

Political Donor Polarization: Observing Consumptive Behavior using a Network Approach *

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American politics has recently been defined by unprecedented levels of partisan polarization. Given the concurrent rise of the amount of money in politics, many have suggested a connection between money in politics and polarization. This paper uses the occurrence of a specific polarizing event, former Wisconsin Governor Scott Walker's introduction and passage of Act 10, to analyze the relationship between donor polarization and mass polarization. Using political donation data from the Wisconsin Campaign Finance Information System (CFIS) and using the network science measure of modularity, this paper shows that political donor networks polarized during the 2012 election cycle at the same time as the electorate. This result suggests that political donors were likely not the main contributors to the polarization in the state and provides evidence for the 'consumption' model of political donations.

Keywords: polarization, political donations, network analysis, state politics

Political campaign finance plays an important role in the American political system. This significance is evidenced by the attention that academic researchers pay to the topic as well as the many different contexts in which campaign finance is studied. For example, research has concluded that: campaign contributions ultimately impact one-third of all congressional roll-call votes (Roscoe and Jenkins 2005), even small contributions can sway politicians' votes (Stratmann 1991), campaign finance is a contributor to gender inequities between the Democratic and Republican parties (Barber, Butler, and Preece 2016; Crowder-Meyer and Cooperman 2018; Kitchens and Swers 2016; Thomsen and Swers 2017), existing professional networks are beneficial to new politicians (Bonica 2017), and candidates spend a significant amount of time fundraising (Torres-Spelliscy 2017). In addition, the concurrent rise of money in politics and political polarization has led to the idea that the two are connected (Francia et al. 2005; McCarty, Poole, and Rosenthal 2006). Barber concludes that "the connection between donors and candidates is an important part of the story of the polarization of American politics" (2016).

*Code and data available at: github.com/rossdahlke

The folk-theory of political donors is of smokey backrooms and access-oriented donors who seek to have a direct influence on policy making. However, the psychological processes of donors are thought to be similar to ordinary voters. Political donations can be thought of an extension of voting. In other words, both actions are political consumption that seek to improve a preferred candidate's chances of winning. Ansolabehere, de Figueiredo and Snyder summarized this idea by stating, "In our view, campaign contributing should not be viewed as an investment, but rather as a form of consumption—or, in the language of politics, participation" (2003).

Donations can be seen as an outlet for motivated citizens to increase their participation beyond just turning out to vote when they perceive the stakes of elections to be high (Hill and Huber 2017). Even when individuals have an economic interest in the outcome of an election, donations are found to be motivated by existing policy agreements and not an expectation of access (Barber, Canes-Wrone, and Thrower 2016). For example, donations from business executives have been found to be "best understood as purchases of 'good will' whose returns, while positive in expectation, are contingent and rare" (Gordon, Hafer, and Landa 2007).

Although the psychological process of making a political campaign contribution can be thought of as similar to voting, there are significant demographic and ideological differences between donors and voters. People with lower incomes, less education, and those not in professional jobs are less likely to be politically engaged, including making political donations (Laurison 2016). Donors to the Democratic and Republican parties were previously summarized as being "Limousine Liberals" and "Corporate Conservatives" (Francia et al. 2005). However, that narrative has shifted recently in the wake of the *Citizens United* supreme court case which allowed individuals to contribute more money to political causes and the rise of small-dollar donors Albert and Raja (2020).

While Democrats and Republicans draw their bases of electoral support from different geographic bases, major campaign donors are highly concentrated geographically. These

“big-donor neighborhoods” are unrepresentative of the country as a whole and point to these communities having a distinct political culture (Bramlett, Gimpel, and Lee 2011). In both parties, donors are more ideologically extreme than non-donating voters (Francia et al. 2003; Hill and Huber 2017) and wealthy donors who make up the “big money” in politics are especially partisan (McCarty, Poole, and Rosenthal 2006).

This ideological extremity shown by political donors has led some scholars to suggest that political donors are contributors to the partisan polarization of the politics of the United States (Francia et al. 2005). This idea is supported by the observation that both political polarization and campaign spending have risen in conjunction (McCarty, Poole, and Rosenthal 2006). Despite this observed correlation, there is little evidence for the causal relationship of donors causing political polarization. Many have found that political donations don’t influence polarization (Harden and Kirkland 2016; Keena and Knight-Finley 2019; Raja and Wiltse 2012). Furthermore, it is likely the case that the causal arrow flows the other direction, and it is a more polarized electorate and candidates that have led to more polarized donors (Harden and Kirkland 2016; Keena and Knight-Finley 2019; Raja and Wiltse 2012).

If political donors were found to be contributors of political polarization, the ever-growing amount of money in politics could forewarn even greater levels of polarization. In addition, political donors contributing to polarization would mean that donors are even more influential in the broader realm of politics than previously believed. Further, there could be policy responses to attempt to curb the amount of money in politics and therefore political polarization, such as more strict contributions limits.

Studying polarization, particularly among political donors can be difficult because of the myriad of potential confounding factors that can contribute to polarization (Harden and Kirkland 2016). In addition, polarization is generally a phenomenon that gradually increases or decreases over time (Pew Research Center 2017). However, this paper leverages a singular event, former Wisconsin Governor Scott Walker’s proposition and passage

of Act 10, a “budget repair bill” that ended collective bargaining for teachers unions, to examine political donor polarization in the state of Wisconsin. Given the recent research which has pointed to the polarization of political donors as being *reactive* instead of *causal* to broader polarization, we could expect political donors to follow the trend of voters and polarize after introduction of Act 10 and subsequent events. Particularly, we would expect levels of polarization among political donors to be closely aligned with levels of polarization of the mass electorate under the consumption model of political donations where donations are a participatory extension of voting.

H_1 : Political donors in the State of Wisconsin polarized during the 2012 election cycle compared to the 2010 election cycle and maintained their level of polarization in the 2014 election cycle.

If this hypothesis is supported, these results would strengthen the evidence for politics donors being *reactive* to their political environment as we would expect under Ansolabehere, de Figueiredo and Snyder’s consumption model of political giving.

Alternatively, if political donors are instead *contributors* to mass and legislative polarization, as is suggested by some scholars, we would expect to see hypothesis two.

H_2 : Polarization levels stayed the same in the 2012 election cycle compared to the 2010 election cycle.

If hypothesis two is supported, the result would suggest that political donors helped to create the polarized political environment that we see today.

In addition, this paper makes a methodological contribution to campaign finance research by taking a network approach to measuring donor polarization by using modularity as a measure of polarization. This method has been used elsewhere in the social sciences to study congressional polarization (Waugh et al., n.d.; Zhang et al. 2008) and polarization in social media networks (Conover et al. 2011; Garcia et al. 2015; Guerra et al. 2013). This paper conceives of the political donor landscape of donors and candidates acting as nodes who are connected by donations that act as edges. This method is impor-

tant in studying political donor networks because it takes into consideration real-world actions, such as in network studies of polarization among member of congress where voting records (Guerra et al. 2013) and co-sponsorships (Zhang et al. 2008) are used to study polarization opposed to surveys administered to donors that rely on self-reported ideology and partisanship.

Wisconsin Context

Both Wisconsin's legislators and mass public are among the most polarized in the nation (Cramer 2016), and the state has been used by academics to examine how political actions unfold in contentious and highly divisive environment (Bode et al. 2018). Although many state legislatures are also experiencing polarization (Shor 2015), Wisconsin is unique in that there is a single event that many point to in creating "the most politically divisive place in America" (Kaufman 2012).

In 2011, newly-elected Republican Governor Scott Walker introduced Act 10, a "budget reconciliation bill" that stripped public school teachers of collective bargaining via their union. Up to 100,000 people protested this "anti-union bill" at the State Capitol and even occupied the capitol building for a period of time (Sewell 2011). Democratic lawmakers fled to Illinois in an effort to delay or stop the bill from passing into law (Layton 2011). In 2012 there was an unsuccessful election to recall Governor Walker.

Wisconsin Governor Scott Walker's self-anointed "divide and conquer" politics (Blake 2012) has left a political divide in Wisconsin that persists to today. The result is that "divisive politics ruled Wisconsin over the last decade" (Marley and Beck 2019). The Marquette Law School poll headed by Charles Franklin has called public opinion in Wisconsin a "lesson in the two worlds of Wisconsin" where "it seems often as if people have not only differing opinions but differing views of facts and realities" (Borsuk 2017). This discrete event and its long-lasting consequences provides a unique opportunity to study massively polarized politics that can be attributed back to a single event. In addition,

Wisconsin is a competitive swing state that reflects a roughly 50-50 split similar to the country as a whole.

Methodology

All data on political contributions came from the Wisconsin Campaign Finance Information System (“Wisconsin Campaign Information System,” n.d.). I exported all contributions to State Assembly, State Senate, and Gubernatorial races from the 2010, 2012, and 2014 elections. This dataset does not include donations to party committees, although it does include disbursements from these committees. I manually created a table of the parties of each of all the campaigns receiving contributions in this timeframe and added the party of the campaign receiving the donation to this dataset.

To clean and analyze my data I used the statistical programming language R (R Core Team 2013; Wickham et al. 2019). I started with 1,499,603 donations. I then filtered out 3,503 unitemized/ anonymous donations, removed punctuation from the names of the donors, and used Open Refine (Kelli 2013) via the `ref_inr` R package (Muir 2018) to standardize names (for example, Jim versus James). Next, I created a unique identifier for donors by combining their standardized name with their zip code. This identifier was created to be able to link donors who contributed across multiple campaigns in multiple years without considering two different people, with the same name, from different locations to be the same person. Identity resolution is notoriously difficult and is the biggest limitation of this paper. This study uses some of the most advanced identity resolution techniques available, but there is always inevitable error. While this potential error is difficult to calculate, we expect that any error would be random distributed across the three election cycles. In other words, one can assume that the levels of error are the same across all three election cycles. So, the directional conclusions of this paper remain valid.

Next, I estimated the partisanship of each donor in each election cycle by taking the percent of donations that each donor gave to Republicans divided by their donations to

Republicans and Democrats. I took that “percent donated to Republicans” and rescaled it from -1 to 1, where -1 represents the most Democratic donors, and 1 the most Republican donors. I also calculated each individual’s party bin: if more than 75% of donations were to Democrats, they were labeled as a Democrat; if more than 75% of donations were to Republicans, they were labeled as a Republican; if their donations were somewhere inbetween, they were labeled as being a bipartisan donor.

To quantify the levels of polarization in each election cycle, I calculated two statistics: network modularity and average absolute partisanship of donors.

First, political donations can be thought of as a network where donors and candidates are nodes and donations connecting donors and candidates are edges. This conceptualization of the political donor landscape as network allows us to examine the network structure and calculate network statistics on the graph of donors and candidates. One of the most useful network statistics for measuring polarization in a network is modularity (Newman 2006).

The modularity of a graph measures the strength of the division of groups (such as political parties) by calculating “the number of edges falling within groups minus the expected number in an equivalent network with edges placed at random” (Newman 2006). The modularity of a network falls in the range $[-1/2, 1]$. If the modularity is positive, the number of edges that remain within each group is greater than the expected number to remain in-group based on chance. The higher the modularity, the greater the concentration of edges within each groups. In other words, the higher the modularity of a network, the higher the polarization among the groups. Formally, the equation to calculate modularity Q is:

$$Q = \frac{1}{2m} \sum_{ij} \left[A_{ij} - \frac{k_i k_j}{2m} \right] \delta(g_i, g_j)$$

In this equation $m = \frac{1}{2} \sum_i k_i$ is equal to the strength of all the ties in the network,

$k_i = \sum_j A_{ij}$ is the strength/ weighted degree of the i th node, g_i is the group (in this case, party/ party bin) to which the i belong, and $\delta(g_i, g_j) = 1$ if i and j belong to the same group (party/ party bin) and 0 if they do not belong to the same party/ party bin.

I calculated the modularity of the network graphs of each election cycle (2010, 2012, 2014) using the igraph R package (Csardi and Nepusz 2006). I used candidates' declared parties and donors' party bin as the groups for the modularity calculation. The modularity of the network graph of each election is in Table 1.

In addition to calculating the change in modularity of each of the election cycles, I also analyzed the change in mean absolute partisanship of the donors in each election cycle.

I defined a donor's absolute partisanship as the absolute value of their partisanship score (which is on a scale from -1 to 1). Therefore, the larger a donor's absolute the partisanship, the higher percentage of their money that they contributed to a single party. To calculate the significance in the difference of the mean absolute partisanship, I use a bootstrap methodology with 1,000 replications using the infer R package (Bray et al. 2020). This paper uses a non-parametric permutation method because of the non-Normal distribution of partisanship of the donors (98% of donors across all election cycles only contribute to a single party).

To conduct a hypothesis test with a permutation, you first compute the mean for each group of the measure that you're interested in—in this case we are calculating the mean partisanship of donors in two given election cycles. This is difference $d = \bar{X}_1 - \bar{X}_2$. Next, you pool the data and randomly draw new groups of data of equal sizes of the original groups. You then calculate the difference in sample means from these random draws and compare them to your original sample mean (Wilcox 2003). This paper uses 1,000 replications, primarily to be able to calculate a p-value to the thousandths place. A p-value for a permutation test is done by calculating the proportion of randomly drawn d 's that are greater than the original d (Butar and Park 2008). In essence, you figure out what proportion of your randomized draws have a sample mean that is greater or less

than your observed data—how many randomized groupings having a result that is as or more extreme than your observed groups. If your p-value is below your pre-specified level (this paper uses the standard .05) you can reject the null hypothesis.

The results of the bootstrap are found in Table 2.

These measures of polarization among Wisconsin donors can be compared to party identification among likely voters in Wisconsin in Table 3 for 2010 (Kniss 2010), 2012, and 2014 (Franklin, n.d.).

Results

The results of this analysis show that political donors in Wisconsin polarized during the 2012 election cycle, the same time that mass polarization occurred in the state (see Table 3). This phenomenon is best visualized in Figure 1. This figure uses the Yifan Hu layout algorithm (Hu 2005) in the Gephi software (Bastian, Heymann, and Jacomy 2009), a force-directed graphical layout of networks that seeks to repulse clusters of nodes from one another. The Yifan Hu layout algorithm is a standard among social scientists studying networks such as online networks (Adalat, Niazi, and Vasilakos 2018; Hemsley et al. 2015; Khonsari et al. 2010; Rehman et al. 2020). This visual representation shows two distinct clusters of donors (Democrats and Republicans) that are reasonably close to one another in the 2010 election cycle and then polarize significantly in the 2012 election cycle and remain polarized in 2014.

This graphical representation reflects statistical measures of polarization within the networks. Table 1 shows the modularity of the networks in the 2010, 2012, and 2014 election cycles. In 2010, the modularity of the donor network is 0.4. The modularity for the 2012 cycle climbs to 0.49 and settles in at 0.48 during the 2014 cycle. The interpretation of modularity is the higher the number, the more observed polarization within the network. As such, the rise in modularity in the 2012 cycle depicts polarization within the donor network in 2012 compared to 2010. And then the steady modularity in the 2014 cycle

reveals a stabilization of the level of polarization observed in the 2012 election cycle.

One limitation of a modularity calculation is that it does not quantify uncertainty. To validate the results of the modularity calculation, I conducted a hypothesis test. Table 2 compares the average absolute donor partisanship in 2012 compared to 2010 and 2014 compared to 2012. As the table shows, donors in the 2012 election cycle became much more partisan with an average change in their absolute partisanship of 0.04209 (CI = 0.04012-0.04396, p-value = <.001). However, there was not a statistically significant change in mean absolute partisanship in the 2014 election cycle compared to the 2012 election cycle (-2.8×10^{-4} , CI = -0.00085-0.00032, p-value = 0.35).

In comparison, in 2010, 40% of likely voters in Wisconsin identified as independents. By election day 2012 that number dropped to 12%. Just as both the modularity calculation and the absolute partisanship calculation showed levels of polarization dropping slightly in 2014 compared to 2012, the percent of likely voters who identified as independents slightly rose by election day 2014 to 17%. See Table 3 for levels of partisanship polarization among likely voters in Wisconsin.

Taken together, the results of the modularity calculations and the hypothesis tests support the rejection of the null of H_1 and fail to reject the null of the alternative H_2 . In other words, political donors in Wisconsin had a statistically significant increase in polarization in the 2012 election cycle and slightly less polarization in 2014—the exact same pattern observed among the electorate of Wisconsin.

Additional data and graphs are provided to contextualize the polarization that is observed among political donors in Wisconsin. Figure 2 shows the partisan flow of political donors across the election cycles, including the massive influx of new donors in both the 2012 and 2014 election cycles. Figure 3 shows the distribution of the size of donors (amount contributed) by partisanship. Figure 4 shows the partisan shift of donors who contributed in both the 2010 and 2012 election cycles. And Figure 5 is a map representing the geographic polarization among political donors.

Discussion

The failure to reject the null of H_1 suggests that political donors were likely not the main contributors to the extreme levels of polarization first seen in the state in 2012. Other factors, such as a more polarized primary electorate in the wake of the Tea Party in 2010 (Jacobson 2012) electing Governor Walker in the first place, appear to be the contributors to mass polarization and political donors in Wisconsin.

These results also provide evidence for the ‘consumption’ model of political donations. Ansolabehere, de Figueiredo and Syder’s (2003) conclusion that political donations are similar to voting in that they are both acts of political consumption are borne out in the results of this paper. Polarization of political donors happened in unison with the polarization of the electorate. The conclusion that we can draw is that the polarization of these two groups of people were a behavioral, participatory response to a changing political environment. Both the electorate and donors have specific acts of political consumption (voting and donating, respectively) that were both impacted in the same way at the same time. Previous scholars speculation that the concurrent rise of money in politics and broader polarization are connected. However, the results of this analysis concur with the more recent studies that conclude that if anything, the causal arrow flows from increased levels of mass polarization to more political contributions out of a sense of the heightened stakes of an election.

Further evidence for this consumption model is the idea that political donations are an extension of voting in the broader realm of political participation. The inflow of new donors show in Figure 2 suggests that the same mechanism that triggered mass polarization also spurred members of the mass electorate to go beyond voting and make political contributions. Previous research by Oklobzija (2016) reached a similar conclusion that “politically polarizing events bear dividends for extremist lawmakers” in California who raised more money as a result of polarizing political events. This inflow of new donors

was primarily small-dollar donors who were partisan in their contributions (see Figure 3).

One of the biggest shifts in the study of campaign finance is the recent rise of small-dollar donors for both Democrats (Albert and Raja 2020) and Republicans (Lott 2019). Although small-dollar donors were not as prominent in the 2010, 2012, and 2014 state-level elections in Wisconsin as recent national politics, they certainly played a role in the data of this study. Figure 3 shows that 100% partisan donors were on average much smaller donors than non-100% donors. Previous research into small-dollar donors has asked, “Are Small Donors Polarizing?” and came to the same conclusion as this paper of a more polarized politics as potentially spurring the rise of small-dollar donors and not the other way around (Keena and Knight-Finley 2019).

The “M-shaped” distribution of the donor sizes in different partisan bins in Figure 3 is noteworthy. Purely partisan donors were the smallest donors, but then the next most partisan donors were on average the largest donors. Further research should investigate the differences in purely-partisan donors and nearly-partisan donors. Potentially, these larger nearly-partisan donors are the ones who are the most strategic in their contributions whether their motivation is policy, surrogate representation (Baker 2019), or access.

Even though new donors are the likely explanation for most of the polarization observed in the networks, donors that contributed in both the 2010 and 2012 election cycles also showed significant movement. The shift among donors whose contributions are not purely partisan shift to be more Republican. Figure 4 shows how donors who were not pure partisan donors in 2010 much more often became purely Republican donors compared to Democratic donors. This sort of previously bi-partisan donors becoming single-party donors is very similar to the decreases in split-ticket voting (Bump 2016; Desilver 2016; Skelley 2018), especially if we conceive of political donations as an extension of voting on a participatory spectrum.

Finally, a major theme of polarization studies is the rise of geographic sorting. There

has been “an increased concentration of partisan behavior” that emphasizes “a local residential spatial pattern of geographic polarization” (Kinsella, McTague, and Raleigh 2015). “Partisan migrants” are found to “prefer to relocate in areas populated with copartisans” (Cho, Gimpel, and Hui 2012). The geography of political polarization in Wisconsin is well-studied by Cramer (2016) who uses the term “urban-rural divide” to explain the geographic polarization of the state. This divide that Cramer documents across Wisconsin manifests itself in the geography of polarization among political donors (Figure 5). Bases of strong Democratic support (mostly in urban centers) saw their donors become more Democratic. Conversely, areas that are strong bases of electoral support for Republicans also saw their donors become more Republican. Although the major sources of Democratic and Republican campaign money were once thought to occupy the same geographic area (Bramlett, Gimpel, and Lee 2011), the triumvirate of the rise of small-dollar donors, a decrease in bi-partisan donors across election cycles, and an urban-rural divide with geographic partisan sorting results in a geographic landscape of donor polarization that mirrors the polarization of the populous.

In short, it appears that political donations are an extension of voting, an outlet for political participation when individuals perceive the stakes of the election to be high (Hill and Huber 2017). And so it would be reasonable to find that political donors are not the cause of political polarization. But in fact, more polarized donors are a reflection of polarization seen elsewhere in American politics.

In addition to the substantive contribution of evidence of political donations as a “consumptive” behavior, this paper also makes a methodological contribution of applying network analysis to campaign finance. Previous research showed the value of using campaign finance data to map donors and candidates onto a spatial common-space (Bonica 2014). This paper demonstrates the value of a network approach to studying campaign finance. While the term “donor networks” is often used to refer to individuals with formal or informal connections that are generally social or business-related, this paper conceives

of “donor networks” more literally. This literal interpretation uses donors and candidates as nodes in a network that are connected by donations or edges. Placing campaign finance data within this network structure allows one to use the gambit of social network analysis methodologies—such as this paper using modularity to measure polarization.

Future research can build upon this conception of the campaign finance landscape as a network and apply other developed network science techniques such as clustering to identify donors that are most statistically similar to one another or directed networks to study the flow of money across the landscape.

Conceiving political donations as a literal network structure allows for new ways to look at existing campaign finance debates. This paper uses the novel network science measurement of modularity to study the polarization of political donors. The conclusion of this research is support for the consumption model of politics and the idea of political donations being a participatory extension of voting.

Tables

Table 1: Modularity calculation for the donor networks in each election cycle. Higher modularity means more polarization.

Election Cycle	Modularity
2010	0.3987161
2012	0.4912367
2014	0.4797913

Table 2: Bootstrapped difference-in-means test with 1,000 replications comparing mean partisanship of donors.

Election Cycle	T	CI	p
2012 compared to 2010	0.04209	0.04012-0.04396	<.001
2014 compared to 2012	-0.00028	-0.00085-0.00032	0.35

Table 3: Party Identification among Wisconsin likely voters.

Party	7/15/2010 poll	10/28/2012 poll	10/26/2014 poll
Democrat	24%	42%	33%
Republican	26%	42%	45%
Independent/ Neither	40%	12%	17%

Figures

Figure 1

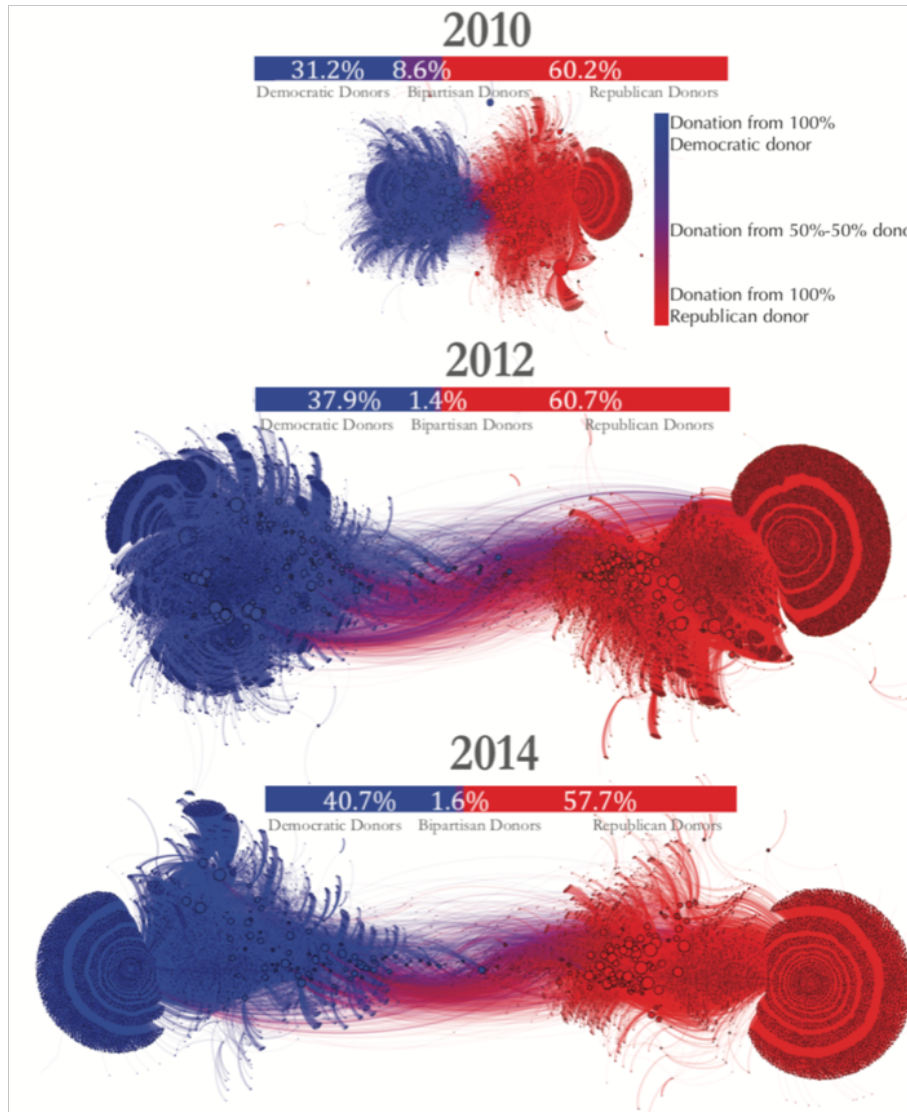


Figure 1: Visual representation of Wisconsin donor networks in the 2010, 2012 and 2014 election cycle using the Yifan Hu layout algorithm. Each dot/ node is a donor or campaign and lines/ edges connecting them are donations. Nodes sized by in-degree (incoming donations). Nodes and edges are colored by the partisanship of the donor. Percentages on the bars represent the percent of donors in each party bin.

Figure 2

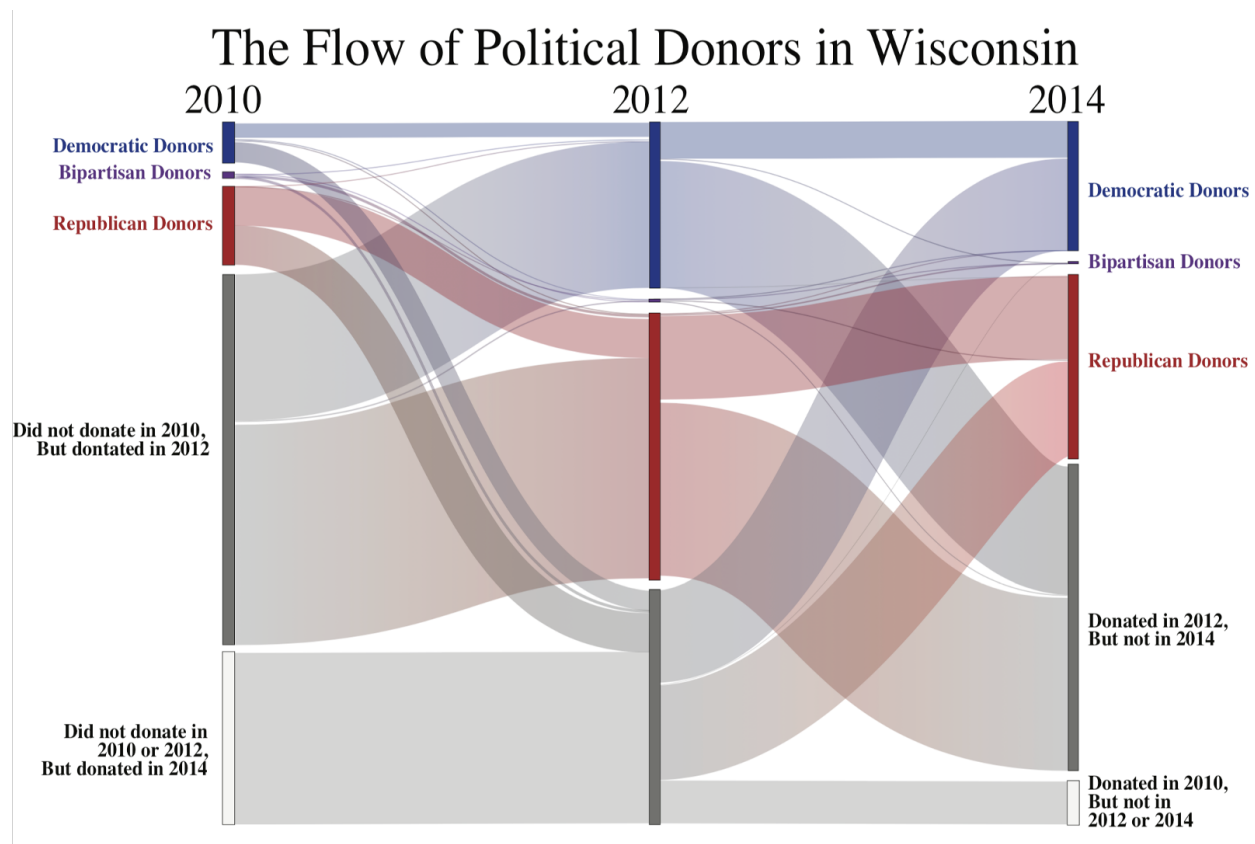


Figure 2: Sankey diagram of the flow of political donors in 2010, 2012, and 2014 election cycles in Wisconsin. The vertical bars are proportional to the number of donors in each bin.

Figure 3

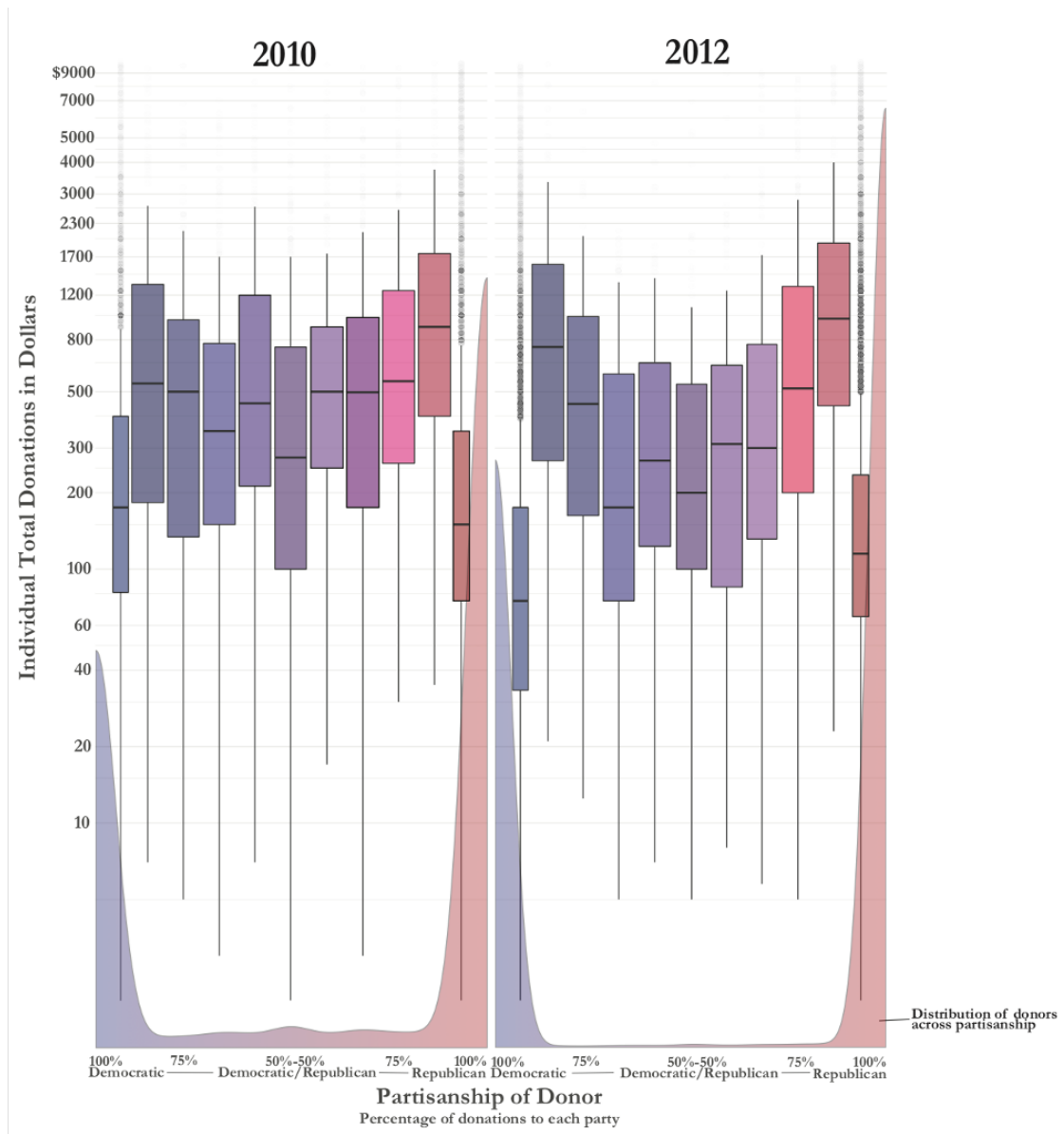


Figure 3: This box and whisker plot is grouped by the partisanship of the donors in the 2010 and 2012 election cycles. Note that the y-axis is shown on a log10 scale for clarity. The partisan distribution is shown along the bottom of the x-axis.

Figure 4

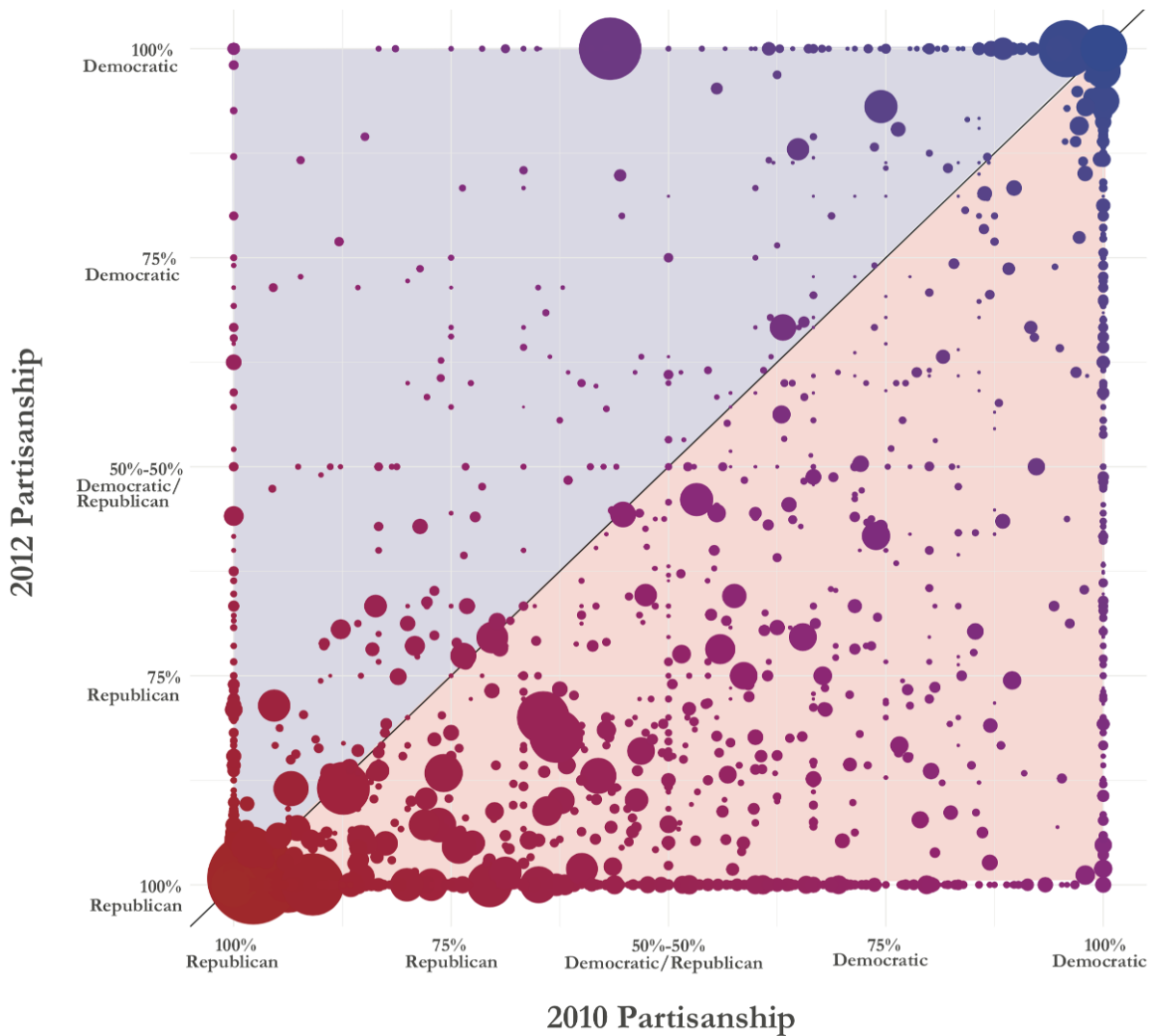


Figure 4: Every dot is a donor who contributed in 2010 and 2012. The bigger the dot, the more money they contributed. The x-axis is their partisanship in the 2010 election cycle and the y-axis is their partisanship in the 2012 election cycle. If the donor is to the right of the center diagonal line, they became more Republican. If they are to the left of the line, they became more Democratic.

Figure 5

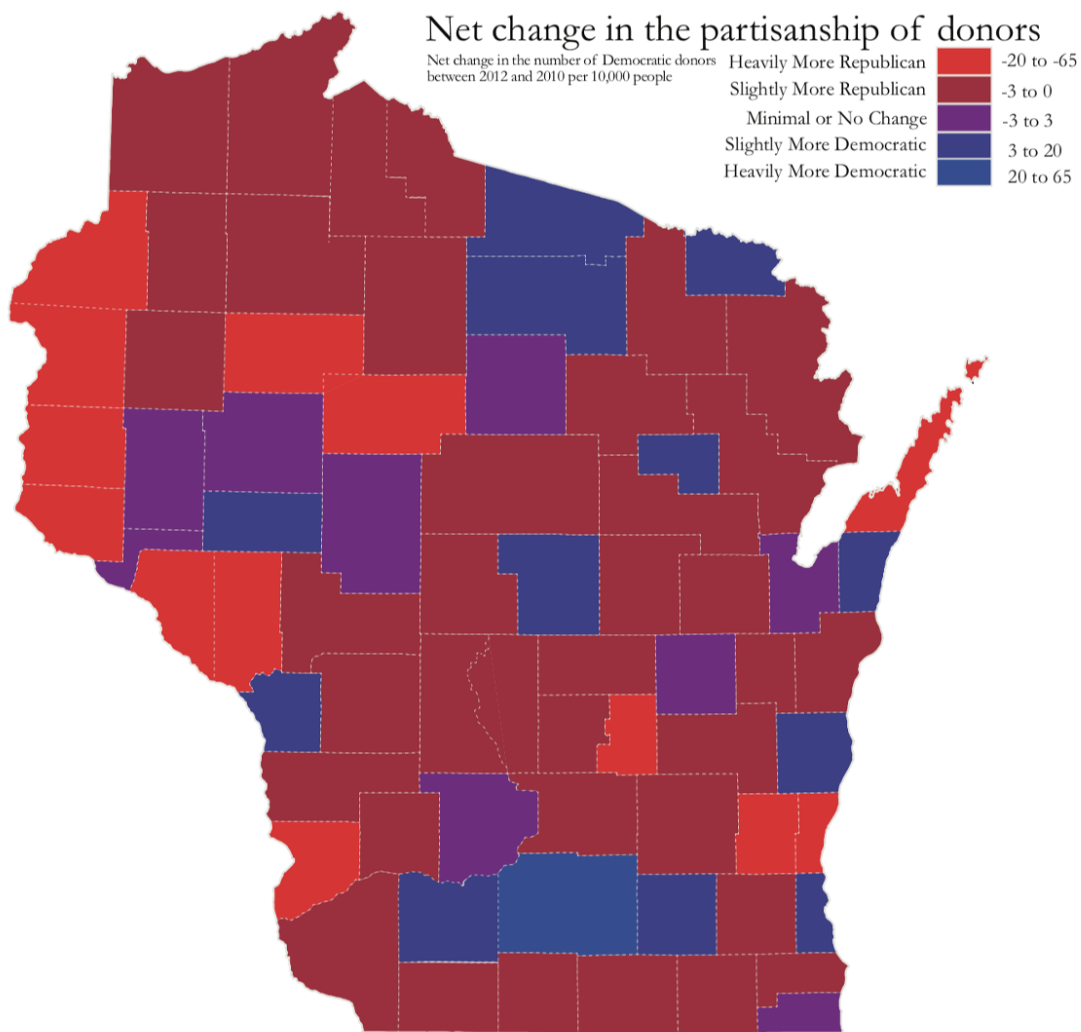


Figure 5: This map shows the polarization of donor networks across Wisconsin's counties based on the net change of donors in each county per 10,000 residents. The red counties had a net increase in Republican donors, blue counties had a net increase for Democrats, and the purple counties had little or no change.

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