traits. In this study, we manipulate network structure by randomizing a sample of adults into one of two social networks that vary in structure: in one, people's contacts tend to be connected to each other and to reside in the same region of the network; in another, people's contacts tend not to be connected and to reside in different network regions. We asymmetrically seed competing pieces of information in each network and track their spread. By comparing these cases, we find that people connected to distant network regions are exposed to dominant and underdog information at near equal rates. As a result, they learn about both sides

1. http://www.slate.com/articles/news_and_politics/map_of_the_week/2012/10/campaign_spending_map_shows_where_obama_romney_campaigns _spend_.html (accessed February 25, 2016).

The Journal of Politics, volume 79, number 2. Published online January 24, 2017. http://dx.doi.org/10.1086/689972 © 2017 by the Southern Political Science Association. All rights reserved. 0022-3816/2017/7902-0024\$10.00

Samara Klar (klar@email.arizona.edu) is assistant professor at the University of Arizona, School of Government and Public Policy, Tucson, AZ 85721. Yotam Shmargad (yotam@email.arizona.edu) is assistant professor at the University of Arizona, School of Information, Tucson, AZ 85721.

This study was conducted in compliance with relevant laws and was approved by the appropriate institutional and/or national research ethics committee. Support for this research was provided by National Science Foundation grant no. 1160104 to Samara Klar. Data and supporting materials necessary to reproduce the numerical results in the paper are available in the *JOP* Dataverse (https://dataverse.harvard.edu/dataverse/jop). An online appendix with supplementary material is available at http://dx.doi.org/10.1086/689972.

of an issue and shift their attitudes toward the underdog. When contacts tend to be connected to the same region, on the other hand, a single perspective dominates. We use computer simulations to generalize our findings. Our study provides important evidence regarding the influence of network structure on preferences.

THE INFLUENCE OF SOCIAL NETWORK STRUCTURE ON INFORMATION EXPOSURE

Encounters with unfamiliar information may provide individuals with opportunities to become more informed, but individuals tend to dismiss information that casts doubt on their predetermined allegiances (see Kraft et al. [2015] for a review). Yet, despite their distaste for counter-attitudinal information, it can still make its way into people's information environments in unexpected ways (e.g., Walsh 2004; Wojcieszak and Mutz 2009).

We expect that when the structure of an individual's social network provides access to distant regions of the network, exposure to both perspectives on an issue should increase (hypothesis 1). A key contributor to the transmission of novel information is what network scientists call *weak ties*. Strong ties, by virtue of being one's close contacts, tend to know each other. Weak ties, like casual acquaintances, are less likely to know each other and more likely to reach distant regions of the broader social network (Granovetter 1973). The advantage of weak ties is their tendency to connect individuals to regions of the network to which they otherwise could not access and, subsequently, to expose them to new information (Barbera 2014). Indeed, weak ties drive the majority of information diffusion on social network platforms like Facebook (Bakshy et al. 2012).

Individuals who are able to reach distant regions of a social network through their contacts should therefore be more likely to learn about new perspectives on an issue as compared with those whose contacts who are connected to similar regions (hypothesis 2). Exposure to opposing view-points increases the ability of individuals to justify their own perspective and to understand why others may disagree with them (Price and Cappella 2002). Such exposure should increase citizens' learning and engagement with an issue (Levitan and Wronski 2014).

In addition, people with exposure to underdog views should form opinions that better reflect those views (hypothesis 3). Existing work demonstrates that diverse settings promote even-handed choices (Price and Cappella 2002), loosen prior attachments (Mutz 2002), and weaken opinions (Klar 2014). Whereas Huckfeldt et al. (1995) show that social networks with low cohesion more accurately reflect the distribution of opinions in the broader information environment, we expect a previously untested benefit of social networks that facilitate access to distant regions: the distribution of opinions should be less dependent on the balance of information seeded in the network.

PROCEDURE

A key challenge of studying the effects of information exposure is that "we do not know how many and what kind of people are exposed to which messages" (Prior 2013, 102). In our study, we observe the information to which people are exposed with two controlled cases. We invited 756 adult subjects to participate in a mock social network,² which we called Political Pulse. Forty-six percent agreed to participate (n = 348). Following Centola (2010), we randomly assigned each subject to a node in either a clustered lattice or a random network.3 The former is characterized by large distances between nodes; in the latter, nodes reach many regions of the network through their contacts. Each subject in our study was connected to exactly six other subjects. In each network, we seeded information about two policy issues: genetically modified organisms (GMOs) and electric cars. For each issue, two individuals received information about one perspective (either pro or con), while only one received information about the opposing perspective. This created asymmetry in seeding across perspectives.

On each day of our study, subjects received e-mails alerting them about information that their connections viewed the previous day, along with links to the information that those contacts had viewed. Each link contained a unique eight-digit number identifying the subject receiving the e-mail, so we could use Google Analytics to infer which subjects viewed what content when. We allowed information to flow through both networks for eight days. At the end of the study, we administered a final survey to ask subjects about their learning experience and opinions on the issues. Fifty-eight percent of the subjects completed this final survey (n = 201).

RESULTS

Our first hypothesis states that individuals in the random network will experience greater relative exposure to lowseeded information in comparison to those in the clustered lattice. The top panel of figure 1 depicts cumulative views of the dominant information in support of GMOs and the

^{2.} ResearchNow, a research firm, provided this study's subjects. See the appendix for full detail on the sample, design, content, and survey questions we used.

^{3.} We randomly assigned 60 subjects to a control group. This group completed a survey about the issues in our study, which provides a baseline level of support for these issues.



Figure 1. Across two issues, support or not of genetically modified foods and electric cars, the dominant perspective was viewed more frequently than the underdog perspective in the clustered lattice but not in the random network.

underdog information against GMOs over time. In the clustered lattice, the dominant perspective received more than twice as many views as the underdog perspective. The bottom panel of figure 1 illustrates the diffusion of underdog pro-electric car and dominant anti-electric car information. Across both issues, the dominant perspective was viewed more frequently than the underdog perspective in the clustered lattice but not in the random network.

Hypotheses 2 and 3: Learning and preference formation

The left bars in figure 2 depict subjects' responses to the questions: "To what degree did you become more informed about genetically modified organisms?" and "To what degree did you become more informed about electric cars?" A fully-labeled 7-point response scale was provided, ranging from "I did not learn any new information" (1) to "I gained a very large amount of new information" (7). We collapse across issues and show the percentage of subjects who report that they learned a large or very large amount of new information. In the clustered lattice (dark bars), 10% of the subjects reported becoming more informed, while 19% of those in the random network (light bars) became more informed.

Next, we examine perhaps the most consequential outcome of network structure: opinion change. Recall that, in each network, we seed one perspective (i.e., the underdog's) half as frequently as the other. Our examination so far suggests that networks enabling broad access allow underdogs to achieve exposure equal to dominant perspectives. In networks where one's contacts are connected to the same region (i.e., clustered lattices), underdogs are underexposed. Our third hypothesis states that wider exposure causes opin-



Figure 2. Participants reported greater learning and support for the underdog in the random network than in the clustered lattice. ions to shift in favor of the underdog. We first examine the percentage of subjects who agreed with the underdog view-point in the final survey. The questions read: "Would you say you oppose or support genetically modified foods?" and "Would you say you oppose or support electric cars?" Responses ranged from "definitely oppose" (1) to "definitely support" (7). In figure 2, we collapse across the issues and show, in the right two bars, the percentage of subjects who report that they "support" or "definitely support" the underdog perspective. In the clustered lattice, 28% support the underdog, which is lower than the 38% who support the underdog in the random network.

Our results should be viewed as preliminary evidence of the effects of network structure on underdog exposure, learning, and opinion formation. In our appendix, available online, we provide three statistical analyses and discuss in more detail the assumptions about independence that underlie them. First, we model learning and support for the underdog to explore how the effect of network structure varied by party affiliation. In doing so, we find that the additional exposure to underdog viewpoints only changed attitudes among those who did not already agree with them. Second, we incorporate the control (no network) condition to show that opinions changed because subjects in the random network moved away from the baseline level of support; opinions among subjects in the clustered lattice did not significantly differ from the baseline. Finally, we show that there was no selective attrition that might have affected the findings.

Robustness: Computer simulations

An underlying assumption of many statistical tests is that observations are statistically independent from one another. A limitation of our study is that we explicitly impose dependence: a participant's likelihood of viewing information is dependent on the viewing decisions of those assigned to the same network condition. While our two networks serve as useful cases, statistical analyses comparing information exposure in the clustered lattice to those in the random network rely on the independent observations assumption to hold.

We therefore ran computer simulations to test if differences in exposure across networks (hypothesis 1) hold across a large number of replications and under different assumptions about network structure and the diffusion process. We first start with the clustered lattice and rewire its links to create one thousand random networks. We simulate a thousand diffusion processes of underdog (once-seeded) and dominant (two-seeded) information in each random network, and a million processes of each type of information in the clustered lattice. These simulations confirm that, in clustered lattices, views of dominant information surpass those of the underdog. In the random network, on the other hand, there was not a significant difference in the views.

We then varied network structure by altering the probability that links in the clustered lattice are randomly rewired, testing 10 values: 0 (clustered lattice), 10^{-2} , $10^{-1.75}$, $10^{-1.5}$, $10^{-1.25}$, 10^{-1} , $10^{-0.75}$, $10^{-0.5}$, $10^{-0.25}$, and 1 (random network). We also tested click rates of 0.09, 0.14, 0.19 (the click rate in our study), 0.24, and 0.29. For each network and click rate, we simulated 20 thousand diffusion processes, split across dominant and underdog information. These simulations confirm that what we observe for random networks holds across click rates, as well as for low rewiring probabilities. Complete details can be found in the appendix.

CONCLUSION AND DISCUSSION

We present two cases in which subjects are randomly assigned to networks that vary in structure. The diffusion we observe in these networks suggests that campaigns competing with fewer resources can compensate for their resource disadvantages by seeding messages in networks with connections between distant regions. We find that exposure to these viewpoints can have important consequences: people report that they learn more from this exposure and, ultimately, preferences shift in favor of the underrepresented perspective (particularly among those who did not already support it, as we show in the appendix).

Much of what we know about political mobilization (Bond et al. 2012), the spread of health behaviors (Centola 2010), and emotional contagion (Kramer, Guillory, and Hancock 2014) emerges from controlled studies of online networks. Such studies suffer from particular limitations, and ours is no exception. First, these studies are carried out online, and so they may or may not generalize to face-to-face interactions, which can have unique implications (e.g., Klar 2014). Second, the subjects employed in this study are anonymous, which might facilitate different behaviors than what we might see when identities are known, especially since information about an individual's contacts and their actions can influence his behavior (McCubbins, Enemark, and Weller 2014). Third, subjects in our study automatically received links to information that their contacts viewed. This passive sharing expedites diffusion beyond what we might expect with active sharing. Finally, perhaps the main limitation of this exploratory study is the small sample size: we only study two network cases. We urge scholars to extend this work to a larger scale and across different settings.

Karl Llewellyn described the tendency to rely only on widely available information as "the threat of the available" (1931, 95). Biased information environments encourage attitudinal polarization (Klar 2014) and depress knowledge (Jerit and Barabas 2012), and they can lead to inequality (Mansbridge 1983) and segregation (Mendelberg and Olseke 2000). Our study suggests that under-resourced viewpoints can gain traction in particular network structures, helping to mitigate bias in the information to which individuals are exposed.

ACKNOWLEDGMENTS

We are very grateful to Jennifer Merolla and three anonymous reviewers for invaluable feedback. We are indebted to John B. Ryan for thorough and thoughtful comments on an earlier draft of this paper. We thank David Lazar, Becky Morton, and David A. Siegel for insightful feedback at various stages of this project. We also thank participants at the NYU-CESS 9th annual Experimental Political Science Conference, participants at the 8th annual West Coast Experiments Conference, participants at the 2015 annual meeting of the American Political Science Association, and participants at the 2014 annual meeting of the International Network of Social Network Analyses.

REFERENCES

- Bakshy, Eytan, Itamar Rosenn, Cameron Marlow, and Lada Adamic. 2012. "The Role of Social Networks in Information Diffusion." In *Proceedings* of the 21st International Conference on World Wide Web. New York: ACM, 519–28.
- Barbera, Pablo. 2014. "How Social Media Reduces Mass Political Polarization. Evidence from Germany, Spain, and the US." Presented at the annual meeting of the 2014 American Political Science Association, Washington, DC.
- Bond, Robert M., Christopher J. Fariss, Jason J. Jones, Adam D. I. Kramer, Cameron Marlow, Jaime E. Settle, and James H. Fowler. 2012. "A 61-Million-Person Experiment in Social Influence and Political Mobilization." *Nature* 489:295–98.
- Centola, Damon. 2010. "The Spread of Behavior in an Online Social Network Experiment." *Science* 329:1194–97.
- Granovetter, Mark S. 1973. "The Strength of Weak Ties." *American Journal of Sociology* 78 (6): 1360–80.
- Huckfeldt, Robert, Paul Allen Beck, Russell J. Dalton, and Jeffrey Levine. 1995. "Political Environments, Cohesive Social Groups, and the Com-

munication of Public Opinion." American Journal of Political Science 39 (4): 1025-54.

- Jerit, Jennifer, and Jason Barabas. 2012. "Partisan Perceptual Bias and the Information Environment." *Journal of Politics* 74 (3): 672–84.
- Klar, Samara. 2014. "Partisanship in a Social Setting." American Journal of Political Science 58 (3): 687–704.
- Kraft, Patrick W., Milton Lodge, and Charles S. Taber. 2015. "Why People 'Don't Trust the Evidence': Motivated Reasoning and Scientific Beliefs." *Annals of the American Academy of Political and Social Science* 685 (1): 121–33.
- Kramer, Adam D. I., Jamie E. Guillory, and Jeffrey T. Hancock. 2014. "Experimental Evidence of Massive-Scale Emotional Contagion through Social Networks." *Proceedings of the National Academy of Sciences of the United States of America (PNAS)* 111 (24): 8788–90.
- Levitan, Lindsey C., and Julie Wronski. 2014. "Social Context and Information Seeking: Examining the Effects of Network Attitudinal Composition on Engagement with Political Information." *Political Behavior* 36 (4): 793–816.
- Llewellyn, Karl N. 1931. "Legal Tradition and Social Science Method: A Realist's Critique." In *Essays on Research in the Social Sciences*. Washington, DC: Brookings Institution Press, 89–120.
- Mansbridge, Jane N. 1983. *Beyond Adversary Democracy*. Chicago: University of Chicago Press.
- Manski, Charles F. 1993. "Identification of Endogenous Social Effects: The Reflection Problem." *Review of Economic Studies* 30 (3): 531–42.
- McCubbins, Mathew, Daniel Enemark, and Nicholas Weller. 2014. "Knowledge and Networks: An Experimental Test of How Network Knowledge Affects Coordination." *Social Networks* 36:122–33.
- Mendelberg, Tali, and John Olseske. 2000. "Race and Public Deliberation." Political Communication 17:169–91.
- Mutz, Diana C. 2002. "The Consequences of Cross-Cutting Networks for Political Participation." American Journal of Political Science 46:838– 55.
- Price, Vincent, and Joseph N. Cappella. 2002. "Online Deliberation and Its Influence: The Electronic Dialogue Project in Campaign 2000." *IT and Society* 1:303–28.
- Prior, Markus. 2013. "Media and Political Polarization." Annual Review of Political Science 16:101–27.
- Sinclair, Betsy. 2012. The Social Citizen. Chicago: University of Chicago Press.
- Stimson, James. 2004. *Tides of Consent*. Cambridge: Cambridge University Press.
- Walsh, Katherine Cramer. 2004. *Talking about Politics: Informal Groups and Social Identity in American Life*. Chicago: University of Chicago Press.
- Wojcieszak, Magdalena, and Diana Mutz. 2009. "Online Groups and Political Discourse: Do Online Discussion Spaces Facilitate Exposure to Political Disagreement?" *Journal of Communication* 59:40–56.