

It's Not Just What You Have, but Who You Know: Networks, Social Proximity to Elites, and Voting in State and Local Elections

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Individual-level studies of electoral turnout and vote choice have focused largely on personal attributes as explanatory variables. We argue that scholars should also consider the social network in which individuals are embedded, which may influence voting through variation in individuals' social proximity to elites. Our analysis rests on newly discovered historical records revealing the individual votes of all electors in the 1859 statewide elections in Alexandria, Virginia and the 1874 municipal elections in Newport, Kentucky, paired with archival work identifying the social relations of the cities' populations. We also replicate our core findings using survey data from a modern municipal election. We show that individuals more socially proximate to elites turn out at a higher rate and individuals more socially proximate to a given political party's elites vote disproportionately for that party. These results suggest an overlooked social component of voting and provide a rare nineteenth-century test of modern voting theories.

Citizens' personal attributes and attitudes have long been the dominant explanations of individual electoral turnout and vote choice. An innovative and growing body of work demonstrates, however, that many individual voting decisions arise not through the individual's own attributes, but rather through social influence—when one's political behavior depends on that of their associates. Our friends, family, and coworkers help us to learn about politics (Ahn et al. 2013; Eveland and Hively 2009), influence whether we participate in politics (Klofstad 2010; McClurg 2006; Mutz 2006; Nickerson 2008; Rolfe 2012), and help us decide how to vote (Richey 2008; Ryan 2011; Sinclair 2012; Sokhey and McClurg 2012).

People are connected to one another in a large social network, but most research on social influence in voting focuses on only a few close relationships. People often form hundreds of interpersonal relationships—friends, families, coworkers, neighbors—and are connected to even more individuals through a series of intermediaries—friends' families, coworkers' neigh-

bors, and so on. Compelling theories suggest this extended social network exerts important influence on individual decisions (e.g., Siegel 2009; 2013), but measurement challenges have impeded their empirical examination. Most observational work on the subject relies on survey batteries asking respondents to identify the three to five people with whom they talk most frequently. These batteries cannot possibly include all of the relevant associates with whom a respondent interacts, nor can they measure effects of associates' friends and other people they are connected to only through intermediaries. Thus, the extended social network's influence on voting has received little attention. Though our friends and family may indeed exert the strongest social influence in our lives, they comprise only a small subset of our relationships—and we know little about whether and how these remaining relationships influence voting (Eulau 1980).

We focus on one mechanism by which these relationships may influence voting, exploring the extent to which individuals' turnout and vote choice depend on their social proximity to elites—the number of intermediaries in the social network between individuals and political elites such as candidates running for office. We rely on three datasets that allow us to measure this social proximity. The first two rest on newly-discovered historical records that reveal the individual votes of all electors in the 1859 statewide elections in Alexandria, Virginia and the 1874 local elections in Newport, Kentucky. Nineteenth century Virginia and Kentucky employed *viva voce* election law, requiring all votes to be cast by voice at assigned polling places. Every voter's preference for every office was thus recorded by election clerks. Alexandria and Newport are the only two cities under this voting law for which the complete poll books have been recovered. We have conducted archival work to pair these individual votes with detailed profiles for all known inhabitants of these mid-sized U.S. cities at the times of these elections. Our records reveal inhabitants' personal attributes such as age, wealth, place of birth and social attributes including religious affiliation, family structure, occupation,

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and place of residence. Our third dataset comes from a survey of students at the College of William & Mary prior to the 2010 municipal election in Williamsburg, Virginia. This modern election is useful because it shares with the nineteenth century datasets the ability to identify the network locations of potential voters and a candidate running for local elected office.

In each dataset, we show that people who are more proximate to elites in the social network are more likely to vote. When voting, people more proximate to elites from a particular party are more likely to support that party and less likely to support their opponents. Sensitivity analysis and a placebo test suggest these relationships are robust to confounds arising from the clustering of interests and environmental influences within the network. These results suggest that individuals' voting behavior in state and local elections depends not only on their personal attributes and attitudes, but also the vast social network in which they are embedded. This work contributes to the voting literature by specifying a mechanism through which not only individuals' close associates, but also their extended networks, influence individual voting behavior. It also contributes by providing a rare individual-level test of voting theories in an era predating scientific polling.

Social and Atomistic Explanations of Voting

At least since Paul Lazarsfeld and his colleagues from Columbia University surveyed residents of 1940 Erie County, Ohio and 1948 Elmira, New York, empirical studies of political behavior have explored social influence. The Columbia team's focus on individual communities, with respondents in close proximity to one another, emphasizes the interdependence in individual political behavior (Berelson, Lazarsfeld, and McPhee 1954; Lazarsfeld 1948). The rise of the "Michigan model," focusing on nationally-representative samples rather than individual communities, obscured this interdependence. Social influence is difficult to observe when individuals under study are treated as atomistic actors, geographically distant and socially isolated from one another. Since *The American Voter* (Campbell et al. 1960), the nationally-representative sample has become the dominant paradigm in political behavior research. As a result, our understanding of social influence in voting has developed slowly, relative to the rapid progress made understanding the personal attributes that shape political behavior.

The literature on personal attributes suggests the best predictors of turnout in U.S. national elections are socioeconomic resources such as education, wealth, and occupational status (Leighley and Nagler 1992; Verba, Scholzman, and Brady 1995; Wolfinger and Rosenstone 1980) as well as characteristics including race and church membership (for a review of this literature, see Campbell 2013). Resources also predict turnout in local elections, as do other attributes such as age and home ownership (Oliver and Ha 2007). For candidate choice, the predominant explanations focus on political attitudes such as ideology and parti-

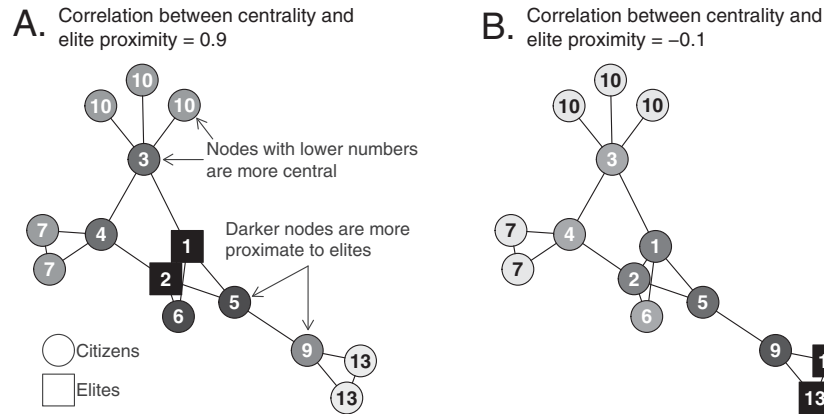
san identification (Bartels 2000; Campbell et al. 1960; Campbell, Green, and Layman 2011). Models of candidate choice also frequently employ as explanatory variables the resources and demographics used to explain turnout (e.g., Miller and Shanks 1996; Oliver and Ha 2007). In modern U.S. national elections, economic resources are strongly associated with candidate support (Gelman et al. 2009) and this relationship appears to extend back to the nineteenth century (DeCanio 2007). Citizens tend to be uninformed about local political issues, but those who show up to vote tend to be highly informed (McLeod, Scheufele, and Moy 1999). As a result, the predictors of candidate choice in high-information national elections also perform well in local elections (Oliver and Ha 2007).

Over the last few decades, we have made progress understanding social influence by returning to the community-centered designs pioneered by the Columbia team. Huckfeldt, Sprague, and colleagues reinvigorated this literature with their studies of 1984 South Bend (Huckfeldt and Sprague 1995) and 1996 Indianapolis and St. Louis (Huckfeldt, Johnson, and Sprague 2004). Their surveys include name-generator batteries, which ask respondents to name a handful of individuals with whom they discuss politics. The researchers then interview some of these discussants, allowing them to examine social influence arising between main respondents and their discussants. This methodology provides detailed measurement of a few close relationships, but also comes with a cost. By focusing attention on only a few immediate relationships, name generators obscure social influence arising between less socially-proximate relations.

Social influence is likely to extend well beyond the handful of individuals identified in a name generator. We are all connected in a vast network of relationships. You are connected to your neighbors by geography and your neighbors are connected to their coworkers by occupation. Your neighbors thus act as intermediaries, connecting you to their coworkers. Siegel (2009) develops a computational model of simulated voters to demonstrate the implications of this point. His model suggests that the influence of direct associates is inexorably intertwined with the extended pattern of relationships in the network. Whether two individuals influence each other depends not only on their dyadic relationship, but also on the pattern of relationships they each hold with others, the pattern of relationships that each of their associates hold with others, and so on. Siegel's model suggests we can improve our understanding of voting by developing theory that incorporates these extended network effects, combined with more comprehensive measurement of the social network in which individuals are embedded.

A Theory of Social-Proximity Voting

To move toward a better understanding of how networks influence turnout and candidate choice, we focus on the role of elites. Social ties to elites have long played a prominent role in theories of turnout, if not

Figure 1. Centrality does not Necessarily Imply Social Proximity to Elites

Note: The figure shows two hypothetical networks where circular nodes represent ordinary citizens, squares represent elites, and lines represent relationships between these individuals. The numbers rank the nodes by eigenvector centrality, with lower values indicating greater centrality. Darker nodes are more proximate, on average, to the elites in the network based on the number of intermediaries separating them.

candidate support. People participate when asked and these requests for participation often come from elites' mobilization efforts (Rosenstone and Hansen 1993; Verba, Schlozman, and Brady 1995). People are more likely to receive mobilization appeals when they are involved in a variety of civic organizations because these associations lead to more frequent interaction with community leaders and other elites (Verba, Schlozman, and Brady 1995). While this literature has focused on direct contact with these mobilizing elites, we should also expect influence to spread from individual to individual across the network. If you attend church with the county treasurer, she may encourage you to vote, and you may in turn encourage your neighbor to vote. Indeed, several recent studies suggest mobilization appeals influence not only recipients, but also their associates, though the effect dissipates as it spreads from person to person (Bond et al. 2012; Nickerson 2008).

Thus, we should expect the influence of elites to extend beyond initial recipients, spreading through the network. This expectation is consistent with two prominent studies—Nie, Junn, and Stehlik-Barry (1996) and Rolfe (2012)—each arguing that individuals' positions in the social network covary with their levels of personal resources, causing scholars who have focused only on the latter to overlook an important social component of voter turnout. Nie, Junn, and Stehlik-Barry (1996) argue that socioeconomic resources influence participation, in part, by strengthening resource-rich citizens' social connections with elites. Drawing from the terminology of social network analysis, Nie, Junn, and Stehlik-Barry conceive of individuals as nodes in a network, linked together by relationships formed on the basis of geography, occupation, hobbies, and a multitude of other domains. Educated individuals are more central in this network, where network centrality refers to how “well-connected” an individual is relative to others in the network. They argue,

[E]ducational attainment has a profound effect on the positions of individuals by placing them in more- or less-central network positions. Those with higher levels of formal education are substantially more likely to be found closer to the central nodes of politically important social networks, while those with less education are much more likely to be found at the periphery. Citizens who are at the center of society also end up at the center of political networks [45].

In their view, better educated people participate more frequently because education increases individuals' centrality in society at large, which subsequently increases their social proximity to elites. Better-educated individuals are more central generally because they interact with broader sets of the community; they hold occupations with greater leadership and supervisory responsibilities, accumulate greater wealth, and participate in more voluntary organizations (see also Rosenstone and Hansen 1993, 80–8). Nie, Junn, and Stehlik-Barry (1996, 45) assume this interaction with diverse parts of the network leads to greater interaction with elites, but this assumption warrants testing.

Because people tend to interact with others of similar age, class, and background (McPherson, Smith-Lovin, and Cook 2001), elites and ordinary citizens are likely to occupy different social spheres. Given the paucity of elites relative to ordinary citizens, the most-connected citizens may still be distant socially from elites. Consider the two hypothetical networks displayed in Figure 1. In part A, the elites (represented by squares) are the most central actors.¹ Consistent with

¹ Several common measures of centrality exist. Perhaps the most common measure, degree centrality, equals the number of relationships an individual has in the network. We use instead eigenvector centrality because it takes into account intermediaries, increasing as one forms relationships with more individuals and, unlike degree centrality, as the individuals one is connected to themselves form

Nie, Junn, and Stehlik-Barry's expectations, ordinary citizens (circles) who are more central in this network are also more socially proximate to elites. Consider the third most central individual (Node number 3). She is highly central because she interacts with five people and several of these people also interact with many others. She is also proximate to the elites, interacting with one (Node 1) and only a single intermediary (Node 4) separating her from the other elite (Node 2). In part B, the elites cluster in a separate part of the network. Here an individual's centrality provides little information about her proximity to elites. Node 3's centrality remains unchanged from part A, but she is now less proximate to elites because she must pass between three intermediaries (Nodes 1, 5, and 9) before reaching either elite.

If elites are not the most central actors, centrality and social proximity to elites may have distinct effects on voting. The effect of network centrality should depend on the voting behavior of the most central actors. Individual decisions on whether and how to vote tend to follow those of their social relations (Huckfeldt and Sprague 1995; Lazarsfeld 1948; Rolfe 2012; Sinclair 2012). When well-connected people abstain or when they vote Republican, they will encourage many others to abstain or vote Republican as well. Therefore, in elections where the most central actors do not vote, increasing any individual's centrality should decrease her likelihood of voting (Fowler and Smirnov 2005).

In contrast, we expect social proximity to elites to consistently encourage voting. Social proximity increases people's access to political information, reducing the costs of voting; it magnifies their voice in the political system, strengthening their sense of political efficacy; and it increases social pressure to vote, which has a powerful effect on turnout (Gerber, Green, and Larimer 2008; Panagopoulos 2010). Our first hypothesis follows from these mechanisms:

Elite Proximity-Turnout Hypothesis: Individuals more socially proximate to a city's elites should be more likely to turnout to vote in elections.

While Nie, Junn, and Stehlik-Barry (1996) and Rolfe (2012) focus only on the effect of social proximity on turnout, we argue social proximity to elites should also affect candidate support. Individuals with direct relationships with local candidates tend to feel favorably toward these candidates (Oliver and Ha 2007, 404, footnote 15). These positive feelings should spread throughout the network through informational mechanisms and social pressure. In the informational route, social proximity to strong supporters of a particular party biases the stream of information an individual is likely to receive, producing more favorable messages about the party and less favorable messages about opposing parties (Downs 1957; Huckfeldt, Pietryka, and

Reilly 2014). Likewise, social pressure coupled with the desire to avoid cognitive dissonance (Festinger 1957; 1964), should encourage conformity in voting patterns amongst associates (Huckfeldt, Johnson, and Sprague 2002; 2004; Sinclair 2012). Both of these effects should be strongest for those closest to a party's elites and dissipate with social distance and proximity to elites from opposing parties:

Elite Proximity-Support Hypothesis I: Individuals more socially proximate to elites from a focal political party should be more likely to support that party in elections.

Elite Proximity-Support Hypothesis II: Individuals more socially proximate to elites from a political party opposing the focal party should be less likely to support the focal party in elections.

RESEARCH DESIGN

To test our hypotheses, we need data identifying the elites and ordinary citizens within a community and relationships connecting these individuals. Previous work on the topic has lacked these data. Nie, Junn, and Stehlik-Barry measure social proximity to elites using survey questions from a nationally representative sample, asking respondents whether they were personally acquainted with various elected representatives and media personnel. They show that people who report more acquaintances with these elites also report voting more frequently, even after controlling for personal resources. Rolfe uses survey data to demonstrate that more central people—measured as respondents who discuss politics with more people—tend to report voting at a higher rate, after controlling for personal resources and attitudes. Nie, Junn, and Stehlik-Barry lack a measure of general centrality while Rolfe lacks elite proximity. Therefore neither study can separate the effect of centrality from that of elite proximity and, without more detailed network data, neither approach can study individuals connected to elites only through intermediaries.²

To better specify the relationships among individuals, we must draw from Huckfeldt and colleagues, focusing on small, well-defined communities. Much like the Columbia studies before them, Huckfeldt and colleagues make no claims that their samples are representative of the nation as a whole (Huckfeldt and Sprague 1995, 25). Rather, they chose to sacrifice representativeness to better observe the interdependence among their respondents. Since then, studies in the lab (e.g., Ahn, Huckfeldt, and Ryan 2014; Carlson and Settle 2016), the field (e.g., Nickerson 2008; Sinclair, McConnell, and Michelson 2013), and computer simulations (e.g., Fowler and Smirnov 2005; Rolfe 2012; Siegel 2013) have followed this model, contributing new insight into interdependence while eschewing nationally representative samples.

relationships with more individuals. Social influence studies commonly use eigenvector centrality instead of degree or alternatives such as closeness and betweenness because it imposes fewer assumptions about the paths that information travels through the network (Borgatti 2005, 62).

² Without these data, Rolfe identifies several demographic characteristics to serve as proxies for position in the network, demonstrating their impact on turnout. She also shows that areas more geographically proximate to the homes of electoral candidates tend to have higher turnout rates (Rolfe 2012, Chap. 8).

If we hope to make progress, we must accept that no single dataset is likely to provide both detailed measurement of social interaction and easily-generalizable results. Without representative samples, studies of social influence must demonstrate their generality through replication in new contexts. This replication process poses a challenge because social data are rarely gathered in the same way and the unique features of each context and measurement strategy may limit their comparability. Nonetheless, studies of social influence can bolster their generality by demonstrating that the same patterns of social influence arise in different contexts and under different measurement strategies.

With these considerations in mind, we test our theory in three settings: the 1859 state elections in Alexandria, Virginia, the 1874 municipal elections in Newport, Kentucky, and the 2010 municipal election in Williamsburg, Virginia. The central commonality between these studies is they each identify ordinary citizens, elites, and relationships between them. The differences in time, place, and measurement make comparisons between these studies difficult, but allow us to test the robustness of our results across different contexts and specifications. Together, these three settings provide insight into the social logic of voting that no single dataset could accomplish on its own.

1859 Alexandria and 1874 Newport

Alexandria, Virginia rests across the Potomac River from Washington, DC. By 1859, it was a prosperous commercial city, relying heavily on slave labor. Roughly 1,400 of its 13,000 residents were slaves, while another 1,500 were free blacks. Alexandria otherwise featured relatively little ethnic diversity, with eight percent Irish-born, two percent German-born, two percent born elsewhere in Europe, and the remainder born in the United States. Through most of the 1850s, the voters of Alexandria favored the Whigs, but by 1859 with the Civil War approaching rapidly, the Whigs had disintegrated and the Democratic Party had become increasingly dominant in the state. Virginia's 1859 statewide elections therefore pitted the Democratic Party against the newly formed "Opposition Party," comprised largely by former Whig members.

Like Alexandria, Newport also sits on a river—the Ohio—across from a larger urban center—Cincinnati. Unlike Alexandria's mercantile economy, Newport was largely industrial. While similar in size to Alexandria, with almost 16,000 residents, fewer than one percent of 1874 Newport's residents were African American. Newport was nonetheless ethnically diverse, with Irish- and German-born residents comprising about eight and 18 percent of the city's population and another six percent born elsewhere in Europe. The city was in the midst of an economic depression, sparked by a financial panic, that came after years of sustained economic growth. By the 1874 municipal elections, the Democratic and Republican Parties had become the dominant electoral contestants.

With these and myriad other considerations in mind, voters in each city filed into their assigned polling places and each called out their votes for the various offices. These votes were all recorded and we use these records to construct the outcome variables in our analysis. To compliment these poll books, we have worked with a large interdisciplinary team to assemble public records about the cities' social networks at the time of the elections. Using census rolls, tax lists, plat maps, city directories, and other sources, we have identified the familial relations, street addresses, and occupations of each known resident in Alexandria and Newport. Lastly, we have culled church membership lists to link residents to the specific churches they attended.³ In addition to these social data, the records tell us residents' age, gender, accumulated wealth, and if they own or are in the process of purchasing their home.

These nineteenth century elections are useful for understanding voting more generally because they provide unobtrusive and comprehensive measurement of all known citizens and local candidates in each city. In addition to official records listing attributes such as wealth and home ownership, these are the only data available that provide a census of who voted in an election *and* for whom each citizen voted. Only with unbiased measures of candidate choice can we test Elite Proximity-Support Hypotheses I and II. These data provide an opportunity to define the network using the same variables—family, neighbors, occupation, and church—that best predict relationships in modern political discussion networks (Huckfeldt et al. 1995, 1032). More generally, these measures predict many forms of human interaction across a range of regions and eras (for a review of this literature, see McPherson, Smith-Lovin, and Cook 2001) including nineteenth century U.S. cities (Tilly 2007, 80). We have these measures for the entire city populations, providing a more detailed and comprehensive image of the cities' social networks than can be afforded with most survey samples.

Despite the advantages of these data, nineteenth century elections differ in many ways from those of today. Dramatic changes in social and political life have occurred since the nineteenth century, limiting their comparability to modern elections. These changes raise a question: are theories developed through the use of modern survey data relevant for understanding voting in this earlier period (Shortridge 1980, 617)? We are aware of no individual-level analyses of turnout and few of vote choice that test modern theories of voting in the nineteenth century.⁴ Thus, a secondary contribution of our analysis is to study the extent

³ We have a complete census of all 13 churches known to exist in Alexandria at the time of the elections, but our data includes only six of the 15 known churches in Newport.

⁴ For earlier individual-level studies of nineteenth-century vote choice using poll books, see Bohmer (1977; 1978), Bourke and DeBats (1977; 1980; 1985; 1987), DeBats (2009), and Rozett (1977). For work using nineteenth-century county directory data to study the relationship between individuals' resources and self-reported partisanship, see Hammarberg (1977) and DeCanio (2007). For earlier individual-level work examining spatial, rather than social, distance in shaping nineteenth-century voting behavior and partisan preferences, see Bourke and DeBats (1995), DeBats (2004; 2008; 2011)

to which resource-based models help us understand nineteenth-century voting.

We would also like to test our theory in a modern election, but the archival methods we use to collect the nineteenth century data cannot be replicated for recent elections. Without viva voce poll books, no archival work can provide a measure of vote choice. One could instead document public demonstrations of support such as displaying yard signs (e.g., Makse and Sokhey 2012) and identify voters using state voter files. This approach will not disclose the identities of individuals who do not display a sign or register to vote, posing a significant selection problem for an individual-level study. One could pair these data with U.S. census records to enumerate an entire city population, but this path is precluded because individual census records remain closed for 72 years. The data we use to identify relationships between individuals would also be difficult to obtain. We rely on the census for occupational data, city tax records for neighbors, and a combination of the two for family. The city tax records are a historical artifact; nineteenth century cities taxed personal as well as real property, encouraging tax collectors to visit every household and enumerate all residents therein. Modern tax and property records may reveal the owner of a parcel, but provide no information about who resides there—impeding measures of family or neighbors. We identify religious membership from the archival records of the religious institutions, including member lists, baptisms, marriages, and deaths which have been deposited in public archives or which the institutions were willing to make available as historical documents. Religious institutions are unlikely to share such information about recent members.

For these reasons, we instead test our theory in a modern election using a survey of college students. College records overcome the challenge of enumerating all community members while a survey-based name generator, combined with housing records, map relationships between these individuals. These data provide a more detailed map of the network than can be obtained through either modern archival work or larger-community studies including Huckfeldt and Sprague's innovative survey designs. To test our theory, we also require elites in the network, but few college students run for elected office. We leverage a rare exception in which a student ran for—and won—a city council seat.

2010 Williamsburg

The May 2010 Williamsburg municipal election occurred in the wake of several recent city council decisions affecting student life at the College of William & Mary, a liberal arts college located in the city. In particular, the council recently adopted a strict noise ordinance and banned more than three unrelated individuals from living together in a single dwelling. A central decision for voters in this election was whom

to elect for city council. One of these candidates, Scott Foster, was concurrently a William & Mary student.

We rely on a survey fielded by Daniel Maliniak, Patrick Miller, and Ronald Rapoport during the election targeting all current William & Mary students (Miller et al. 2015). Most of our data are drawn from the preelection wave, which received responses from 2,740 students, representing just under 50% of the student body. We also draw some data from a postelection wave which received 992 valid responses constituting a sixth of the student body.⁵ The social network we explore contains all students at the school, including Scott Foster, allowing us to measure each student's social proximity to this elite. We use two methods to identify relationships between students. First, the preelection survey asked respondents to identify their five closest friends at William & Mary. Second, we obtained housing records for all 3,655 students living on campus, allowing us to identify relationships between roommates, even if they did not respond to the survey. These housing data add substantially to the network because 64% of students live on campus.

The Williamsburg data address several limitations of the nineteenth century studies. First, they allow us to test our theory in a modern municipal election because they provide measures of students, an elite, and the relationships connecting them. Second, by using a name generator and roommate records to identify the network, we obtain more concrete measures of relationships than those afforded by archival work—relationships are explicitly identified rather than imputed based on shared attributes. By relying on name generators, we test the robustness of our results using the typical measurement strategy of recent research on social influence in political behavior. Third, the survey measures attitudes including partisan identification and behaviors such as participating in previous elections, providing controls for established predictors of voting unavailable through archival work.

Limitations

Our three datasets share an ability to measure the attributes of everyday citizens, elites, and relationships connecting these groups. Like previous studies of social influence, the features of the data that make this measurement possible also limit their generality. The elites in all three datasets are candidates for local offices and elite proximity may have different effects in national elections, which feature broader, less-connected

and DeBats and Lethbridge (2005). For an assessment of the continuing divergence between the explanatory power of individual and aggregate political data, see Bourke, DeBats, and Phelan (2001).

⁵ The survey was administered online and all 5,726 students were invited via email to participate. Students who did not respond within the first two weeks were recontacted with an abridged version of the survey, including only key demographic items and the name generator. The AAPOR participation rate for nonprobability internet panels is equal to the number of usable responses divided by the number of people invited to participate (The American Association for Public Opinion Research 2015, 40). We define usable responses as the respondents who affirmed in the survey that they were 18 years or older and had at least one direct connection in the social network, yielding a preelection participation rate of $\frac{2590}{5726} = 0.45 = 45\%$ and a postelection participation rate of $\frac{992}{5726} = 0.17 = 17\%$.

constituencies and larger mobilization efforts. Yet local elections are of great significance; due to the vast number of local governments, local elections occur in much greater number than national elections and locally-elected officials are responsible for implementing policy and allocating a large proportion of the nation's resources (Trounstein 2009).

The college-age citizens in the Williamsburg data may be more susceptible to social influence than older citizens. The nineteenth century data overcome this problem, but the public nature of the vote may also amplify social influence. Worse, it may have facilitated patronage—the partisan use of public office to appoint individuals dedicated to the welfare of a political machine. If present in our data, patronage may increase elites' social influence relative to modern elections. But patronage systems typically arose in large cities which provided governing parties with many job openings to exchange for party loyalty. Nineteenth century Alexandria and Newport were too small and provided too few services to generate many such jobs. We have nonetheless searched for evidence of patronage in these cities, finding evidence that it occurred, but only rarely. The public vote also opens the possibility of vote buying and coercion, but like patronage we believe these behaviors were rare in these elections. See Online Appendix A for the evidence on which we base these conclusions.

The many differences between the nineteenth century data and the Williamsburg data limit our ability to draw connections between them. For instance, the nineteenth century networks feature many elites from different parties while the Williamsburg network features only a single elite. Citizens embedded in these different networks are likely to encounter different information and face different forms of social pressure. The network measures also differ with these two approaches. We assume in the nineteenth century networks that people from the same family, occupation, church, or block interact with one another. This assumption is often correct (Huckfeldt 1983), but produces only a probabilistic measure of the network.⁶ This network undoubtedly overlooks some important relationships while also imputing connections between people who do not know each other directly. The Williamsburg data avoid this problem, providing more certain measures of relationships. By focusing on close friends and roommates, these relationships are likely to be stronger and more personal than those captured in the nineteenth century data. The many differences between these studies provide a strong test of the robustness of our results, but they prevent us from identifying the specific mechanisms that generate these broad patterns. As we proceed, we therefore pay careful attention to the confounds that may lead us to ascribe social influence where none exists.

⁶ In this manner, these measures are analogous to aggregated relational data, which are obtained through survey items that probabilistically approximate the social distance between survey respondents and various subgroups of interest (Killworth et al. 1998; McCormick et al. 2013).

Perhaps the largest set of confounds we face are not limited only to our design, but pervade virtually all cross-sectional studies of social influence (Fowler et al. 2011). Regardless of how the network is measured, people who are socially proximate to one another inhabit similar environments and share similar interests. In our study, shared environments can confound our conclusions when people who are socially proximate to each other encounter the same external pressures to vote. A pair of neighbors or college roommates may each turn out to vote, not because one influenced the other, but because they were each canvassed by the same get-out-the-vote drive. Shared interests arise because people interact primarily with others who are similar to themselves (McPherson, Smith-Lovin, and Cook 2001). Those shared interests can confound our estimates if they lead socially proximate individuals to vote similarly to one another. In the nineteenth century data, people who attend the same church as an elite may share the elite's economic interests, beliefs, and values. These shared traits may also influence their political behavior. In the Williamsburg data, people may be friends because of their shared interest in politics and hence the similarity in their voting may be a cause of the friendship rather than the consequence. Our data allow us to control for some of these shared interests, but we cannot eliminate these confounds. We therefore subject our estimates to sensitivity analysis, originally developed by VanderWeele (2011), in order to assess the robustness of apparent social influence effects to these and other unobserved confounds.

Constructing the Nineteenth Century Networks

Our goal is to measure the social proximity of an individual, i , to a city's elites. We also must measure separately the social proximity of an individual to each party's elites. We define the social distance between individual i and a set of elites [to be denoted $\text{Social Distance to Elites}_i$] as the average—across all members of the elite set—distance between individual i and elite j .

The distance between individual i and elite j , d_{ij} , is measured by first determining the social distance between every pair of individuals in the city. To measure this distance between any pair, we assume two individuals are more closely connected the more direct social connections between them. We therefore count the number of direct social connections between each pair using each of four possible means that can be observed in our data:

- whether they are in the same family (i.e., they share the same surname and live in the same residence),
- whether they are neighbors (i.e., they live on the same block and the same side of the street),
- whether they are confreres (i.e., they share an occupation), and
- whether they are affiliated with the same church.

We also allow for paths passing through intermediaries; if Tom and Dick attend the same church and Dick and Harry have the same occupation, then Tom and Harry are connected on a path passing through Dick. With these assumptions, we apply Dijkstra's shortest path algorithm, which finds the shortest path between any two individuals (for details, see Cormen et al. 2009, 658). The resulting social distance measure holds larger values for pairs of individuals with more intermediaries between them and fewer direct social relationships along the path. To arrive at d_{ij} , we then extract the distance between each individual i and elite j . Individuals not connected to an elite by any path length are coded as one plus the maximum path length to this elite.

In Alexandria, we define elites as the candidates who ran for city-wide office in the local elections held March 1, 1859.⁷ These elections were held just three months before the May 29 statewide elections on which we base our outcome measures of turnout and vote choice. In total, 19 Opposition Party candidates, seven Democrats, and three candidates with no known party affiliation ran for these positions, including multiple candidates of the same party competing against each other for several positions.⁸ In Newport, we define elites as the candidates running that year for one of the 11 city-wide elected offices⁹—the same elections on which we base our outcome measures of turnout and vote choice. Republicans fielded a candidate for each office, but Democrats did not field nominees for City Treasurer or City Engineer. We could not determine which of the six people named James Smith in our records was the Democratic candidate for Weights and Measures. We are hence left with eight Democratic and 11 Republican elites on whom we base our measures of elite proximity.

Tables B1 and B2 in Online Appendix B provide descriptive statistics for the social distance measures and all other variables used in the analyses. In Alexandria, the average social distance between an eligible voter and each candidate, regardless of party, is 9.9 (standard deviation = 2.15). Eligible voters average the same distance to each party as well, with an average distance of 9.9 (SD = 2.14) to each Opposition Party candidate and 9.8 to each Democratic candidate (SD = 2.25). In Newport, eligible voters average a social distance of

11.0 (SD = 1.99) to each elite, 11.2 to each Democratic candidate (SD = 1.98), and 10.8 to each Republican candidate (SD = 1.99).¹⁰ For all analyses, we standardize these measures with mean = 0 and SD = 1. To aid in presentation, we then multiply these distance measures by -1 so larger values indicate greater proximity to elites.

Constructing the Williamsburg 2010 Network

To measure Social Distance to the Elite $_i$ in Williamsburg, we rely on the name generator from the preelection survey, which asks respondents to identify “the first and last names of up to five of your closest friends who attend William & Mary.” There is a direct connection from individual i to individual j , if i named j as one of her five closest friends. Using the on-campus housing records, we also code a pair of individuals as connected if they share a room. We restrict our analysis to the 2,590 respondents who had at least one direct connection in this network.

Social Distance to the Elite $_i$ equals $1 +$ the number of intermediaries on the directed shortest path running from individual i to Foster, the William & Mary student running for city council. For the 85 students who cannot reach Foster on a path of any length, their social distance measure equals $1 +$ the maximum distance in the sample (maximum = 11).¹¹ Again, this approach yields a measure of distance between pairs of individuals where higher values indicate more intermediaries between the pair. Respondents were on average connected to Foster on a path length of 6.2 (SD = 1.8).¹² As with the nineteenth century data, we standardize this elite distance measure with mean = 0 and SD = 1 and then multiply by -1 so larger values indicate greater proximity to the elite.

EMPIRICAL RESULTS

Using these measures, we first explore social proximity's relationship with turnout and vote choice in the nineteenth century. We then examine the 2010 election.

Turnout in the Nineteenth Century

Table 1 displays individual-level logistic regressions modeling turnout in Alexandria and Newport. For each city, model 1 includes only personal attributes as explanatory variables. Model 2 introduces as an explanatory variable an individual's weighted eigenvector network centrality. Model 3 introduces our measures of

⁷ These positions were Mayor, Auditor, Attorney, Superintendent of Gas, Superintendent of Police, Chief Engineer of the Fire Department, Clerk of the Market, Measurer of Wood and Bark, Gauger, Surveyor, Measurers of Lumber (two positions), and Assessors (two positions).

⁸ Since the states were not in control of a ballot, we identify party affiliation using either the party ticket or newspaper advertisements which listed the party affiliation of candidates. In the few cases where candidates' party affiliations were not listed in either source, we checked the votes of those candidates in the May state-wide elections. We found no cases in which a candidate voted in the May elections against their party, as identified with the party ticket or newspaper ad.

⁹ These positions were President of the Council, City Clerk, City Treasurer, City Attorney, City Jailor, City Engineer, City Physician, Market Master, Wharf Master, Weights and Measures, and Street Commissioner.

¹⁰ These distances are inflated only slightly by individuals who are disconnected from all elites. In Alexandria, omitting the 43 disconnected individuals produces a mean elite distance of 9.7, Opposition distance of 9.7, and Democratic distance of 9.6. In Newport, omitting the 32 disconnected individuals reduces the mean elite distance to 10.9, Democratic distance to 11.1, and Republican distance to 10.7.

¹¹ We have replicated these analyses instead replacing students who cannot reach the candidate on a path of any length with the average distance to the candidate, yielding similar results.

¹² After removing the 85 students who are not connected to the candidate, the mean decreases from 6.2 to 6.0.

Table 1. Estimates from a Model Predicting an Individual's Probability of Voting Using his Social Proximity to Elites and Other Variables (Based on the Statewide Elections in 1859 Alexandria, VA and the Local Elections in 1874 Newport, KY)

	Alexandria			Newport		
	(1)	(2)	(3)	(1)	(2)	(3)
Household wealth (thousands of dollars)	−0.011* (0.004)	−0.010* (0.004)	−0.010* (0.004)	−0.001 (0.003)	−0.001 (0.003)	−0.002 (0.003)
ln(household wealth)	0.063* (0.021)	0.041 (0.022)	0.043* (0.022)	−0.072* (0.015)	−0.066* (0.015)	−0.067* (0.015)
Mid-status occupation	0.687* (0.103)	0.417* (0.119)	0.290* (0.123)	0.145 (0.082)	0.333* (0.100)	0.191 (0.104)
High-status occupation	0.864* (0.186)	0.633* (0.193)	0.587* (0.194)	0.247 (0.189)	0.435* (0.198)	0.404* (0.200)
Owens home?	0.765* (0.150)	0.802* (0.151)	0.781* (0.152)	1.641* (0.097)	1.652* (0.098)	1.578* (0.099)
Age (years)	−0.091* (0.020)	−0.097* (0.020)	−0.099* (0.021)	−0.004 (0.015)	0.000 (0.015)	−0.007 (0.015)
ln(age)	4.378* (0.808)	4.615* (0.810)	4.738* (0.817)	−0.141 (0.524)	−0.342 (0.529)	0.010 (0.537)
Is church member?	0.881* (0.106)	0.879* (0.107)	0.582* (0.125)	—	—	—
Is African American?	—	—	—	1.440* (0.425)	1.315* (0.428)	1.298* (0.427)
Is U.S. born?	0.723* (0.119)	0.606* (0.123)	0.543* (0.124)	0.129 (0.086)	0.171 (0.087)	0.211* (0.088)
Network centrality (Z score)	—	−0.162* (0.038)	−0.235* (0.041)	—	0.096* (0.029)	−0.008 (0.035)
Social proximity to elites (Z score)	—	—	0.557* (0.125)	—	—	0.717* (0.136)
Intercept	−13.500* (2.146)	−13.806* (2.148)	−14.244* (2.167)	−0.564 (1.322)	−0.213 (1.331)	−1.366 (1.357)
N	2216	2216	2216	3416	3416	3416
Log likelihood	−1338.754	−1329.364	−1317.761	−2121.453	−2115.955	−2098.883
AIC	2697.508	2680.728	2659.522	4262.907	4253.910	4221.766

Notes: Reported coefficients are from logistic regressions (with standard errors in parentheses). Models are restricted to eligible voters (in Alexandria, white males at least 21 years of age; in Newport, males at least 21 years of age). The omitted reference category for occupational status includes low-status occupations, the unemployed, and those lacking occupational data. In each city, a likelihood ratio test suggests that model 3—which includes the social network measures—provides a better fit than does model 1 (in Alexandria, χ^2 (DF = 2) = 42; $p < 0.001$. In Newport, χ^2 (DF = 2) = 45.1; $p < 0.001$).

* $p < 0.05$.

elite proximity. The models are restricted to eligible voters.

The models control for several resource-based covariates, including wealth, occupational status, and home ownership. In Alexandria, where we have complete church records, we control for whether someone was on a church membership list (Campbell 2013, 38). Recall that we use church attendance as a form of connection in the social network. We include this control to defend against the possibility that the relationships exhibited by the elite proximity measures are spurious, emerging instead due to differences between church members and nonmembers.¹³ In Newport, where blacks had the right to vote, we control for race. In both cities, we also control for age. We expect wealth and age to have diminishing returns (see Rolfe

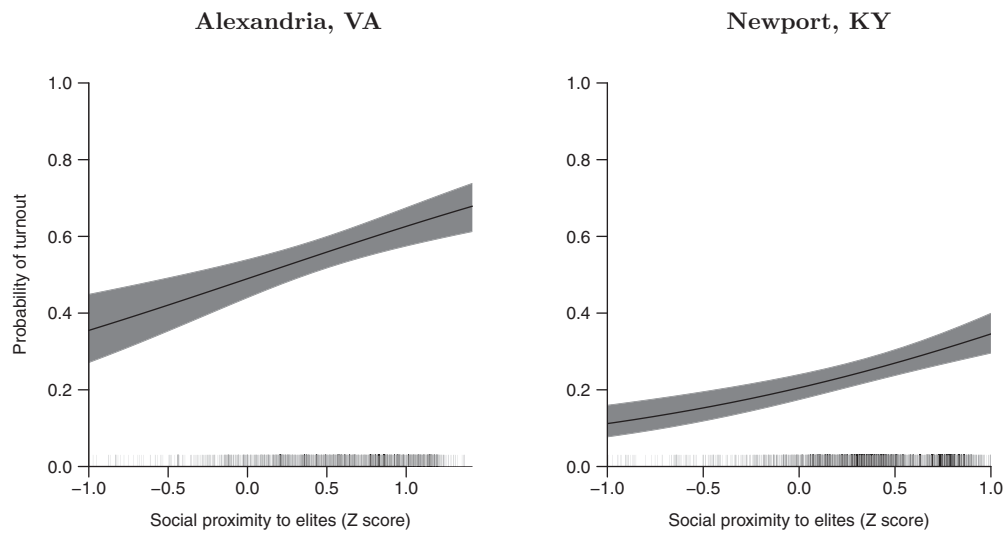
2012, 140) and therefore also include the natural log of both variables (we set the minimum wealth to \$1 before this transformation).

Predictors of turnout commonly used on modern data receive mixed support in these elections. Socioeconomic resources perform as expected in Alexandria. The models suggest, all else equal, people who have greater wealth, have higher-status occupations, and own homes are all more likely to vote.¹⁴ In Newport, greater resources do not consistently distinguish voters

¹³ Tables C1 and C2 in Online Appendix C provide models with additional controls for religion and church membership.

¹⁴ For instance, the results from Alexandria model 3 suggest that someone with only a dollar of household wealth (the minimum) will have a 0.55 probability of voting compared to 0.60 for people with \$1,000 of wealth, an increase of 0.05 (first difference = $0.60 - 0.55 = 0.05$; 95% confidence interval [0.01, 0.09]). People with \$2,000 of wealth, however, are no more likely to vote than those with \$1,000. These and all reported probabilities and confidence intervals are calculated using simulations from the posterior distribution while setting other explanatory variables to their medians.

Figure 2. An Individual's Predicted Probability of Voting Given his Social Proximity to Elites (Based on the Statewide Elections in 1859 Alexandria, VA and the Local Elections in 1874 Newport, KY)



Notes: The lines indicate the predicted probabilities and the shaded regions show the 95% confidence intervals. All other covariates are set to their medians. Rug plots show the distribution of social proximity to elites, jittered to better show frequencies. Source: Table 1, model 3.

from nonvoters. All else equal, wealthier individuals turnout at *lower* rates and the difference in turnout between people with mid- and low-status occupations is not statistically significant. People with high-status occupations do vote more frequently than people with low- or mid-status occupations and home owners vote more frequently than nonowners. Aside from these resources, the other covariates perform as expected, with voters in each city disproportionately comprised of older citizens, the U.S. born, and (in Alexandria) church members. In Newport's 1874 elections—only a few years after the Fifteenth Amendment extended the franchise to African Americans—blacks were significantly more likely to vote than were whites, all else equal.

Our Elite Proximity-Turnout Hypothesis receives support in both cities. In each, model 3 shows a positive and statistically significant coefficient associated with elite proximity, suggesting people more proximate to elites vote at higher rates. Figure 2 plots individuals' predicted probabilities of turnout by their social proximity to elites. In Alexandria, the predicted probability of turnout is 0.36 for people one standard deviation below the mean in elite proximity, 0.49 for people at the mean (first difference = $0.49 - 0.36 = 0.13$; 95% confidence interval [0.09, 0.18]), and 0.63 for people a standard deviation above (FD = $0.63 - 0.49 = 0.14$; 95% CI [0.09, 0.18]). In Newport, the corresponding probabilities are 0.11, 0.21, and 0.35 (and again the differences in these probabilities are each statistically significant). These differences are similar in magnitude to the differences in the turnout between people with high- and mid-status occupations. In Alexandria, people with high-status occupations have a 0.13 greater

probability of turnout than otherwise similar people with mid-status occupations (95% CI [0.06, 0.19]). In Newport, this difference is 0.09 (95% CI [0.02, 0.16]).

These results also help disentangle the role of general centrality from that of proximity to elites. While both Nie, Junn, and Stehlik-Barry (1996) and Rolfe (2012) assume a strong association between these two measures, they are only weakly related in our data; Pearson's $r = 0.21$ in Alexandria and 0.23 in Newport. In Alexandria, the Table 1 models 2 and 3 coefficients associated with network centrality are negative and statistically significant; all else equal, more central citizens were less likely to vote than those more disconnected from the community. In model 3, both centrality and social proximity exhibit a distinct relationship with turnout, but in opposite directions, and thus one cannot serve as a proxy for the other. We should not expect centrality to discourage participation in all elections, however. The computational models of Fowler and Smirnov (2005) as well as Siegel (2009; 2013) suggest the effect of centrality will depend on the overall levels of participation in the election.

Vote Choice in the Nineteenth Century

While social proximity to elites has a strong relationship with turnout, how does it relate to voters' candidate choices? Table 2 models vote choice as a function of the same explanatory variables used to predict turnout.¹⁵ In Alexandria, the outcome variable is equal to the proportion of votes an individual

¹⁵ We observe candidate preferences only for those who cast a vote, creating concern that our vote choice estimates may be biased

Table 2. Estimates from a Model Predicting the Proportion of an Individual's Votes Cast for the Majority Party in the City (Based on the Statewide Elections in 1859 Alexandria, VA and the Local Elections in 1874 Newport, KY)

	Alexandria			Newport		
	(1)	(2)	(3)	(1)	(2)	(3)
Household wealth (thousands of dollars)	0.002 (0.001)	0.002 (0.001)	0.002 (0.001)	-0.003 (0.002)	-0.002 (0.002)	-0.003 (0.002)
ln(household wealth)	0.001 (0.006)	-0.001 (0.006)	-0.004 (0.006)	-0.001 (0.004)	0.003 (0.004)	0.002 (0.004)
Mid-status occupation	0.042 (0.031)	0.008 (0.034)	0.010 (0.035)	-0.075* (0.022)	0.059* (0.027)	0.041 (0.027)
High-status occupation	0.028 (0.045)	-0.002 (0.047)	0.002 (0.047)	-0.035 (0.050)	0.093 (0.051)	0.085 (0.050)
Owns home?	0.042 (0.036)	0.041 (0.036)	0.048 (0.036)	-0.055* (0.027)	-0.048 (0.027)	-0.047 (0.027)
Age (years)	0.003 (0.006)	0.002 (0.006)	0.003 (0.006)	0.005 (0.004)	0.007 (0.004)	0.008 (0.004)
ln(age)	-0.106 (0.243)	-0.070 (0.243)	-0.095 (0.241)	-0.178 (0.158)	-0.270 (0.154)	-0.278 (0.153)
Is church member?	0.062* (0.027)	0.063* (0.027)	0.094* (0.034)	—	—	—
Is African American?	—	—	—	-0.275* (0.092)	-0.369* (0.090)	-0.370* (0.090)
Is U.S. born?	0.224* (0.036)	0.204* (0.037)	0.198* (0.038)	-0.059* (0.024)	-0.029 (0.023)	-0.030 (0.023)
Network centrality (Z score)	—	-0.029* (0.013)	-0.008 (0.014)	—	0.065* (0.008)	0.056* (0.009)
Social proximity to majority party elites (Z score)	—	—	0.225* (0.072)	—	—	0.297* (0.067)
Social proximity to minority party elites (Z score)	—	—	-0.304* (0.072)	—	—	-0.204* (0.070)
Intercept	0.610 (0.651)	0.556 (0.650)	0.655 (0.647)	1.040* (0.402)	1.144* (0.392)	1.134* (0.391)
N	1128	1128	1128	1381	1381	1381
R ²	0.049	0.054	0.069	0.030	0.078	0.096
Adj. R ²	0.042	0.045	0.059	0.023	0.072	0.088
RMSE	0.432	0.431	0.428	0.368	0.358	0.355

Notes: Reported coefficients are from ordinary least squares regressions (with standard errors in parentheses). Models are restricted to voters. The outcome variable is equal to the proportion of votes cast for the majority party in the city (The Opposition Party in Alexandria; The Democratic Party in Newport). The omitted reference category for occupational status includes low-status occupations, the unemployed, and those lacking occupational data. In each city, an *F* test suggests that model 3—which includes the social network measures—provides a better fit than does model 1 (In Alexandria, $F(DF = 3, 1115) = 7.7$; $p < 0.001$. In Newport, $F(DF = 3, 1368) = 33.2$; $p < 0.001$).

* $p < 0.05$.

cast for the Opposition Party across the five races for national or state government: U.S. Congress, VA House of Delegates, VA Governor, VA Lieutenant Governor, and VA Attorney General. In Newport, the outcome variable is equal to the proportion of votes cast for the Democratic Party across the nine citywide positions contested by both parties.¹⁶ We use OLS

(Heckman 1979). We therefore present selection models addressing the same question in Table D1 of Online Appendix D. The substantive results are unchanged from those presented here.

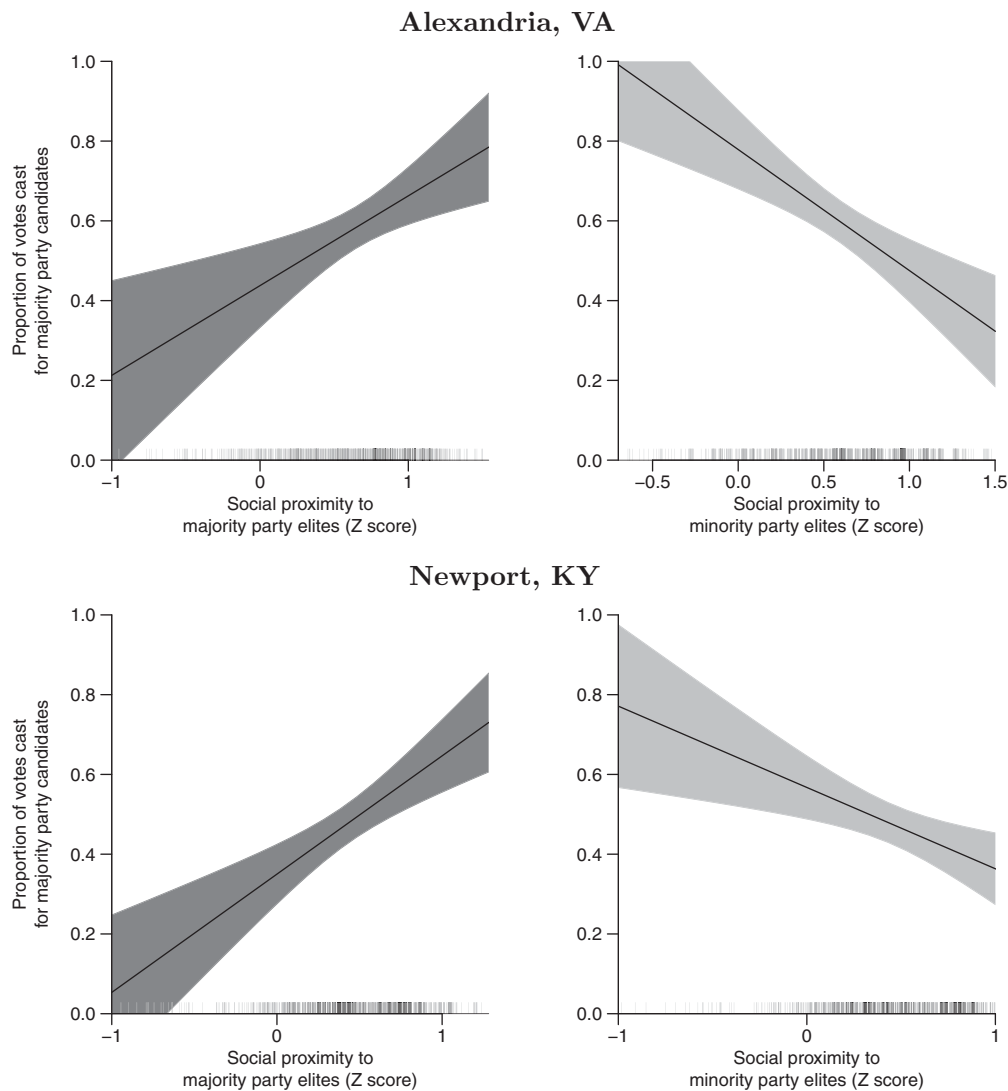
¹⁶ In both cities, we choose the party receiving the most votes as the focal party. The estimates change slightly when the other party is used as the focal party due to a few races that feature independents, but the substantive and statistical significance of the results remain unchanged.

for these models.¹⁷ As above, the first model for each city includes only personal attributes as explanatory variables, the second adds the centrality measure, and the third adds the social proximity measures.

Both elections were closely divided, with Opposition candidates securing 59% of votes in Alexandria and Democratic candidates securing 52% in Newport. Table 2 shows that, unlike modern elections (Gelman et al. 2009), resource-based measures were poor predictors of vote choice in these elections. Unsurprisingly in the wake of the Civil War, model 3 in Newport suggests that, all else at its median, African Americans were 37 percentage points less supportive of Democratic

¹⁷ Table E1 in Online Appendix E replicates this analysis with ordered logistic regressions, yielding the same substantive conclusions.

Figure 3. The Predicted Proportion of an Individual's Votes Cast for the Majority Party in the City Given his Social Proximity to Each Party's Elites (Based on the Statewide Elections in 1859 Alexandria, VA and the Local Elections in 1874 Newport, KY)



Notes: The lines indicate the model predictions and the shaded regions show the 95% confidence intervals. All other covariates are set to their medians. Rug plots show the distribution of the variable on the X axis, jittered to better show frequencies.

Source: Table 2, model 3.

candidates than were people of other races (95% CI $[-55, -19]$).

In model 3 from each city, the positive and statistically significant coefficient associated with proximity to the majority party provides support for Elite Proximity-Support Hypothesis I. Likewise, the negative and statistically significant coefficient associated with proximity to the minority party provides support for Elite Proximity-Support Hypothesis II. Figure 3 plots the predictions from the models as the social proximity measures vary. The left panel of the figure shows the proportion of votes an individual is predicted to give to the majority party in the city as social proximity to this focal party increases. The right panel shows how

this support decreases with increases in social proximity to the minority party. The figure shows that, all else equal, citizens closer socially to majority-party candidates cast a greater proportion of their votes for that party. Compared to people at the mean social proximity to the majority party, those one standard deviation above are predicted to have 0.23 greater support for the majority party in Alexandria (95% CI $[0.09, 0.37]$) and 0.30 greater support in Newport (95% CI $[0.16, 0.44]$). The right panel shows that voters closer to minority-party candidates are decreasingly likely to vote for the majority party. Compared to people at the mean social proximity to the minority party, those one standard deviation above are predicted to have 0.30 less

support for the majority party in Alexandria (95% CI $[-0.44, -0.07]$) and 0.20 less support in Newport (95% CI $[-0.34, -0.06]$).

If social proximity encourages candidate support, local party organizations can profit by recruiting candidates with strong social connections to their communities. Alternatively, our results may arise if candidates in our elections ran *because* of their social proximity to their supporters (or because of unobserved factors that covary with social proximity). Our cross-sectional data cannot provide a conclusive test of these mechanisms. Though candidates may have emerged due to their social proximity to typical party supporters, we see little evidence that candidates emerged because they were socially proximate to citizens generally. If social proximity predicts candidate emergence, we should expect candidates to be particularly central in the network, which they were not. Candidates had an average weighted eigenvector centrality of 0.04 in Alexandria and 0.13 in Newport, slightly less than the 0.17 and 0.14 averages for eligible voters in each city. If elites were exceptionally central, the distinction between social proximity to elites and general centrality would be muted; the most central citizens would also be the most proximate to elites.

Turnout and Participation in 2010 Williamsburg

We next examine the 2010 Williamsburg election, where we model turnout and political participation in support of Foster's campaign as functions of social proximity to Foster. We focus on campaign participation rather than vote choice because we lack variation in respondents' vote choices; of the respondents casting a ballot, 96% report voting for Foster. Participation provides a useful alternative because it allows us to study other forms of candidate support and campaign engagement that are not available in the nineteenth century data.

To measure turnout, we obtained the Williamsburg voter file for the election. Our outcome variable equals one if the respondent's name was in the file and zero otherwise. By using this validated voting measure, we avoid the problem of overreporting turnout that is common to public opinion surveys (Clausen 1968; Holbrook and Krosnick 2010). In addition to the social proximity measure discussed above, we control for respondents' family income, race/ethnicity, gender, and place of residence. At the time of the election, Foster was a senior government major and hence his close social circle was likely to be older and more politically interested. We therefore control for academic year, past political participation (including voting in 2008 and an additive index of political participation in 2008 and the 2009 VA gubernatorial race),¹⁸ and various

political attitudes (interest in national politics, trust in government, and partisan extremity).¹⁹ All controls are measured using the pre-election wave of the survey. Online Appendix B discusses measurement details and Table B3 displays summary statistics for these variables. King et al. (2001) argue that missing data pose a large problem for survey research, asserting that multiple imputation of missing values provides an improvement over listwise deletion. Our models, displayed in Table 3, therefore rely on multiple imputation via the Amelia package in R (Honaker et al. 2011). For details on the imputations, see Online Appendix F.²⁰

The first three models of the table examine turnout. In model 3, the coefficient associated with social proximity to the elite is positive and statistically significant, lending further support for our Elite Proximity-Turnout Hypothesis. The top-left panel of Figure 4 graphs this relationship. Consider individuals with five intermediaries between themselves and Foster, roughly corresponding to the mean social distance (social proximity to the elite Z score ≈ 0). With other variables at their medians, the model predicts they will have a 0.31 probability of voting. If these individuals made friends with someone separated from Foster by only two intermediaries, they would then be separated by only three intermediaries rather than five—roughly a standard deviation increase in social proximity (social proximity to the elite Z score ≈ 1). In this scenario, their predicted probability of turnout would increase to 0.41 ($FD = 0.41 - 0.31 = 0.10$; 95% CI $[0.06, 0.13]$). For comparison, the top-right panel graphs the relationship between previous campaign participation and turnout—one of the strongest turnout predictors (Brody and Sniderman 1977; Gerber, Green, and Shachar 2003). Compared to an individual who completed the mean number of campaign activities in 2008 and 2009 (mean = 1.6), someone who completed a standard deviation more activities ($SD = 2.2$) is predicted to have a 0.03 greater probability of voting ($FD = 0.35 - 0.32 = 0.03$; 95% CI $[0.01, 0.06]$). We are not arguing that the strength of the social proximity relationship exceeds that of previous participation; voting in the 2008 general election, a different measure of

an on-line forum or message board, writing a letter to a newspaper/magazine, or "other." These acts provide no age restrictions and thus were not limited to respondents who were voting age in these earlier elections. In contrast, some freshmen and sophomores may have wished to vote in 2008, but were not yet old enough and are thus coded as nonvoters on those measures.

¹⁹ Some of these controls such as trust in government may themselves be influenced by proximity to the elite and thus the estimated coefficient for proximity to the elite may be artificially deflated. We believe including these measures is preferable to omitting them because of the threats of shared environments and interests discussed above. In practice, omitting these controls does not alter the magnitude of the social proximity coefficient or its *p* value.

²⁰ We have also run these models omitting respondents who were not registered to vote in Williamsburg. The substantive results are unchanged with this specification. We present the models that include respondents who were not registered because registration in Williamsburg may itself be influenced by social proximity to the elite.

¹⁸ For each election, these participatory acts include the following: attempting to persuade someone to vote for a particular candidate, working on a campaign, attending meetings or rallies, displaying a political yard sign, bumper sticker, or wearing a campaign button, giving money to a candidate, writing on a blog or participating in

Table 3. Estimates from Models Predicting an Individual's Probability of Electoral Turnout and Number of Campaign Activities Completed in Support of Foster (Based on the Local Election in 2010 Williamsburg, VA)

	Turnout			Number of Campaign Activities		
	(1)	(2)	(3)	(1)	(2)	(3)
Family income	0.048 (0.062)	0.045 (0.062)	0.037 (0.062)	0.116 (0.064)	0.119 (0.064)	0.131* (0.061)
Is Caucasian, non-Hispanic?	0.518* (0.145)	0.522* (0.146)	0.438* (0.147)	0.241 (0.142)	0.242 (0.142)	0.175 (0.137)
Is female?	-0.183 (0.105)	-0.171 (0.106)	-0.165 (0.107)	-0.163 (0.105)	-0.170 (0.106)	-0.126 (0.107)
Is sophomore?	-0.409* (0.185)	-0.423* (0.186)	-0.478* (0.186)	0.020 (0.176)	0.033 (0.176)	0.030 (0.171)
Is junior?	-0.283 (0.189)	-0.287 (0.190)	-0.436* (0.191)	0.281 (0.176)	0.284 (0.176)	0.244 (0.176)
Is senior?	-1.022* (0.203)	-1.035* (0.204)	-1.260* (0.207)	0.333 (0.181)	0.340 (0.181)	0.179 (0.174)
Lives on campus?	0.507* (0.155)	0.498* (0.155)	0.436* (0.157)	0.092 (0.135)	0.096 (0.136)	0.017 (0.133)
Voted in 2008 primary	0.044 (0.130)	0.044 (0.130)	0.033 (0.131)	-0.008 (0.124)	-0.012 (0.124)	-0.038 (0.121)
Voted in 2008 general election	0.581* (0.187)	0.581* (0.188)	0.578* (0.187)	0.107 (0.178)	0.109 (0.179)	0.055 (0.173)
Number of campaign activities, 2008–2009	0.092* (0.025)	0.089* (0.025)	0.072* (0.025)	0.157* (0.018)	0.159* (0.018)	0.148* (0.018)
Interest in national politics	0.158 (0.082)	0.162* (0.082)	0.148 (0.083)	0.196* (0.094)	0.195* (0.094)	0.192* (0.091)
Trust in government	-0.025 (0.139)	-0.026 (0.140)	-0.015 (0.143)	0.039 (0.129)	0.034 (0.130)	0.021 (0.114)
Partisan strength	-0.067 (0.058)	-0.069 (0.058)	-0.055 (0.060)	-0.063 (0.055)	-0.063 (0.055)	-0.049 (0.054)
Party ID	—	—	—	-0.071 (0.058)	-0.073 (0.058)	-0.082 (0.057)
Ideology	—	—	—	0.094 (0.065)	0.096 (0.065)	0.095 (0.062)
Network centrality (Z score)	—	0.049 (0.035)	0.035 (0.035)	—	-0.027 (0.025)	-0.044 (0.030)
Social proximity to the elite (Z score)	—	—	0.418* (0.070)	—	—	0.473* (0.075)
Intercept	-2.683* (0.371)	-2.670* (0.373)	-2.544* (0.376)	-1.728* (0.384)	-1.738* (0.385)	-1.833* (0.355)
N	2590	2590	2590	992	992	992
Log likelihood	-1213.311	-1212.316	-1192.726	-1172.282	-1171.963	-1148.435
AIC	2454.623	2454.631	2417.453	2378.565	2379.925	2334.870

Notes: Reported coefficients are from logistic regressions for the Turnout models and negative binomial regressions for the Campaign Activities models (with standard errors in parentheses). All estimates are based on five imputations. The omitted reference category for academic year is freshman. For both outcome variables, a likelihood ratio test suggests that model 3—which includes the social network measures—provides a better fit than does model 1 (for Turnout, $\chi^2(\text{DF} = 2) = 41.2$; $p < 0.001$). For the Campaign Activities models, $\chi^2(\text{DF} = 2) = 47.7$; $p < 0.001$).

* $p < 0.05$.

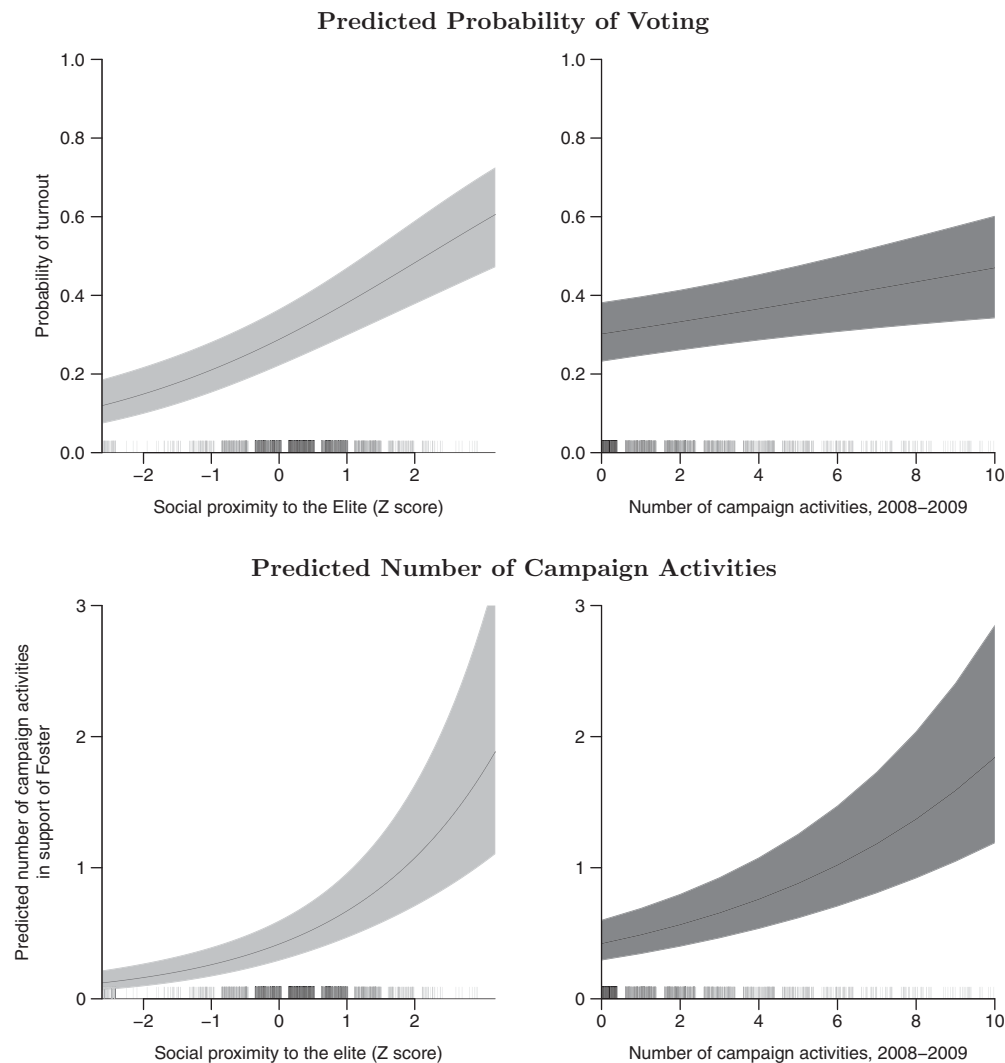
past participation, also predicts turnout.²¹ Rather, we wish to emphasize the similarity between the results from this modern election to those from the nineteenth century. A strong relationship persists between turnout and social proximity to the elite despite the absence of

viva voce voting, and even after controlling for past participation and political attitudes.

Our theory suggests that social proximity to the elite should encourage turnout in the 2010 municipal election, but it should have no effect on turnout in the 2008 election, which occurred two years before Foster ran for office and before many students arrived at William & Mary. Thus, turnout in 2008 provides a placebo test for our model (Sekhon 2009). In the test, we model

²¹ Compared to 2008 nonvoters, 2008 voters have a 0.11 greater probability of voting in 2010 ($\text{FD} = 0.32 - 0.21 = 0.11$; 95% CI [0.05, 0.18]).

Figure 4. An Individual's Predicted Probability of Voting and Predicted Number of Campaign Activities in Support of Foster Given her Social Proximity to the Elite (Based on the 2010 Williamsburg Election)



Notes: The lines indicate the predicted probabilities of turnout (top row) or number of campaign activities in support of Foster (bottom row). The shaded regions show the 95% confidence intervals. All other covariates are set to their medians. Rug plots show the distribution of the variable on the X axis, jittered to better show frequencies.

Source: Table 3, Turnout model 3 (top row) and Campaign Activities model 3 (bottom row).

2008 turnout as the outcome variable using as explanatory variables the remaining covariates from Table 3 model 3. If the coefficient associated with social proximity remains large and statistically significant in this model, it contradicts our interpretation of the results, suggesting instead that people more proximate to Foster participate at higher rates due to shared interests or reasons other than social proximity. As Table G1 of Online Appendix G demonstrates, however, the coefficient is almost zero, lending support to our approach.

The last three models in Table 3 display the relationship between social proximity and participation in favor of Foster's campaign, providing a new test of

our Elite Proximity-Support Hypothesis I. The models are estimates from negative binomial regressions²² where the outcome variable is a count of the number of campaign activities in which the respondent engaged specifically in support of Foster, measured in the post-election wave.²³ In addition to the controls from the turnout models, we control for party ID and ideology.

²² Likelihood ratio tests suggest these negative binomial regressions significantly outperform poisson models.

²³ These activities include the same type of acts measured in the 2008-2009 participation index as well as becoming a fan of Foster's Facebook page. These participation measures may suffer from an overreporting bias, but the survey's online administration may

The positive and statistically significant coefficient associated with social proximity suggests, all else equal, more proximate respondents participated in more campaign activities in support of Foster. The bottom row of Figure 4 plots this relationship as well as that between the 2008–2009 participation index and participation in support of Foster. Someone at the mean of social proximity is predicted to perform 0.4 acts in support of Foster compared to 0.7 acts for someone one standard deviation above ($FD = 0.68 - 0.42 = 0.26$; 95% CI [0.15, 0.40]). For campaign activism in 2008–2009, someone at the mean is predicted to perform 0.6 acts in support of Foster, compared to 0.8 acts for someone one standard deviation above ($FD = 0.76 - 0.55 = 0.21$; 95% CI [0.14, 0.32]). Like the turnout model, this analysis suggests a strong relationship between social proximity to the elite and participation in support of the elite, even after controlling for individuals' past engagement in similar forms of participation.

In the analyses above, we show in three different settings that voting has a strong relationship with social proximity to elites, even after including a variety of controls. Nonetheless, these apparent social influence effects may instead arise due to unobserved confounds arising from associates' shared environments and interests. We therefore subject all of our estimates to sensitivity analysis, originally developed by VanderWeele (2011). In Tables H1–H6 of Online Appendix H, we show that all of our social influence estimates prove robust to large levels of bias from these or other confounds.

CONCLUSION

Most people do not know personally someone who has run for elected public office (Nie, Junn, and Stehlik-Barry 1996, 48), but many may know someone who knows such an elite. We are all connected to elites through this social network, some of us more closely than others. We have shown that people more closely connected to elites vote at higher rates and, when choosing among candidates, often choose members of the party they are closest to socially. This study was only possible because the three elections we examine share the ability to identify ordinary citizens, elites, and relationships between them. The attributes of these communities that allowed this measurement also make them unrepresentative in many ways. Indeed, recent work in social influence suggests these relationships will differ from context to context (e.g., Huckfeldt, Johnson, and Sprague 2004; Siegel 2009). It is for this reason that we replicate our analysis in three different elections, which vary in time, place, and measurement strategy. The large differences between these studies—different age groups, different voting institutions, different candidate pools—demonstrate the robustness of our results, but also limit our ability to examine the

specific mechanisms driving these patterns. To what extent do the patterns we observe emerge from social pressure, information diffusion, candidate emergence or asocial processes such as networks of shared interests? We must leave it to future work to answer these questions. Despite these limitations, we think these results contribute to our understanding of social influence in political behavior, local elections, and voting more generally.

Over the last three decades, work on social influence in political behavior has relied primarily on name-generator survey batteries, exploring social influence arising among people's closest friends and family. These few individuals comprise only a small fraction of the hundreds of people we interact with directly and the thousands we are linked to through intermediaries. Our theory presented here specifies *how* these remaining relationships influence individual voting decisions, with influence flowing from elites to their immediate associates and then spreading to less proximate individuals in the network. Our work suggests close associates can connect individuals to the broader climates of opinion within their community. As such, we bridge recent research on close relationships to earlier contextual studies focusing on the correlation of attitudes or behaviors of individuals within neighborhoods (e.g., Tingsten 1963), religions (e.g., Langton and Rapoport 1976), and social classes (e.g., Berelson, Lazarsfeld, and McPhee 1954). We demonstrate that individuals are not atoms disconnected from the broader context, nor are they social sponges, indiscriminately absorbing the prevailing political culture. They are tied to the broader context by specific and idiosyncratic networks of associations. By specifying this network we can better understand the ways individuals respond to the contexts in which they are embedded.

The elites we study are all candidates for local office and our work provides clarity into the social forces underlying these important, but understudied elections. Our work extends to local elections the logic of friends-and-neighbors voting—a phenomenon of interest to political scientists for over half a century (e.g., Gimpel et al. 2008; Stokes and Miller 1962; Tatalovich 1975). This theory, largely developed and tested in congressional districts and statewide electoral offices, suggests geographic proximity to a candidate encourages voters to “back the home-town boy” (Key 1949, 41). Our Newport results suggest that in local elections, where all candidates are from the hometown, social proximity rather than geographic proximity alone may drive this relationship.²⁴ Our analysis of Virginia's state elections demonstrates that local social ties also help shape state-level elections, lending support to the claim that friends-and-neighbors voting occurs through the transmission of information (Bowler, Donovan, and Snipp 1993). While recent work on the topic suggests more salient races may influence down-ticket elections (Meredith 2013), this result suggests less-salient races

mitigate this problem because overreporting is less common in online settings (Holbrook and Krosnick (2010); but see Ansolabehere and Hersh (2012)). This analysis is the only outcome on which we rely that is self-reported rather than objectively validated.

²⁴ In Online Appendix I, we show that our network measures add additional explanatory power over simple measures of geographic proximity.

may also impact more salient ones. The electoral fate of geographically-distant statewide candidates is linked to the social network within voters' communities.

Finally, our analysis of two nineteenth century elections demonstrates both the usefulness and limitations of voting theories originally developed with modern data. Much like modern elections, people with more socioeconomic resources at stake are more likely to participate. Unlike more modern elections, socioeconomic resources provide little insight into the divisions between supporters of competing parties. In contrast, the social network measures we develop clearly separate voters from nonvoters and supporters of a party from supporters of another—suggesting we can better understand elections by considering not only what citizens have, but also who they know. Unfortunately, specifying the relationships between individuals proves difficult using the random samples that dominate the voting literature. The challenge for electoral scholars is to specify these relationships without abandoning the many advantages that come from random samples.

SUPPLEMENTARY MATERIALS

To view supplementary material for this article, please visit <https://doi.org/10.1017/S000305541600071X>

REFERENCES

- Ahn, T. K., Robert Huckfeldt, Alexander K. Mayer, and John Barry Ryan. 2013. "Expertise and Bias in Political Communication Networks." *American Journal of Political Science* 57 (2): 357–73.
- Ahn, T. K., Robert Huckfeldt, and John Barry Ryan. 2014. *Experts, Activists, and Interdependent Citizens: Are Electorates Self-Educating?* New York, NY: Cambridge University Press.
- Ansolahehere, Stephen, and Eitan Hersch. 2012. "Validation: What Big Data Reveal About Survey Misreporting and the Real Electorate." *Political Analysis* 20 (4): 437–59.
- Bartels, Larry M. 2000. "Partisanship and Voting Behavior, 1952–1996." *American Journal of Political Science* 44 (1): 35–50.
- Berelson, Bernard R., Paul F. Lazarsfeld, and William N. McPhee. 1954. *Voting: A Study of Opinion Formation in a Presidential Campaign*. Chicago: University of Chicago Press.
- Bohmer, David. 1977. "The Causes of Electoral Alignments: Some Considerations on How Partisan Behavior is Shaped." In *Law, Society, and Politics in Early Maryland: Proceedings of the First Conference on Maryland History*, eds. Aubrey C. Land, Lois Green Carr, and Edward P. Papenfuss. Baltimore: Johns Hopkins University Press.
- Bohmer, David. 1978. "The Maryland Electorate and the Concept of a Party System in the Early National Period." In *The History of American Electoral Behavior*, eds. Joel H. Silbey and Allan G. Bogue. pp. 146–73.
- 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 (London)*. Princeton New Jersey: Princeton University Press, 489 (7415): 295–8.
- Borgatti, Stephen P. 2005. "Centrality and Network Flow." *Social Networks* 27 (1): 55–71.
- Bourke, Paul F., and Donald A. DeBats. 1977. "Identifiable Voting in Nineteenth-Century America: Toward a Comparison of Britain and the United States Before the Secret Ballot." *Perspectives in American History* 11: 259–88.
- Bourke, Paul F., and Donald A. DeBats. 1980. "Individuals and Aggregates: A Note on Historical Data and Assumptions." *Social Science History* 4 (2): 229–50.
- Bourke, Paul F., and Donald A. DeBats. 1985. "Restoring Politics to Political History." *The Journal of Interdisciplinary History* 15 (3): 459–66.
- Bourke, Paul F., and Donald A. DeBats. 1987. "The Structures of Political Involvement in the Nineteenth Century: A Frontier Case." *Perspectives in American History* 3: 207–38.
- Bourke, Paul, and Donald A. DeBats. 1995. *Washington County: Politics and Community in Antebellum America*. Baltimore: Johns Hopkins University Press.
- Bourke, Paul, Donald DeBats, and Thomas Phelan. 2001. "Comparing Individual-Level Returns with Aggregates a Historical Appraisal of the King Solution." *Historical Methods: A Journal of Quantitative and Interdisciplinary History* 34 (3): 127–34.
- Bowler, Shaun, Todd Donovan, and Joseph Snipp. 1993. "Local Sources of Information and Voter Choice in State Elections Microlevel Foundations of the "Friends and Neighbors" Effect." *American Politics Quarterly* 21 (4): 473–89.
- Brody, Richard A., and Paul M. Sniderman. 1977. "From Life Space to Polling Place: The Relevance of Personal Concerns for Voting Behavior." *British Journal of Political Science* 7 (03): 337–60.
- Campbell, David E. 2013. "Social Networks and Political Participation." *Annual Review of Political Science* 16 (1): 33–48.
- Campbell, Angus, Philip E. Converse, Warren E. Miller, and Donald E. Stokes. 1960. *The American Voter*. New York: John Wiley and Sons.
- Campbell, David E., John C. Green, and Geoffrey C. Layman. 2011. "The Party Faithful: Partisan Images, Candidate Religion, and the Electoral Impact of Party Identification." *American Journal of Political Science* 55 (1): 42–58.
- Carlson, Taylor N., and Jaime E. Settle. 2016. "Political Chameleons: An Exploration of Conformity in Political Discussions." *Political Behavior* 38 (4): 817–59.
- Clausen, Aage R. 1968. "Response Validity: Vote Report." *Public Opinion Quarterly* 32 (4): 588–606.
- Cormen, Thomas H., Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein. 2009. *Introduction to Algorithms*. Cambridge, Massachusetts: MIT Press.
- DeBats, Donald Arthur. 2004. The Politics of German Americans: Three Case Studies from an Industrial Age. In *German-American Immigration and Ethnicity in Comparative Perspective*, eds. W. Helbig and Walter D. Kamphoefner. Madison: MaxKade Institute for German-American Studies, University of Wisconsin Press.
- DeBats, Donald A. 2008. "A Tale of Two Cities: Using Tax Records to Develop GIS Files for Mapping and Understanding Nineteenth-Century U.S. Cities." *Historical Methods: A Journal of Quantitative and Interdisciplinary History* 41 (1): 17–38.
- DeBats, Donald A. 2009. "Using GIS and Individual-Level Data for Whole Communities A Path Toward the Reconciliation of Political and Social History." *Social Science Computer Review* 27 (3): 313–30.
- DeBats, Donald A. 2011. "Political Consequences of Spatial Organization." *Social Science History* 35 (4): 505–41.
- DeBats, Donald A., and Mark Lethbridge. 2005. "GIS and the City: Nineteenth-Century Residential Patterns." *Historical Geograph* 33: 78–98.
- DeCanio, Samuel. 2007. "Religion and Nineteenth-Century Voting Behavior: A New Look at Some Old Data." *Journal of Politics* 69 (2): 339–50.
- Downs, Anthony. 1957. *An Economic Theory of Democracy*. New York: Harper and Row.
- Eulau, Heinz. 1980. "The Columbia Studies of Personal Influence: Social Network Analysis." *Social Science History* 4 (2): 207–28.
- Eveland, William P., and Myiah Hutchens Hively. 2009. "Political Discussion Frequency, Network Size, and Heterogeneity of Discussion as Predictors of Political Knowledge and Participation." *Journal of Communication* 59 (2): 205–24.
- Festinger, Leon. 1957. *A Theory of Cognitive Dissonance*. Stanford: Stanford University Press.

- Festinger, Leon. 1964. *Conflict, Decision, and Dissonance*. Stanford: Stanford University Press.
- Fowler, James H., Michael T. Heaney, David W. Nickerson, John F. Padgett, and Betsy Sinclair. 2011. "Causality in Political Networks." *American Politics Research* 39 (2): 437–80.
- Fowler, James H., and Oleg Smirnov. 2005. "Dynamic Parties and Social Turnout: An Agent Based Model." *American Journal of Sociology* 110 (4): 1070–94.
- Gelman, Andrew, David Park, Boris Shor, and Jeronimo Cortina. 2009. *Red State, Blue State, Rich State, Poor State: Why Americans Vote the Way They Do (Expanded Edition)*. Princeton: Princeton University Press.
- Gerber, Alan S., Donald P. Green, and Christopher W. Larimer. 2008. "Social Pressure and Voter Turnout: Evidence from a Large-Scale Field Experiment." *American Political Science Review* 102 (01): 33–48.
- Gerber, Alan S., Donald P. Green, and Ron Shachar. 2003. "Voting May Be Habit-Forming: Evidence from a Randomized Field Experiment." *American Journal of Political Science* 47 (3): 540–50.
- Gimpel, James G., Kimberly A. Karnes, John McTague, and Shanna Pearson-Merkowitz. 2008. "Distance-Decay in the Political Geography of Friends-and-Neighbors Voting." *Political Geography* 27 (2): 231–52.
- Hammarberg, Melvyn. 1977. *The Indiana Voter: The Historical Dynamics of Party Allegiance During the 1870s*. Chicago: University of Chicago Press.
- Heckman, James A. 1979. "Sample Selection Bias as a Specification Error." *Econometrica* 47 (1): 153–61.
- Holbrook, Allyson L., and Jon A. Krosnick. 2010. "Social Desirability Bias in Voter Turnout Reports Tests using the Item Count Technique." *Public Opinion Quarterly* 74 (1): 37–67.
- Honaker, James, Gary King, Matthew Blackwell, et al. 2011. "Amelia II: A Program for Missing Data." *Journal of Statistical Software* 45 (7): 1–47.
- Huckfeldt, R. Robert. 1983. "Social Contexts, Social Networks, and Urban Neighborhoods: Environmental Constraints on Friendship Choice." *American Journal of Sociology* 89 (3): 651–69.
- Huckfeldt, Robert, Paul Allen Beck, Russell J. Dalton, and Jeffrey Levine. 1995. "Political Environments, Cohesive Social Groups, and the Communication of Public Opinion." *American Journal of Political Science* 39 (4): 1025–54.
- Huckfeldt, R., P. E. Johnson, and J. Sprague. 2002. "Political Environments, Political Dynamics, and the Survival of Disagreement." *The Journal of Politics* 64 (01): 1–21.
- Huckfeldt, R. Robert, Paul E. Johnson, and John D. Sprague. 2004. *Political Disagreement: The Survival of Diverse Opinions within Communication Networks*. New York: Cambridge University Press.
- Huckfeldt, Robert, Matthew T. Pietryka, and Jack Reilly. 2014. "Noise, Bias, and Expertise in Political Communication Networks." *Social Networks* 36: 110–21.
- Huckfeldt, Robert, and John Sprague. 1995. *Citizens, Politics, and Social Communication: Information and Influence in an Election Campaign*. New York: Cambridge University Press.
- Key, V. O. 1949. *Southern Politics in State and Nation*. New York: A. Knopf.
- Killworth, Peter D., Eugene C. Johnsen, Christopher McCarty, Gene Ann Shelley, and H. Russell Bernard. 1998. "A Social Network Approach to Estimating Seroprevalence in the United States." *Social Networks* 20 (1): 23–50.
- King, Gary, James Honaker, Anne Joseph, and Kenneth Scheve. 2001. "Analyzing Incomplete Political Science Data: An Alternative Algorithm for Multiple Imputation." *American Political Science Review* 95 (1): 49–69.
- Klofstad, Casey. 2010. *Civic Talk: Peers, Politics, and the Future of Democracy*. Philadelphia, PA: Temple University Press.
- Langton, Kenneth P., and Ronald Rapoport. 1976. "Religion and Leftist Mobilization in Chile." *Comparative Political Studies* 9 (3): 277–308.
- Lazarsfeld, Paul F. 1948. *People's Choice: How the Voter Makes Up His Mind in a Presidential Campaign*. New York: Columbia University Press.
- Leighley, Jan E., and Jonathan Nagler. 1992. "Individual and Systemic Influences on Turnout - Who Votes, 1984." *Journal of Politics* 54: 718–40.
- Makse, Todd, and Anand E. Sokhey. 2012. "Yard Sign Displays and the Enthusiasm Gap in the 2008 and 2010 Elections." *PS: Political Science & Politics* 45 (04): 694–9.
- McClurg, Scott D. 2006. "The Electoral Relevance of Political Talk: Examining Disagreement and Expertise Effects in Social Networks on Political Participation." *American Journal of Political Science* 50 (3): 737–54.
- McCormick, Tyler H., Amal Moussa, Johannes Ruf, Thomas A. DiPrete, Andrew Gelman, Julien Teitler, and Tian Zheng. 2013. "A Practical Guide to Measuring Social Structure Using Indirectly Observed Network Data." *Journal of Statistical Theory and Practice* 7 (1): 120–32.
- McLeod, Jack M., Dietram A. Scheufele, and Patricia Moy. 1999. "Community, Communication, and Participation: The Role of Mass Media and Interpersonal Discussion in Local Political Participation." *Political Communication* 16 (3): 315–36.
- McPherson, Miller, Lynn Smith-Lovin, and James M. Cook. 2001. "Birds of a Feather: Homophily in Social Networks." *Annual Review of Sociology* 27: 415–44.
- Meredith, Marc. 2013. "Exploiting Friends-and-Neighbors to Estimate Coattail Effects." *American Political Science Review* 107 (04): 742–65.
- Miller, Patrick R., Piotr S. Bobkowski, Daniel Maliniak, and Ronald B. Rapoport. 2015. "Talking Politics on Facebook Network Centrality and Political Discussion Practices in Social Media." *Political Research Quarterly* 68 (2): 377–91.
- Miller, Warren E., and J. Merrill Shanks. 1996. *The New American Voter*. Cambridge, MA: Harvard University Press.
- Mutz, Diana Carole. 2006. *Hearing the Other Side: Deliberative versus Participatory Democracy*. New York, NY: Cambridge University Press.
- Nickerson, David W. 2008. "Is Voting Contagious? Evidence from Two Field Experiments." *American Political Science Review* 102 (01): 49–57.
- Nie, Norman H., Jane Junn, and Kenneth Stehlik-Barry. 1996. *Education and Democratic Citizenship in America*. Chicago: University of Chicago Press.
- Oliver, J. Eric, and Shang E. Ha. 2007. "Vote Choice in Suburban Elections." *American Political Science Review* (03): 393–408.
- Panagopoulos, Costas. 2010. "Affect, Social Pressure and Prosocial Motivation: Field Experimental Evidence of the Mobilizing Effects of Pride, Shame and Publicizing Voting Behavior." *Political Behavior* 32 (3): 369–86.
- Richey, Sean. 2008. "The Autoregressive Influence of Social Network Political Knowledge on Voting Behaviour." *British Journal of Political Science* 38 (03): 527–42.
- Rolfe, Meredith. 2012. *Voter Turnout: A Social Theory of Political Participation*. New York: Cambridge University Press.
- Rosenstone, Steven J., and John Mark Hansen. 1993. *Mobilization, Participation and Democracy in America*. New York: Macmillan.
- Rozett, John. 1977. "The Social Base of Party Conflict in the Age of Jackson." Ph.D. dissertation. Ann Arbor: The University of Michigan.
- Ryan, John Barry. 2011. "Social Networks as a Shortcut to Correct Voting." *American Journal of Political Science* 55 (4): 753–66.
- Sekhon, Jasjeet S. 2009. "Opiates for the Matches: Matching Methods for Causal Inference." *Annual Review of Political Science* 12 (1): 487–508.
- Shortridge, Ray M. 1980. "Voter Turnout in the Midwest, 1840–1872." *Social Science Quarterly* 60 (4): 617–29.
- Siegel, David A. 2009. "Social Networks and Collective Action." *American Journal of Political Science* 53 (1): 122–38.
- Siegel, David A. 2013. "Social Networks and the Mass Media." *American Political Science Review* 107 (04): 786–805.
- Sinclair, Betsy. 2012. *The Social Citizen: Peer Networks and Political Behavior*. Chicago: University of Chicago Press.
- Sinclair, Betsy, Margaret McConnell, and Melissa R. Michelson. 2013. "Local Canvassing: The Efficacy of Grassroots Voter Mobilization." *Political Communication* 30 (1): 42–57.
- Sokhey, Anand Edward, and Scott D. McClurg. 2012. "Social Networks and Correct Voting." *The Journal of Politics* 74 (03): 751–64.
- Stokes, Donald E., and Warren E. Miller. 1962. "Party Government and the Saliency of Congress." *Public Opinion Quarterly* 77: 945–56.

- Tatalovich, Raymond. 1975. "'Friends and Neighbors' Voting: Mississippi, 1943–73." *The Journal of Politics* 37 (03): 807–14.
- The American Association for Public Opinion Research. 2015. *Standard Definitions: Final Dispositions of Case Codes and Outcome Rates for Surveys*. 8th ed. AAPOR.
- Tilly, Charles. 2007. *Democracy*. New York, NY: Cambridge University Press.
- Tingsten, Herbert. 1963. *Political Behavior: Studies in Election Statistics*. Totowa, NJ: Bedminster Press. Originally published in 1937.
- Trounstein, Jessica. 2009. *Political Monopolies in American Cities: The Rise and Fall of Bosses and Reformers*. Chicago: University of Chicago Press.
- VanderWeele, Tyler J. 2011. "Sensitivity Analysis for Contagion Effects in Social Networks." *Sociological Methods & Research* 40 (2): 240–55.
- Verba, Sidney, Kay Lehman Schlozman, and Henry E. Brady. 1995. *Voice and Equality: Civic Voluntarism in American Politics*. Cambridge, MA: Harvard University Press.
- Wolfinger, Raymond E., and Steven J. Rosenstone. 1980. *Who Votes?* New Haven: Yale University Press.