




ORIGINAL ARTICLE

Do Male and Female Legislators Have Different Twitter Communication Styles?

Daniel M. Butler¹ , Thad Kousser²  and Stan Oklobdzija³ 

¹Department of Political Science, Washington University in St. Louis, St. Louis, MO

²Department of Political Science, University of California, San Diego, La Jolla, CA, USA

³School of Public Policy, University of California, Riverside, Riverside, CA, USA

Corresponding author: Thad Kousser, email: tkousser@ucsd.edu

(Received 12 February 2021; revised 31 August 2021; accepted 22 September 2021)

Abstract

Communication is a fundamental step in the process of political representation, and an influential stream of research hypothesizes that male and female politicians talk to their constituents in very different ways. To build the broad dataset necessary for this analysis, we harness the massive trove of communication by American politicians through Twitter. We adopt a supervised learning approach that begins with the hand coding of over 10,000 tweets and then use these to train machine learning algorithms to categorize the full corpus of over three million tweets sent by the lower house state legislators who were serving in the summer of 2017. Our results provide insights into politicians' behavior and the consequence of women's underrepresentation on what voters learn about legislative activity.

Keywords: state legislators; gender and Twitter; constituent communication

Elected officials' communication efforts are an important part of representation (Fenno 1977; Grimmer 2013). Historically, traditional media outlets have been a major way that politicians have reached their constituents. However, traditional media outlets have demonstrated bias against women, making it harder for them to reach voters (Baitinger 2015; Heldman, Carroll, and Olson 2005; Kahn 1992, 1994). This poses a challenge for women in politics; they need to reach voters but face obstacles in utilizing a primary avenue, traditional media outlets, for doing so. The rise of Twitter and other social media tools represents one way that women may be able to overcome bias in media coverage by allowing politicians to circumvent traditional media and directly reach voters. In this paper, we test whether gender state legislators are more likely to use Twitter. We also explore whether gender predicts how politicians communicate with the public through Twitter.

We evaluate several hypotheses to learn about the differences between men and women. Some of these hypotheses tell us about what motivates politicians when they think about their election prospects. For example, previous work has argued that female politicians have strong incentives to portray themselves as conservative in order to counteract the stereotype that they are more liberal than their male, co-partisan colleagues (Koch 2000). Another stream of research shows that women work harder in political office (Kurtz *et al.* 2006), speak more in the legislature (Pearson and Dancey 2011), put more effort into their constituency service (Thomsen and Sanders 2020), and produce more legislation (Anzia and Berry 2011; Volden, Wiseman, and Wittmer 2013), suggesting that gender may affect how much effort state legislators will put into the time-consuming task of social media communication.

Other hypotheses have important implications for levels of descriptive representation. Previous work has found that politicians' communications can bias the information environment for voters. For example, Grimmer (2013) finds that politicians with more extreme preferences are more likely to communicate about policy issues and moderate politicians are more likely to discuss nonpolicy-related funding that they secure for their district. The differences in how these groups communicate allow extremist views to dominate the public policy debate. In a similar way, if men and women communicate differently, this has implications for what voters hear because women are underrepresented in office. Previous work has argued that women tend to work more on issues like education and health care (Foerstel and Foerstel 1996; Reingold 1992; Saint-Germain 1989; Swers 1998, 2002; Thomas 1994; Thomas and Welch 1991). If women also discuss these issues more, then electing more women will lead voters to hear and learn more about these issues. We test whether gender predicts how much legislators communicate on these issues.

To build the broad dataset necessary to undertake this analysis, we harness the massive trove of communication by American politicians through Twitter. Just as it has become a highly visible mode of political discourse in national politics (Garofoli 2018), social media is now one of the primary modes of political communication for state legislators. As we show, a majority of members of state lower houses have public Twitter handles, with the average lawmaker tweeting over 1,000 times. Together, the lower house state legislators we study produced over three million tweets in our period of study from October 2015 to July 2018.

This wealth of data presents both an opportunity and a challenge for state politics scholars. Lawmakers in statehouses all across the nation are speaking via the same medium, and doing so both during campaign seasons and while they are governing. Holding the medium constant, scholars can study what they have to say before, during, and after elections, whether the electoral rules under which they run affect their ideological positioning, whether citizen lawmakers speak differently from those in professional legislatures, or whether polarized statehouses produced more negative discourse. With user engagement data, scholars can determine what forms of political communication followers are most apt to like or retweet and whether this varies by state and party. But in order to answer such questions, researchers must make sense of a mountain of data (for review of prior work, see Jungherr 2016; Vergeer 2015).

The modern tools of machine learning can aid in the task of classifying the topics, tone, and content of the enormous amount of data that state legislators are producing every day on Twitter. Machine learning techniques for text analysis can be divided into two approaches. In the first, "unsupervised learning," researchers mine data for

attributes, such as the topics that cluster together, and then attribute meaning to the output of these algorithms. We adopt the second approach, “supervised learning,” a hybrid between qualitative and quantitative techniques that begins by applying human judgment to code texts and then uses these codings to train machine learning algorithms (see Grimmer and Stewart 2013; Peterson and Spirling 2018). Only after testing how precisely the algorithms can replicate human codings, and ensuring sufficient accuracy, do we move onto the stage of classifying the full set of state political tweets. This approach, which we detail below, has been used to study political tweets in gubernatorial elections by McGregor et al. (2016), in the Australian Parliament by Kousser (2019), and for US presidential candidates by Kousser and Oklobdzija (2018). Here, we apply it to state legislators, producing the largest set of classified tweets, including 3,580,727 spanning 49 distinct political systems, that we have seen in the literature.

Four main findings emerge from our analysis. First, women communicate more than men. They are more likely to have Twitter accounts and to use them. Second, in contrast to previous work, we find that female legislators’ tweets have a more positive tone than male legislators’ tweets. Third, women do discuss women’s issues more than their male counterparts, tweeting about both education policy and about health care policy more often. Fourth, gender does not appear to predict the ideological content of tweets after we control for legislators’ roll call records.

In what follows, we first draw hypotheses about gender and legislator communication from the previous literature. We then describe and validate our original dataset and use it to test the hypotheses we lay out. We summarize the findings and their implications in the conclusion.

Theories about Gender and Social Media Communication

Allocating Resources to Twitter Communication

Before politicians decide what to tweet, they must first decide whether they will tweet at all and how often they will do so. This choice is a strategic choice because committing to establishing a social media presence requires a significant investment of time. To study the “tweet styles” of Australian legislators, Kousser (2019) draws upon Fenno’s (1977) classic work on the home styles adopted by members of Congress in their districts. Kousser makes an analogy between Fenno’s concept of the allocation of resources that representatives devote to connecting with their districts and the allocation of effort that today’s lawmakers devote to connecting through social media.

While tweeting does not require the pecuniary investments that are necessary to set up and staff a district office or to fly home to meet with constituents, social media communication taxes a lawmaker’s most vital resource: time. According to Fenno (1977, 890), “Of all the resources available to the House member, the scarcest and most precious one, which dwarfs all others in posing critical allocative dilemmas, is his time.” Tweeting consistently requires a significant investment of time and attention from lawmakers. The price of this investment is magnified because most state legislators typically author their own Twitter feeds. They must do so while still fulfilling a host of other job commitments. Kurtz et al.’s (2006) survey, conducted long before social media added yet another demand to the busy

lives of state legislators, demonstrates the immense time commitment required by serving in a statehouse, even one that pays a small salary and is considered a part-time body.

We argue that allocating time toward tweeting is a costly activity whether the lawmaker communicates directly or indirectly to constituents.¹ How should a lawmaker's gender impact this allocational decision? We expect that female legislators will be more likely to establish a social media presence—both by creating a public Twitter account and by tweeting more often—than male legislators. There are a few reasons this might occur.

First, electoral discrimination might lead to “sex-based selection.”² Anzia and Berry (2011) argue that “if voters discriminate against female candidates, only the most talented, hardest working female candidates will win elections” (478; see also Fulton 2012, 2014; Pearson and McGhee 2013). Consistent with this argument, Anzia and Berry (2011) find that female members of Congress in fact outperform men when it comes to securing district funding and sponsoring and co-sponsoring legislation (see also Volden *et al.* 2013). If this same sex-based selection mechanism operates in state legislatures, we should expect female lawmakers to work harder when it comes to social media communication.

Second, women in state politics may be more motivated to devote time to tweeting because they are simply responding to the demands that constituents are making of them. In a field experiment conducted in collaboration with state legislators, Butler *et al.* (2020) find that when men and women legislators make the same outreach to constituents, constituents are more likely to ask women legislators to do more work. Legislators are motivated by a desire to win reelection and so craft their homestyles in order to please voters. If constituents are asking more of women, women may in turn do more in order to be responsive. Although many studies show that female candidates perform very well in general elections (Burrell 1994; Fox 2006; Newman 1994; Seltzer, Newman, and Leighton 1997), this may be because they are doing more to meet voter demands rather than because voters are not demanding more of them.

Third, traditional media outlets might be biased against women (Baitinger 2015; Heldman, Carroll, and Olson 2005; Kahn 1992, 1994). Women legislators might prefer to reach their constituents through traditional news outlets but prior studies reveal that they are simply not covered at the same rates as men (Heldman, Carroll, and Olson 2005; Kahn 1992). If they face obstacles to reaching voters through traditional news outlets, women may get around this issue by using Twitter to directly reach out to voters. Twitter thus allows them to circumvent the agenda power of media and communicate to voters on their own terms. While this is an advantage for both men and women, the gender bias in the media should make this a relatively more attractive option for women, leading to greater uptake of Twitter among female legislators.

¹When state politicians tweet, they are likely speaking directly to their immediate audience of followers but also indirectly others in their communities. Statehouse journalists increasingly cover tweets, and followers share tweets, either digitally or physically, through their social networks. Rosenstiel *et al.* (2015) show that a majority of non-Twitter users have seen tweets; they are exposed to them primarily on television, through friends, and in newspaper articles.

²This is a disputed position in the literature, with many studies finding that women candidates are not discriminated against in elections (see reviews in Brooks 2013; Hayes and Lawless 2016; Lawless 2015; Thomsen 2020).

Whatever mechanism is at work, we predict that female legislators will work harder than men to establish a social media presence. If this is true, it will be consistent with Kurtz et al.'s (2006, 332) finding that women in state legislators "devote an additional 7 percent of a full-time job to their legislative work compared with men." It will also fit with Evans and Clark's (2016) finding that female candidates tweeted more often than male candidates in the 2012 congressional elections and Thomsen and Sanders' (2020) study showing that women put more effort into their constituency service. In the social media realm, we set forth two empirical hypotheses to test the idea that female state lawmakers put forth more effort in this realm than their male counterparts.

Hypothesis 1: Compared to men, female state lawmakers will be more likely to establish a public Twitter account.

Hypothesis 2: Compared to men, female state lawmakers will tweet with greater frequency.

Sentiment

Prior research has tested whether gender predicts how negative politicians are in their public communications (e.g., Evans and Clark 2016). Gender stereotypes are a reason why gender might be correlated with the tone of communication. Society stereotypes women as being more helpful and kind and men as more aggressive and forceful (Fridkin and Kenny 2009; Huddy and Terkildsen 1993). If voters hold these stereotypes, this can shape what voters expect from them and how they respond to the tone of politicians' communications.

The effect of legislators' tone on voter evaluations is unclear. Some scholars conclude that voters punish women when they act in ways that are counter to existing stereotypes (e.g., Kahn 1996). Yet others conclude that taking a more negative tone helps women because it challenges those stereotypes (Lau and Pomper 2004).

In looking at social media, Evans et al. (2014) and Evans and Clark (2016) find that women are more likely to send more negative attack tweets (cf., Parmelee and Bichard 2012). Evans and Clark (2016) also find that the number of negative tweets (coming from both men and women) increases with the number of women in the race. One reason that women may be negative in their tone is that they are more likely to be attacked (Lazarus and Steigerwalt 2018). This may lead them to defend themselves with tweets that have a more negative tone because they are trying to deal with a more hostile political environment. On the other hand, women may feel more pressure regarding reelection (Krook 2020), leading them to try to win over constituents using a more positive tone in their tweets. We test whether this relationship identified at other levels of office holds among state legislators.

Hypothesis 3: Compared to men, female state lawmakers' tweets will be more likely to have a negative tone.

Issue Focus

Men and women may also differ in the policy content of their communication. At the most basic level, they might differ because they work on different issues. Previous

studies have identified health, welfare, and education as “women’s issues” (Saint-Germain 1989; Swers 1998, 2002). Other studies have instead focused on specific issues: for example, focusing on funding for breast cancer as opposed to all health care funding generally (Osborn 2012; Reingold 2000). We focus on the general categories, in part, because of the data. There are few tweets on any given specific issue. Indeed, there are some major categories that are rarely tweeted about. Looking at the more general categories provides greater variation for analysis. However, using general categories is a noisier measure. This is why “[s]tudies that adopt a more specific definition of women’s issues, or those issues that directly affect women, find a closer connection between women’s presence and policy outputs benefitting women” (Osborn 2012, 27). For this reason, our test is a harder test of gender differences in issue coverage.

Theoretically, research suggests that women may be more likely to work on these issues because they have more knowledge about these issues or simply because they personally prioritize these issues (Foerstel and Foerstel 1996; Reingold 1992; Thomas 1994; Thomas and Welch 1991). Either way, previous studies have found that women are more active in policy making on women’s issues.

During the committee stage, women are more likely to advocate for women’s interests (Swers 2002). And committees with more women are more likely to produce legislation that incorporates women’s interests (Berkman and O’Connor 1993; Norton 1999; Swers 2002). Further, female legislators in both the United States (Thomas 1994; Thomas and Welch 1991) and elsewhere (Considine and Deutchman 1994; Heath, Schwindt-Bayer, and Taylor-Robinson 2005) are more likely to serve on committees that deal with issues traditionally considered women’s issues.

Even after legislation leaves the committee, women are more active on these issues with female legislators in Congress participating more in floor debates on women’s issues (Swers 2000; Tamerius 1995). The focus is not simply because women have more opportunities to work on women’s issues. When men and women are given the same requests for help, female legislators are more likely to work on women’s issues than are their male counterparts (Butler 2014).

In identifying women’s issues, we use general categories rather than specific issues. In order to classify tweets by issue areas, we follow the categorizations created by the US Policy Agendas Project Codebook (see Adler and Wilkerson 2014).³ “Education” tweets are ones that fit the topics in 600: Education in that coding system, and include primary and higher education as well as tweets about universal pre-K. Our “Healthcare” category includes policies fitting into 300: Health, including references to Medicaid, the Affordable Care Act, Medicare for All, and prescription drug prices. We did not use a separate “Welfare” category, because tweets about this policy area were so rare. In our training dataset of 10,104 hand-coded tweets, only five used the word “welfare,” with three of these being references to corporate welfare and thus included under macroeconomic policy. After classifying all state tweets, following the machine learning process that we detail below, we test whether women are more likely than men to discuss women’s issues by comparing their rates of tweeting about education and health care.

³The codebook used by Adler and Wilkerson (2014), which we followed and adapted, is available at https://comparativeagendas.s3.amazonaws.com/codebookfiles/Topics_Codebook_2014.pdf, first accessed in June 2016.

Hypothesis 4: Compared to men, female state lawmakers' tweets will include more content on education and health care policy.

Ideology

In this study, we look at how legislators communicate with their constituents. In order to win elections, politicians want to publicly take positions that appeal to voters. This can lead politicians to try to shape their public record to appear to be more in line with their voters' preferences. Gender can affect this dynamic because voters think female, Republican legislators are more liberal than male, Republican legislators and they think female, Democratic legislators are more liberal than male, Democratic legislators (Koch 2000). If this leads voters to think that the voters are out-of-step with their constituents, then legislators have incentives to engage in more conservative position-taking in order to compensate for voters' stereotypes and present themselves as in-line with their constituents.

Hypothesis 5: Controlling for their positions on roll calls, female state lawmakers' tweets will be more conservative than male state lawmaker's tweets.

We might see a partisan difference in how legislators take positions because of their incentives to appeal to primary voters (Brady, Han, and Pope 2007). Democratic women may not try to appear more conservative than their voting record simply because being viewed as liberal can help them in the primary election (Sides et al. 2020). Thus, we may not see any relationship between gender and Twitter ideology among Democrats once we control for their roll call-based ideology. Republican women, by contrast, have incentives to appear more conservative in order to win their primary elections (Koch 2000). This suggests that we may only see Republicans engaging in position-taking to make themselves appear more conservative than they really are.

Hypothesis 6: Among Republicans, female state lawmakers' tweets will be more conservative than male state lawmakers' tweets when controlling for their roll call record.

Case Selection

We study legislators' tweets—and focus on the states—for four reasons. First, legislators control their tweets. In contrast to coverage in traditional media, the legislators are able to control what they write. This is important because it may be that the media systematically covers female politicians differently than male politicians. If we look at the media coverage, then it is unclear if we are measuring the actions taken by the legislators, the biases of the media, or a combination of both. Because we are interested in how politicians choose to portray themselves, looking at Twitter—a communication form they control—allows us to do that (see also Pearson and Dancey 2011; Pearson and McGhee 2013).

Second, at least in some legislatures, women face institutional constraints that affect their ability to influence legislation or other outcomes (Hawkesworth 2003). Twitter is a tool that is not controlled by legislative leaders or legislative institutions

and therefore allows us to measure the legislators' activity free from any biases against them or constraints placed on them.

Third, tweets are public information. We need access to what legislators say in order to measure how legislators portray themselves. Twitter has this information. Also, legislators cannot microtarget tweets. It is not the case that women can send messages only to female followers and men only to male followers. If they could, we might worry that the differences in content might reflect the specific group they were microtargeting. This is not the concern because we are getting the public tweets that they use to speak to all constituents, the media, and fellow legislators.

Fourth, social media and Twitter are an increasingly important form of communication. They are used extensively not only by American state legislators but also by politicians all around the world (Alles and Jones 2016; Jungherr 2016; Vergeer 2015). Understanding how politicians communicate through a medium that they use nearly every day is critical to understanding how they choose to portray themselves—and how the public perceives them—in the modern era.

Focusing on *state* legislators in particular provides a strong empirical ground in which to study the impact of gender on communication styles. In our dataset, there are 1,391 female lawmakers, making up 25.7% of state legislators overall. This includes 535 Republican women and 845 Democratic women. By contrast, during the 116th Congress, only 101 women served in the House, including 88 Democrats and just 13 Republicans.⁴ Compared to Congress, studying the states provides more opportunity to identify systematic patterns and to differentiate the effects of gender from those of party. And studying Twitter in state legislatures can provide a particularly unfiltered view of political communication. The scarcity of staff resources makes it more likely that state legislators send tweets themselves rather than relying upon staff, relative to members of Congress. When studying the impact of an individual attribute such as gender, this ability to observe direct personal behavior is valuable.

Data Collection

In order to combine human coding with machine learning techniques to classify the tweets of all lower house state legislators, we proceeded in four steps:

- Classifying and validating a “training set” of 10,104 hand-coded tweets
- “Pre-processing” the tweets to focus on their essential linguistic characteristics
- Training machine learning algorithms to replicate the hand codes, and testing their accuracy
- Classifying the full corpus of 3,580,727 tweets.

We began by creating a training set of tweets by American politicians over the last several years, categorized by their ideology, sentiment, whether they contained explicitly political subjects or not, the policy area that they address, and whether they constituted an opinion or a factual claim. We did so by building on the work done by Kousser and Oklobdzija (2018) who had a team of multiple research

⁴See “Women Serving in the 116th Congress (2019–2021),” Center for American Women and Politics, Rutgers University, <https://cawp.rutgers.edu/list-women-currently-serving-congress>.

Table 1. Measures of intercoder reliability: humans agreeing with humans

	Agreement rate	Cohen's kappa
Is the Tweet a factual claim or an opinion?	0.78	0.42
Ideology (liberal, neutral, or conservative)	0.78	0.66
Sentiment (negative, neutral, or positive)	0.75	0.60
Is the Tweet political or personal?	0.91	0.55
Topic: Immigration	1.00	0.87
Topic: Macroeconomics	0.97	0.73
Topic: Defense	0.96	0.77
Topic: Law and Crime	0.99	0.86
Topic: Civil Rights	0.98	0.71
Topic: Environment	1.00	0.84
Topic: Education	1.00	0.83
Topic: Health	0.99	0.82
Topic: Government operations	0.98	0.23
Topic: No policy content	0.91	0.78
Asks for a donation?	0.99	0.66
Asks to watch, share, or follow?	0.95	0.65
Asks for miscellaneous action?	0.93	0.57

Note: Based on an analysis of 1,217 tweets coded by rotating pairs of research assistants.

assistants (RAs) hand code a random sample of 8,206 tweets by the 2016 presidential candidates and their SuperPACs. These tweets were downloaded from Twitter's public API every week from October 2015 to July 2018.⁵

We then supplemented the database from Kousser and Oklobdzjia (2018) with 1,898 tweets from upper house state legislators and statewide officeholders that we coded for this project. We coded the tweets from the upper house state legislators because we wanted to make sure we included communications from those serving in a legislative context in our training dataset, but to train our algorithms on a set of tweets that was distinct from the lower house tweets in the full corpus that we later analyze. We downloaded the tweets from state legislators and state officials beginning in June 2018. We also had a group of RAs hand code the tweets using the same procedure as Kousser and Oklobdzjia (2018). In particular, all RAs worked from the same codebook and met regularly but coded tweets independently and were given only the text of the tweet, with no information about who sent it.

Table 1 provides data demonstrating that these coders reliably agreed in their independent categorizations. Using a subset of 1,217 presidential tweets, which were assigned to overlapping pairs of coders, we report two measures of intercoder reliability: the rate of agreement between coders and the Cohen's Kappa, measuring how much more likely our coders were to agree than two coders would be by random chance alone. Our rates of agreement range from 75% on our three-category sentiment measure to perfect agreement on three of our subject areas, with the Cohen's Kappa measures ranging from "fair" to "almost perfect" agreement levels for all but one of our variables. For our measure of ideology, the coders agreed 78% of the time, with a Cohen's kappa of 0.66, demonstrating that they could make this subjective judgment in a reliable, replicable manner.

⁵Kousser and Oklobdzjia (2018) found that their coding led to Cohen's kappa, which measures how much more likely our coders were to agree than two coders would be by random chance alone, range from 75% on the three-category sentiment score to perfect agreement on three of the subject areas.

Table 2. Examples of variables and tweets in each category

What issue did the tweet address?		
Education issue (3.4%)	Health care issue (1.6%)	
Thank you for helping me reach 1,400 followers. I appreciate your support for my campaign to #StopCommonCore in #AZ	The #opioidcrisis is very much here. This is a great initiative by @GovernorTomWolf and steps to link people to...	
Thanks to @SenatorBaldwin for fighting to expand access to #HigherEd	I've personally witnessed how Medi-Cal has changed lives. 1/3 of CA's population (13 million children & adults) is on Medi-Cal	
RT @HouseDemsLL: Superintendents call on @GovRauner to do his job and #SignSB1 to ensure our children receive a quality education	Live call w/ @ChrisMurphyCT thanking him for leading opposition to #TrumpCare. Watch. https://t.co/WZH1Ot0LUa @ActTogetherCT @womensmarchct	
Ideology		
Liberal (8.3%)	Neutral (72.1%)	Conservative (19.6%)
Medicaid enrollment has slowed in recent years, but it still serves nearly a quarter of our population	Top 5 AZ consumer fraud complaints and 5 warning signs: https://t.co/fznWrvoBr6	I was proud to stand up for the #righttolife again this week! #prolife #alpolitics
Workers' Rights: Check this video out w/ local verizon workers on strike in EB and what their cause is about. http://t.co/OFPMMr9 via @youtube	Last night, we went to Seussical. My cousin Wes was one of the leads and all three of my neighbors were in it. They all did great!	New approach to business tax filings greatly reduces gov't red tape—was recommended by our streamlining initiative
Sentiment		
Negative (19.0%)	Neutral (22.0%)	Positive (59.1%)
Lots that could have been fixed if WI GOP had not quit early: student loan debt, transpo funding, voucher accountability... #WICanDoBetter	Discrimination Lawsuit Filed Against Used Car Dealership in Mesa	Pa. Medical Marijuana legislation back on track #SenAHW #CoSponsor #SB3 #MedicalMarijuana #PA via
In Senate Approps being asked to spend \$1.9 million to cover @SchuetteOnDuty's fight against marriage equality. Waste of taxpayer dollars	Intel predicts a \$7 trillion self-driving future	Governor Brown signs 5 #EquityAndJustice bills today! A BIG step to promote safety, rehabilitation & family cohesion

In Table 2, we provide examples of tweets by state politicians that fit into the key categories that we focus on in this analysis. We show what types of text would highlight to our coders that a tweet had liberal, neutral, or conservative ideological content, as well as whether it conveyed negative, neutral, or positive sentiment. We report tweets that were identified as falling into the education or health care policy realms, two types of “women’s issues” highlighted by prior research (Saint-Germain 1989; Swers 1998, 2002) and following the policy categories used by the Policy Agendas Project to code federal bills by Adler and Wilkerson (2014). For each category, we report how prevalent it was in the full corpus of state lower house tweets, according to our classifications.

Table 3. Measures of classification accuracy: computers replicating humans

	Final testing accuracy	Cohen's kappa
Is the Tweet an opinion versus a factual claim?	0.65	0.29
Ideology (liberal, neutral, or conservative)	0.62	0.40
Sentiment (negative, neutral, or positive)	0.60	0.38
Is the Tweet political or personal?	0.83	0.38
Topic: Education	0.92	0.03
Topic: Health	0.97	0.20
Topic: Immigration	0.98	0.29
Topic: Macroeconomics	0.93	0.27
Topic: Defense	0.93	0.10
Topic: Law and crime	0.98	0.07
Topic: Civil rights	0.97	0.26
Topic: Environment	0.20	0.00
Topic: Government operations	0.98	0.07
Topic: No policy content	0.79	0.56
Asks for a donation?	0.99	0.00
Asks to watch, share, or follow?	0.95	0.50
Asks for miscellaneous action?	0.29	0.02

Note: Based on an analysis of a final testing set of 1,010 tweets after training on 8,084 tweets and testing on 1,010 tweets.

With this training set in hand, we then pre-processed the tweets through a series of steps that are commonly used in text analysis. We made every word lower case, removed URLs as well as additional links and emails, and deleted all alphanumeric text. Depending on whether it improved prediction accuracy for individual characteristics of tweets, we also removed unnecessary stop words such as “the,” “a,” or “an,” and removed screen names.

We then used the remaining text of each tweet, along with the human codings of their characteristics, to train a set of algorithms that fit models connecting the text to the codings. The algorithms that we used in this stage of the analysis were all taken from the scikit-learn Python library.⁶ To train the algorithms, we divided our training set of 10,104 tweets, using 80% of them to train, 10% to test the accuracy and select the most accurate algorithm, and 10% to use as a “final testing set,” which avoids overfitting a model. Table 3 reports the results of these final tests. The first column shows accuracy for each variable, which is rate at which the algorithm was able to correctly replicate the human coding. The second column reports the Cohen’s kappa, which is the improvement in accuracy over what we would expect by random chance if the algorithm always placed a tweet in the most prevalent category. For the policy variables, which take on only two values, accuracy is consistently high, registering over 90% in final testing accuracy for every policy area other than environmental policy (which does not feature in our analysis). While accuracy is lower for sentiment and ideology, reflecting the increased difficulty of correctly coding a variable that takes on three values, Cohen’s kappa values fall just over or just below the “moderate” threshold (Landis and Koch 1977, 165) for both variables.

⁶This library can be accessed at <https://scikit-learn.org/stable/>. We used the algorithm that produced the best accuracy for each tweet characteristic, including Multinomial Naïve Bayes (for sentiment, political, ideology, no policy content, factual claims or opinions, and whether a tweet made a miscellaneous ask), Bagging Classifier (for immigration, macroeconomic, health care, national security, crime, and whether a tweet asked for donations), and Linear SVC (for civil rights, governance, and whether a tweet asked a follower to watch, share, or follow). We also adjusted the tuning parameters to identify the best fit for each model.

Finally, we used the trained algorithms to classify an original dataset of the tweets of all lower house state legislators. To collect these tweets, we began by working with undergraduate RAs to search for the Twitter handles of all legislators serving in lower houses of 49 states—excluding Nebraska’s unicameral, nonpartisan house—in the summer of 2017. The RAs generally started by first performing a search for a one of the legislators on their list. In some states, especially more professional states, the handles were sometimes publicly listed together. More often, the RAs first found the Twitter account for one legislator and then found that the other legislators in their party in the state often linked to that account.

The RAs were unable to find accounts for many legislators, even after using several variants of the legislator’s names in the search. If the RAs were unable to find an account after searching for several minutes, they moved on to the next account. RAs also limited the sample to publicly listed accounts because we are interested in how legislators portray themselves to the public. Once the RAs identified a likely match, they looked at several tweets in the accounts to confirm that they had correctly identified the legislator’s account. In a few cases, the legislator did not have an account, but accounts were set up to parody the legislator. In other cases, legislators had multiple accounts. RAs looked through these accounts and identified the account(s) that were used as the legislator’s account during the legislative session. In some cases, legislators had multiple accounts that met these criteria; in those cases, all accounts were used in the study. In many of these cases, the dates the accounts were used did not overlap, suggesting that it may simply have been a case where the legislator forgot their password and decided to simply create a new account.

The RAs also recorded the genders,⁷ party affiliations, and districts represented by these 5,413 state legislators.⁸ Of these lawmakers, 2,014 (37%) did not have a public Twitter handle that we could identify. For the 3,399 (63%) of state legislators who did have a social media presence, we collected all available tweets from Twitter’s public API that were available in April 2019. This produced a dataset of 3,580,727 tweets. We then classified the features of these tweets and then calculated the average rates of each type of tweet for each tweeting legislator, along with their total tweet count. We merge this dataset with our data on legislator characteristics, successfully matching 3,129 state legislators to their tweet records. Finally, we appended data ideology based on statehouse roll call voting and national survey responses for all state legislators elected before 2016, using Shor and McCarty’s (2011), updated with data from all legislators elected before 2016 through their website.⁹

In order to explore the validity of the classifications the algorithms produced, we can compare our tweet-based ideology score with the roll call-based measures for state legislators collected by Shor and McCarty (2011). Although lawmakers may choose to vote and to communicate in slightly different ways (e.g., *Hypotheses 5 and 6*), there should be a strong correlation between the ideological positions that legislators take on the floor and the images that they convey on social media.

⁷In internet searches, we have not identified any state legislators who have made a public declaration of a non-binary gender identification as of May 2019.

⁸This is two more legislators that serve at any one time in state lower houses (5,411), likely because some legislators who were selected in special elections to replace others were also included.

⁹We collected roll call ideology scores in May 2019 from <https://americanlegislatures.com/data/>.

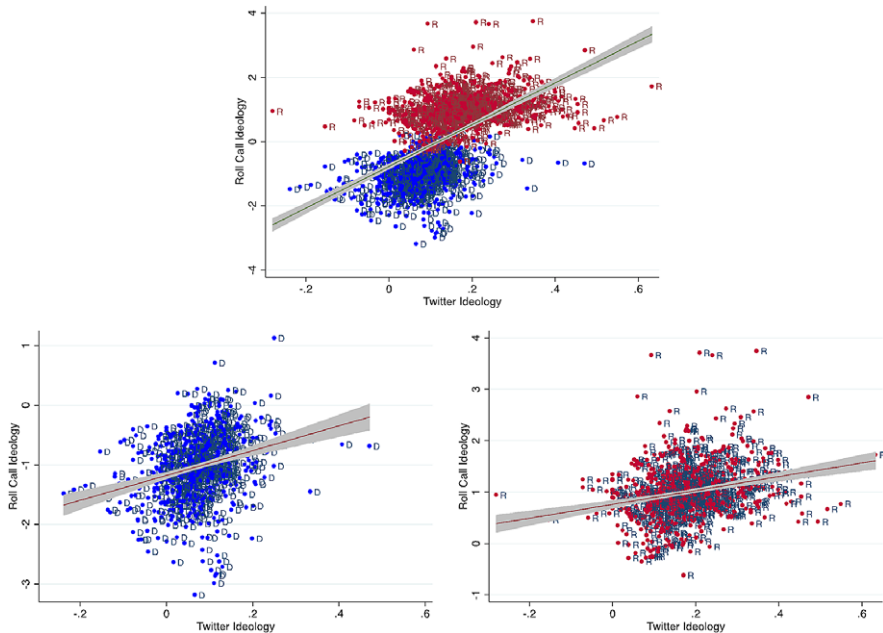


Figure 1. Testing the validity of tweet-based ideology measure.

Notes: All graphs compare our measure of the average ideology of each legislator's tweets with her roll call ideology, taken from updates to the dataset originally collected by Shor and McCarty (2011). Observations are all state lower house legislators elected before 2016 with more than 10 tweets.

As Figure 1 shows, a legislator's Twitter ideology score¹⁰ is positively correlated with her roll call ideology score produced by Shor and McCarty (2011). The scatterplot in the top panel, which includes a fit line with a 95% confidence interval, combines the data for both Democrats and Republicans and shows there is a positive relationship between the two measures. A regression of roll call ideology on Twitter ideology, reported in Supplementary Appendix Table 1, shows that this relationship is statistically significant at the 99% confidence level. Importantly, our tweet-based scores also predict roll call ideology within parties. The two scatterplots in the lower panel of Figure 1 show that this link holds within the Democratic Party and Republican Party. Regressions in Supplementary Appendix Table 1 demonstrate that these relationships are statistically significant. In fact, this relationship holds even within party for a model with state fixed effects. That means that when two legislators are in the same party and members of the lower house in the same state, the

¹⁰We calculate this score by multiplying a legislator's liberal tweets by negative one, neutral tweets by zero, and conservative tweets by one, and taking the sum. This yields a score that can range from negative one to one, with larger values representing a higher rate of conservative versus liberal tweeting. In Figure 1, we display data only for legislators with more than 10 tweets, to guard against small-sample outliers, and also remove the -0.75 score of Mississippi legislators Earle Banks, the most liberal frequent tweeter in our dataset, whose Twitter record consists almost entirely of campaign messages during his 2012 run for state Supreme Court.

lawmaker who tweets more conservatively is also likely to vote more conservatively, an important validation of this measure.

Analysis

In using the information on the 5,413 state legislators and their tweets to test our hypotheses, we compare the raw, bivariate differences between female and male legislators and then present full multivariate tests (see Oklobdzija, Kousser, and Butler 2022 for replication code and data). In these tests, we control for legislators' party affiliation and also include state fixed effects. These fixed effects capture the impact of all measurable features of a state—its level of legislative professionalism, its political culture, the party balance in its statehouse—along with all idiosyncratic characteristics that are fixed. These multivariate models with controls for party and with state fixed effects are the focus of our main analysis. Later, we present extensions that add additional factors to probe the robustness of the impact of gender and to explore other social media dynamics, often with a subset of our cases. We look at the impact of a legislator's racial and ethnic identity, as well as its intersection with gender, in a section of the paper devoted to this question. We explore the effects of a state's legislative professionalism (Squire 2017) and of how recently a lawmaker was elected in analyses reported in our Supplementary Appendix. Each of these reveals important lessons but does not alter the clear relationship between gender and social media activity and messaging. That is the central focus of our main analysis presented below.

We first test relationship between gender and a lawmaker's allocation of time to establishing a social media presence. [Hypotheses 1](#) and [2](#), respectively, predict that women will be more likely to have a Twitter handle and will tweet more frequently if they do. Regarding [Hypothesis 1](#), 71.9% of the 1,391 female state legislators in our dataset had a public Twitter account we could identify. For the 4,022 male state legislators, this figure was only 60.0%. This significant gender gap in social media presence also holds when we estimate regressions that control for the legislator's partisanship and state fixed effects. According to this model, reported in [Table 4](#), women establish handles at a 10.6% higher rate, all else equal, a result that is significant at the 99% confidence level.

Conditional on establishing an account, women also appear to tweet more often, tweeting an average of 1,200 times compared to 1,032 for men over the full history of their political Twitter account. That represents a 16% increase in how much more often women tweet compared to their male counterparts. In a multivariate model of tweet frequency, conditional on having an account, we also see that women send an estimated 121 more tweets, which again is strongly significant. The final model in [Table 4](#) ties these two aspects of social media together into a single estimation. Our assumption here is that there is a latent variable measuring each lawmaker's "tweet effort." For those who take on high enough values of this variable to establish an account, we can directly observe their effort through their number of tweets. For those who have no account, our observation of their effort is censored at zero tweets. This sort of censoring can be corrected for by a "Tobit" maximum likelihood model, with left-censoring at zero. The estimated impact of gender from this model, which is determined both by women's higher rates of tweeting and the greater likelihood that

Table 4. Does gender affect twitter activity?

	Does the legislator have a handle?	Tweet count	Tobit model
	0.106** (0.014)	118.771** (46.580)	327.635** (49.510)
Female legislator	0.053** (0.013)	314.705** (45.036)	354.528** (45.836)
Democratic legislator	Included	Included	Included
State fixed effects			
Observations	5,422	3,134	5,422
Adjusted <i>R</i> -squared	0.159	0.081	

Notes: Observations are all state lower house legislators in the first and third models, and all state legislators with Twitter accounts in the second model. Standard errors in parentheses, ** $p < 0.01$, * $p < 0.05$. Dependent variables are a dichotomous measure of whether a legislator had an official Twitter handle (in the first column) and the count of tweets from that handle (in the second).

Table 5. Does gender affect sentiment and attention to “women’s issues”?

	Sentiment score	Women’s issues	Education	Health care
	3.774** (0.949)	0.716** (0.148)	0.308** (0.104)	0.408** (0.106)
Female legislator	−4.972** (0.978)	0.876** (0.135)	0.280** (0.093)	0.596** (0.097)
Democratic legislator	Included	Included	Included	Included
State fixed effects				
Observations	3,134	3,134	3,134	3,134
Adjusted <i>R</i> -squared	0.107	0.080	0.086	0.052

Notes: Observations are all state lower house legislators with Twitter accounts. Standard errors in parentheses, ** $p < 0.01$, * $p < 0.05$. Dependent variables are the percentage of tweets: (1) with a positive sentiment, (2) addressing women’s issues, (3) addressing education issues, and (4) addressing health care issues.

they will establish a social media presence in the first place, is that female state legislators score an estimated 330 tweets higher on this scale of tweet effort.

Our test of [Hypothesis 3](#) explores the finding of Evans et al. (2014) and Evans and Clark (2016) that female candidates tweet with a more negative tone than their male counterparts. At first glance in our dataset, it appears that gender has little impact on sentiment, with women registering a 48.8% in our summary measure of sentiment¹¹ and men a 47.1%. Yet the multivariate analysis reported in [Table 5](#) shows that gender does have an apparent effect that is hidden, in a bivariate comparison, by its correlation with party affiliation. Women tweet an estimated 3.8 percentage points more positively, while Democrats tweet 5 percentage points more negatively. The full extent of this gender gap is only revealed when we control for party because the majority of women serving in office belong to the Democratic Party and Democrats tweet more negatively. So when we compare female legislators to their partisan counterparts, we see that women exhibit a more positive sentiment in their tweets. Even with state fixed effects, these effects are significant at the 99% confidence level. Sentiment patterns in our sample of state legislator tweets run contrary to the patterns observed in congressional campaigns by Evans et al. (2014) and Evans and Clark (2016); when tweeting from statehouses, it appears, female legislators strike a more positive tone than their male counterparts.

¹¹We calculate this score by multiplying a legislator’s negative tweets by negative one, neutral tweets by zero, and positive tweets by one, and taking the sum. A higher score denotes a higher frequency of positive tweets relative to negative tweets.

Table 5 also reports our tests of whether female legislators are more likely to tweet about education and health care (i.e., *Hypothesis 4*). These are policy realms in which they may have a great interest and expertise, and where their opportunities to claim credit for their work may be magnified if voters view them as more expert in these areas. To be sure, tweeting about these issues or indeed any specific policy realm is a rare occurrence for legislators of either gender. Over 66% of state legislative tweets have no clearly identifiable policy content, a trend that Kousser (2019) also identifies among members of the Australian Parliament. Still, the rate of tweeting about education or health care does significantly vary by gender in the states. Men address these issues in 3.5% of their tweets, while women do so in 4.4% of tweets. Controlling for party and for state fixed effects, our multivariate model estimates that female lawmakers tweet about women's issue 0.7 percentage points more often, a different that is significant at the 99% confidence level. Our models also show that this is because they tweet more about each women's issue significantly more often. Women tweet about education 0.3 percentage points more often and about health care 0.4 percentage points more often than men.

Our final tests look at whether women communicate differently than men about their ideology. Because they may seek to counter the stereotype that they are more liberal than their co-partisans (see Koch 2000), women of both parties may take more conservative positions on Twitter than men (*Hypothesis 5*). Because we are interested in how the present themselves *relative* to their roll call positions, we control for their roll call-based ideology in these tests.

In our test of *Hypothesis 6*, we explore the possibility that these incentives may operate differently for Republican and Democratic women. Republican women should have consistent incentives to take positions on social media that are more conservative, because this will position them well for Republican primary elections. Democrats face a countervailing incentive to appear more liberal to improve their chances in the primary election, which may push take them to tweet more liberally than expected given their roll call behavior.

The first model in Table 6 shows no apparent relationship between gender and Twitter ideology when we hold party affiliation and roll call-based ideology constant. As Figure 1 already showed, there is a positive correlation between legislators' twitter ideology and their roll call-based ideal points. Table 6 confirms this relationship with the positive coefficient on the roll call-based measure for legislators' ideal points. Significantly, when we control for their roll call record, party and gender do not predict the ideological content of their tweets. When we estimate separate models for each party (see columns 2 and 3), we also see no relationship between gender and the ideological content of their tweets. In sum, once we control for the legislators' ideal point, gender does not predict the ideological content of tweets. Women are not portraying themselves more conservatively in their tweets than they are in their roll call votes.

Race, Ethnicity, and Intersectionality

Recent studies suggest the importance of taking a broader view of identity. Looking at tweet activity by members of the Congressional Black Caucus during the 2013–2024 session, Tillery (2019) finds that gender was the single most powerful predictor of how often a caucus member tweeted about racial issues, with women tweeting

Table 6. Does gender predict Twitter ideology?

	All legislators	Only democrats	Only republicans
Roll call ideology	0.039** (0.007)	0.025 (0.013)	0.047** (0.010)
Female legislator	-0.009 (0.007)	-0.017 (0.010)	0.003 (0.009)
Democratic legislator	-0.012 (0.016)		
State fixed effects	Included	Included	Included
Observations	1722	782	937
Adjusted <i>R</i> -squared	0.216	0.063	0.149

Notes: Observations are all state lower house legislators elected before 2016 with Twitter accounts. Standard errors in parentheses, ** $p < 0.01$, * $p < 0.05$. Dependent variable in all models is average ideology of each legislator's tweets.

significantly more frequently about race. Barrett's (1995) investigation demonstrates that the policy priorities of Black women in statehouses are shaped by both aspects of their identity, and Fraga et al. (2006) find significant differences between male and female Latinx state lawmakers in the coalitions that they form and how often members of other groups seek their expertise.

Using data collected by the Reflective Democracy Campaign (2017) on the race and ethnicity of state legislators in the 2016–2017 session, we were able to record the race and ethnicity of 3,355 members of our full dataset of 5,422 state lawmakers, including 2,077 of the 3,144 lawmakers with Twitter handles. For this exploratory analysis, we initially analyzed members of each racial and ethnic group individually and found that Black and Latinx legislators were distinct from other legislators. To streamline the analysis and preserve statistical power in the analysis below, we combine the members of the nation's two largest racial and ethnic minority groups.

An initial, descriptive analysis reveals that Black and Latinx women in state legislatures are by far the most active group, with 82.2% having a political Twitter handle, compared with 72.3% of women who are white or members of other groups. Among male state lawmakers, 71.6% of male Black or Latinx representatives tweet, compared with 63.6% of men who are white or members of other groups. Legislators from these racial and ethnic minority groups also tweet more often than their white or other counterparts of the same gender, with Black and Latinx female lawmakers sending 1,209 versus 1,149 tweets and Black and Latinx men sending 1,344 versus 1,040. The multivariate analysis of tweet activity reported in Table 7 shows that this is intertwined with partisan differences. Democratic state legislators are much more active on social media, and Black and Latinx lawmakers are much more likely to be members of the Democratic Party. Controlling for partisanship, we do not find a significant impact of race/ethnicity or its interaction with gender on tweet activity. Yet these patterns raise the question of whether the higher levels of social media engagement by Democrats in state legislatures is partly a function of the more diverse makeup of this party.

Our analyses of the content of tweets, reported in Table 8, again reveal mixed findings but do show a significant effect of race and ethnicity on attention to health care issues as well as the persistent influence of gender. Just as we found in our main analysis, women are more likely to tweet with a positive sentiment and focus more on education and on health care. For the latter issue, Black and Latinx lawmakers are significantly more attentive than white legislators or members of other groups. The

Table 7. How do race, ethnicity, and gender affect Twitter activity?

	Does the legislator have a handle?	Tweet count	Tobit model
	0.090** (0.019)	91.402 (60.531)	248.057** (64.946)
Female legislator	−0.025 (0.031)	39.148 (106.190)	−114.067 (108.246)
Black or Latinx legislator	−0.004 (0.045)	−243.474 (151.930)	−76.890 (161.517)
Female × Black or Latinx legislator	0.063** (0.018)	351.917** (59.200)	422.495** (59.957)
Democratic legislator	Included	Included	Included
State fixed effects	3,355	2,077	3,355
Observations	0.149	0.099	
Adjusted <i>R</i> -squared			

Notes: Observations are all state lower house legislators with race/ethnicity data in the first and third models, and all state legislators with Twitter accounts in the second model. Standard errors in parentheses, ** $p < 0.01$, * $p < 0.05$. Dependent variables are a dichotomous measure of whether a legislator had an official Twitter handle (in the first column) and the count of tweets from that handle (in the second).

Table 8. How do race, ethnicity, and gender affect sentiment and attention to “women’s issues”?

	Sentiment score	Women’s issues	Education	Health care
	4.676** (1.317)	0.903** (0.190)	0.371** (0.141)	0.532** (0.126)
Female legislator	0.787 (2.066)	0.480 (0.344)	0.100 (0.274)	0.379* (0.220)
Black or Latinx legislator	0.905 (2.951)	−1.248** (0.485)	−0.167 (0.419)	−1.081** (0.268)
Female × Black or Latinx legislator	−5.348** (1.264)	0.791** (0.166)	0.232 (0.117)	0.560** (0.112)
Democratic legislator	Included	Included	Included	Included
State fixed effects	2,077	2,077	2,077	2,077
Observations	0.100	0.061	0.044	0.064
Adjusted <i>R</i> -squared				

Notes: Observations are all state lower house legislators with race/ethnicity data and Twitter accounts. Standard errors in parentheses, ** $p < 0.01$, * $p < 0.05$. Dependent variables are the percentage of tweets: (1) with a positive sentiment, (2) addressing women’s issues, (3) addressing education issues, and (4) addressing health care issues.

significant interaction between gender and race/ethnicity shows that this effect is strongest among male Black and Latinx lawmakers and demonstrates the value of taking an intersectional approach to studying Twitter behavior.

Discussion

We have looked at how state legislators use Twitter. We studied this increasingly important way to communicate because it allows us to directly learn about how legislators communicate without looking at how legislators’ efforts are filtered through media. We collected data on state legislative Twitter communication to test hypotheses in four areas.

The largest difference we observed related to the level of effort legislators put into communicating on Twitter. The data show that female legislators are more likely to have Twitter accounts and use them. Previous researchers have argued that women have to work harder in order to get elected. Consistent with that argument, prior

studies have found that women put more effort into their jobs along various dimensions (Anzia and Berry 2011; Kurtz et al. 2006; Pearson and Dancy 2011; Thomsen and Sanders 2020). We confirm that this pattern holds when looking at efforts to use Twitter to communicate with voters.

Our results also confirm previous work regarding how gender relates to the issues that politicians work on. Previous findings have shown that women work more than men on health, education, and other issues considered to be women's issues (Saint-Germain 1989; Swers 1998, 2002). Our data show that women also discuss health and education more on Twitter. Men discuss these issues in 3.5% of their tweets, while women do so in 4.4% of tweets. This means that women discuss these issues 26% more than men do ($0.9/3.5 = 0.26$).

One benefit of having more women in office is that voters learn about more issues. Politicians' communications are an important way for voters to learn about issues and to form evaluations (Arceneaux 2006). If politicians never focus on issues like health care and education, then voters are likely to pay less attention to those issues. If politicians descriptively represent the population, they are more likely to cover a wider range of issues that allows voters to learn about a wider range of issues.

Other results from our analysis contradict previous findings. Evans et al. (2014) and Evans and Clark (2016) found that among candidates for Congress, women were more likely to take a negative tone in their communication. When looking at the basic comparison between men and women we see the same pattern—i.e., women are more negative in tone. However, this is confounded with party. Women are more likely to be Democrats and Democrats are more negative in tone. When we control for partisanship, we find that, among state legislators, women are more positive than men in their Twitter communication.

Also, we find that gender no longer predicts the ideological positions that politicians take after accounting for their actual position. Previous work has found that politicians had incentives to take portray themselves as more conservative in order to counter the stereotype that women are more liberal. We find no evidence for this. Controlling for the legislators' position based on their roll call votes, gender does not predict how legislators portray themselves on Twitter. There is no evidence that women politicians are trying to counter gender stereotypes in their communication.

Future research could explore the determinants in legislators' tone in many ways. Among other things we can think more about how majority status could affect the level of negativity. Politicians in the majority might have more reasons to be more positive. We may have found that Democrats were more negative simply because of their status in the chamber. Alternatively, it might simply have been that our dataset covers the beginning of the Trump presidency (October 2015 through July 2018). Democratic politicians may have simply been responding to Trump and this could have led to a more negative tone.

More generally, research might test whether these patterns will hold in future time periods. Again, our data come from the period when Donald Trump transformed political communication by making Twitter his central means of reaching voters and attracting media attention (Kreis 2017; Ott 2017). It is also a time in which Trump was the center of attention, especially on Twitter and his sexist behavior may have influenced how women and men legislators used this communication tool (Scotto di Carlo 2020).

Social media is an increasingly important tool for legislators to use to communicate with voters. It also provides a fruitful opportunity for researchers to learn more

about representation because the data are public and the legislators are in direct control of the content. As a result, we can directly observe what politicians want to communicate with voters. We have used this tool to study how gender relates to legislators' use of this communication form. Future work will expand this research in many and varied directions.

Supplementary Materials. To view supplementary material for this article, please visit <http://doi.org/10.1017/spq.2022.16>.

Data Availability Statement. Replication materials are available on SPPQ Dataverse at <https://doi.org/10.15139/S3/MHAAZV> (Oklobdzija, Kousser, and Butler 2022).

Acknowledgments. We thank Calvin Tam, Lauryn Johnson, Mohammed Al Elew, Brittany Van Ryder, Arsham Aliaskari, Eamon Aalipour, Chris Trinh, Treya Parikh, Adarsh Parthasarathy, Emily Yeh, Zaire Bailey, Kexin Chen, and Rachel Schoner for their research assistance.

Funding Statement. The authors received no financial support for the research, authorship, and/or publication of this article.

Conflict of Interest. The authors declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

References

- Adler, E. Scott, and John Wilkerson. 2014. Congressional Bills Project: 2014. NSF 00880066 and 00880061. The views expressed are those of the authors and not the National Science Foundation.
- Alles, Santiago, and Mark P. Jones. 2016. "Tweeting in the Lone Star State: Locating State Politicians in a Common Space Using Twitter Followers." State Politics and Policy Conference.
- Anzia, Sarah F., and Christopher R. Berry. 2011. "The Jackie (and Jill) Robinson Effect: Why Do Congresswomen Outperform Congressmen?" *American Journal of Political Science* 55 (3): 478–93.
- Arceneaux, Kevin. 2006. "Do Campaigns Help Voters Learn? A Cross-National Analysis." *British Journal of Political Science*, 36 (1), 159–173.
- Baitinger, Gail. 2015. "Meet the Press or Meet the Men? Examining Women's Presence in American News Media." *Political Research Quarterly* 68 (3): 579–92.
- Barrett, Edith J. 1995. "The Policy Priorities of African American Women in State Legislatures." *Legislative Studies Quarterly* 20: 223–47.
- Berkman, Micahel B., and Robert E. O'Connor. 1993. "Do Women Legislators Matter? Female Legislators and State Abortion Policy." *American Politics Quarterly* 21 (1):102–124.
- Brady, David W., Hahrie Han, and Jeremy C. Pope. 2007. "Primary Elections and Candidate Ideology: Out of Step with the Primary. Electorate?" *Legislative Studies Quarterly* 32 (1): 79–105.
- Brooks, Deborah Jordan. 2013. *He Runs, She Runs: Why Gender Stereotypes Do Not Harm Women Candidates*. Princeton, NJ: Princeton University Press.
- Burrell, Barbara C. 1994. *A Woman's Place Is in the House: Campaigning for Congress in the Feminist Era*. Ann Arbor: University of Michigan Press.
- Butler, Daniel M. 2014. *Representing the Advantaged: How Politicians Reinforce Inequality*. New York: Cambridge University Press.
- Butler, Daniel M., Elin Naurin, and Patrik Öhberg. 2020. "Constituents Ask Female Legislators to Do More." Paper Presented at the 2020 APSA Annual Meeting.
- Considine, Mark, and Iva Ellen Deutchman. 1994. "The Gendering of Political Institutions: A Comparison of American and Australian State Legislators." *Social Science Quarterly* 75 (4): 854–866.
- Evans, Heather K., Victoria Cordova, and Savannah Sipole. 2014. "Twitter-Style: An Analysis of How House Candidates Used Twitter in their 2012 Campaigns" *PS: Political Science and Politics* 47 (2): 454–462.
- Evans, Heather K., and Jennifer Hayes Clark. 2016. "'You Tweet Like a Girl!': How Female Candidates Campaign on Twitter." *American Politics Research* 44 (2): 326–52.

- Fenno, Jr., Richard F. 1977. "U.S. House Members in Their Constituencies: An Exploration." *The American Political Science Review* 71 (3): 883–917.
- Fox, Richard L. 2006. "Congressional Elections: Where are We on the Road to Gender Parity?" In *Gender and Elections*, eds. Susan J. Carroll and Richard L. Fox, 97–116. New York: Cambridge University Press.
- Foerstel, Karen, and Herbert N. Foerstel. 1996. *Climbing the Hill: Gender Conflict in Congress*. Westport, CT: Praeger.
- Fulton, Sarah A. 2012. "Running Backwards and in High Heels: The Gendered Quality Gap and Incumbent Electoral Success." *Political Research Quarterly* 65 (2): 303–14.
- Fulton, Sarah A. 2014. "When Gender Matters: Macro-dynamics and Micro-mechanisms." *Political Behavior* 36 (3): 605–30.
- Fraga, Luis Ricardo, Linda Lopez, Valerie Martinez-Ebers, and Ricardo Ramírez. 2006. "Gender and Ethnicity: Patterns of Electoral Success and Legislative Advocacy Among Latina and Latino State Officials in Four States." *Journal of Women, Politics & Policy* 28 (3–4): 121–145.
- Fridkin, Kim L., and Patrick J. Kenney. 2009. "The Role of Gender Stereotypes in U.S. Senate Campaigns" *Politics and Gender* 5: 301–324.
- Garofoli, Joe. 2018. "After Trump, Politics is All About Social Media," *San Francisco Chronicle*, July 31, 2018.
- Grimmer, Justin. 2013. "Appropriators Not Position Takers: The Distorting Effects of Electoral Incentives on Congressional Representation." *American Journal of Political Science* 57 (3): 624–42.
- Hawkesworth, Mary. 2003. "Congressional Enactments of Race-Gender: Toward a Theory of Raced-Gendered Institutions." *American Political Science Review* 97 (4): 529–50.
- Hayes, Danny, and Jennifer L. Lawless. 2016. *Women on the Run: Gender, Media, and Political Campaigns in a Polarized Era*. New York: Cambridge University Press.
- Heath, Roseanna Michelle, Leslie A. Schwindt-Bayer and Michelle M. Taylor-Robinson. 2005. "Women on the Sidelines: Women's Representation on Committees in Latin American Legislatures." *American Journal of Political Science* 49 (2): 420–436.
- Heldman, Caroline, Susan J. Carroll, and Stephanie Olson. 2005. "'She Brought Only a Skirt': Print Media Coverage of Elizabeth Dole's Bid for the Republican Presidential Nomination." *Political Communication* 22 (3): 315–35.
- Huddy, Leonie, and Nayda Terkildsen. 1993. "Gender Stereotypes and the Perception of Male and Female Candidates." *American Journal of Political Science* 37 (1): 119–147.
- Jungherr, Andreas. 2016. "Twitter Use in Election Campaigns: A Systematic Literature Review." *Journal of Information Technology & Politics* 13 (1): 72–91.
- Kahn, Kim Fridkin. 1992. "Does Being Male Help? An Investigation of the Effects of Candidate Gender and Campaign Coverage on Evaluations of U.S. Senate Candidates." *The Journal of Politics* 54 (2): 497–517.
- Kahn, Kim Fridkin. 1994. "Does Gender Make a Difference? An Experimental Examination of Sex Stereotypes and Press Patterns in Statewide Campaigns." *American Journal of Political Science* 38 (1): 162–95.
- Kahn, Kim Fridkin. 1996. *The Political Consequences of Being a Woman: How Stereotypes Influence the Conduct and Consequences of Political Campaigns*. New York: Columbia University Press.
- Koch, Jeffrey. 2000. "Do Citizens Apply Gender Stereotypes to Infer Candidates' Ideological Orientations?" *Journal of Politics* 62(2): 414–29.
- Kousser, Thad. 2019. "Tweet Style: Campaigning, Governing, and Social Media in Australia." *Australian Journal of Political Science* 54: 183–201; doi:10.1080/10361146.2019.1568966.
- Kousser, Thad, and Stan Oklobdzija. 2018. "The Supply and Demand of Fact v. Opinion in Presidential Tweets." Paper presented at the 2018 Meetings of the American Political Science Association.
- Kreis, Ramona. 2017. "The 'tweet politics' of President Trump." *Journal of Language and Politics* 16 (4): 607–18.
- Krook, Mona Lena. 2020. *Violence Against Women in Politics*. New York: Oxford University Press.
- Kurtz, Karl T., Gary Moncrief, Richard G. Niemi, and Lynda W. Powell. 2006. "Full-Time, Part-Time, and Real Time: Explaining State Legislators' Perceptions of Time on the Job." *State Politics & Policy Quarterly* 6 (3): 322–38.
- Landis, J. R., and Gary G. Koch. 1977. "The measurement of observer agreement for categorical data." *Biometrics* 33 (1): 159–174.
- Lau, Richard R. and Gerry Pomper. 2004. *Negative Campaigning: An Analysis of U.S. Senate Elections*. Lanham, MD: Rowman Littlefield.
- Lawless, Jennifer L. 2015. "Female Candidates and Legislators." *Annual Review of Political Science* 18: 349–66.

- Lazarus, Jeffery, and Amy Steigerwalt. 2018. *Gendered Vulnerability: How Women Work Harder to Stay in Office*. Ann Arbor, MI: University of Michigan Press.
- McGregor, Shannon C., Regina G. Lawrence, and Arielle Cardona. 2016 "Personalization, Gender, and Social Media: Gubernatorial Candidates' Social Media Strategies." *Information, Communication and Society* 20 (2): 264–83.
- Newman, Jody. 1994. *Perception and Reality: A Study Comparing the Success of Men and Women Candidates*. Washington, DC: National Women's Political Caucus.
- Norton, Noelle H. 1999. "Committee Influence Over Controversial Policy: The Reproductive Policy Case." *Policy Studies Journal* 27 (2): 203–216.
- Oklobdzija, Stan, Thad Kousser, and Daniel Butler. 2022. "Replication Data for: Do Male and Female Legislators Have Different Twitter Communication Styles?" *UNC Dataverse*, V1, UNF:6:CZoFF/yYo4C7HlTRja/mmA== [fileUNF]. <https://doi.org/10.15139/S3/MHAAZV>.
- Osborn, Tracy. 2012. *How Women Represent Women*. New York: Oxford University Press.
- Ott, Brian L. 2017. "The Age of Twitter: Donald J. Trump and the Politics of Debasement." *Critical Studies in Media Communication* 34 (1): 59–68.
- Parmelee, John H., and Shannon L. Bichard. 2012. *Politics and the Twitter Revolution: How Tweets Influence the Relationship Between Political Leaders and the Public*. Lanham, MD: Lexington Books.
- Pearson, Kathryn, and Logan Dancy. 2011. "Elevating Women's Voices in Congress: Speech Participation in the House of Representatives." *Political Research Quarterly* 64 (4): 910–23.
- Pearson, Kathryn, and Eric McGhee. 2013. "What it Takes to Win: Questioning 'Gender Neutral' Outcomes in US House Elections." *Politics & Gender* 9 (4): 439–62.
- Peterson, Andrew, and Arthur J. Spirling. 2018. "Classification Accuracy as a Substantive Quantity of Interest: Measuring Polarization in Westminster Systems." *Political Analysis* 26 (1): 120–128.
- Reflective Democracy Campaign. 2017. *Reflective Democracy Research Findings 2016–2017*. Accessed at <https://wholeads.us/research/reflective-democracy-research-findings-2016-2017/> in July 2021.
- Reingold, Beth. 1992. "Concepts of Representation among Female and Male State Legislators." *Legislative Studies Quarterly* 17 (4): 509–537.
- Reingold, Beth. 2000. *Representing Women: Sex, Gender, and Legislative Behavior in Arizona and California*. Chapel Hill, NC: University of North Carolina Press.
- Rosenstiel, Tom, Jeff Sondermon, Kevin Locker, Maria Ivancin, and Nina Kjarval. 2015. *Twitter and the News: How People Use the Social Network to Learn About the World*. Arlington, VA: American Press Institute.
- Saint-Germain, Michelle A. 1989. "Does Their Difference Make a Difference? The Impact of Women on Public Policy in the Arizona Legislature." *Social Science Quarterly* 70 (4): 956–970.
- Scott di Carlo, Giuseppina. 2020. "Trumping twitter: Sexism in President Trump's Tweets." *Journal of Language and Politics* 19 (1): 48–70.
- Seltzer, Richard A., Jody Newman, and M. Voorhees Leighton. 1997. *Sex as a Political Variable*. Boulder, CO: Lynne Reinner.
- Shor, Boris, and Nolan McCarty. 2011. "The Ideological Mapping of American Legislatures." *American Political Science Review* 105 (3): 530–51.
- Sides, John, Christopher Tausanovitch, Lynn Vavreck, and Christopher Warshaw. 2020. "On the Representativeness of Primary Electorates." *British Journal of Political Science* 50 (2): 677–685.
- Squire, Peverill. 2017. "A Squire Index Update." *State Politics and Policy Quarterly* 17 (4): 361–71.
- Swers, Michele L. 1998. "Are Congresswomen More Likely to Vote for Women's Issue Bills Than Their Male Colleagues?" *Legislative Studies Quarterly* 23 (3): 435–448.
- Swers, Michele L. 2002. *The Difference Women Make: The Policy of Women in Congress*. Chicago: University of Chicago Press.
- Tamerius, Karin L. 1995. "Sex, Gender and Leadership in the Representation of Women." In *Gender Power, Leadership, and Governance*, eds. Georgia Duerst-Lahti and Rita Mae Kelly. Ann Arbor, MI: University of Michigan Press.
- Thomsen, Danielle M. 2020. Ideology and Gender in U.S. House Elections. *Political Behavior* 42: 415–42.
- Thomsen, Danielle M., and Bailey K. Sanders. 2020. Gender Differences in Legislator Responsiveness. *Perspectives on Politics* 18: 1017–30.
- Thomas, Sue. 1994. *How Women Legislate*. New York: Oxford University Press.

- Thomas, Sue, and Susan Welch. 1991. "The Impact of Gender on Activities and Priorities of State Legislators." *Western Political Quarterly* 44: 445–56.
- Tillery, Alvin B. 2019. "Tweeting Racial Representation: How the Congressional Black Caucus Used Twitter in the 113th Congress." *Politics, Groups, and Identities* 9 (2): 219–38.
- Vergeer, Maurice. 2015. "Twitter and Political Campaigning." *Sociology Compass* 9 (9):745–60.
- Volden, Craig, Alan E. Wiseman, and Dana E. Wittmer. 2013. "When are Women More Effective Lawmakers Than Men?" *American Journal of Political Science* 57: 326–41.

Author Biography. Dan Butler is a Professor of Political Science at the Washington University in St. Louis. His research focuses on understanding representation.

Thad Kousser is a Professor of Political Science at the University of California, San Diego and co-directs the Yankelovich Center for Social Science Research.

Stan Oklobdzija is a Visiting Assistant Professor of Public Policy at the University of California, Riverside where his research focuses on state/local politics, housing policy, and interest group behavior. He holds a PhD in Political Science from the University of California, San Diego and his prior research has appeared in Electoral Studies, Research & Politics, Socius, and State Politics & Policy Quarterly.

Cite this article: Butler, Daniel M., Thad Kousser, and Stan Oklobdzija. 2023. Do Male and Female Legislators Have Different Twitter Communication Styles? *State Politics & Policy Quarterly* 1–23, doi:10.1017/spq.2022.16