

Modalities of Political Engagement on Social Media and Voting for Donald Trump: Capturing Heterogeneity by Race over Time

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The widespread adoption of social media has radically altered the American political communication environment, transforming how citizens access information, express political opinions, and engage with candidates and campaigns. However, the relationship between modes of political engagement on social media and political actions, such as voting behavior, remains understudied. This study provides a systematic, temporal analysis of the relationships between distinct modalities of political engagement on social media (e.g. liking, following, forwarding, commenting on and posting political content) and support for Donald Trump across three election cycles — 2016, 2020, and 2024. Using self reported measures of social media use from the Cooperative Election Study (CES), we show that among social media users, politically engaged voters were significantly more likely to vote for Trump in all three elections. Exploring racial heterogeneity, we also show that in 2024, the difference in support for Trump between politically engaged and unengaged social media users was the largest among Asian Americans. Moreover, politically engaged Hispanic voters went from being less likely to vote for Trump in 2016 to being significantly more likely to vote for him in 2020. Our findings reveal variation by race and modes of political engagement. We found that social media use itself is negatively associated with voting for Trump, but political engagement on such platforms is positively tied to it.

Significance Statement

Second of three dissertation papers.
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Social Media | Political Engagement | Voting Behavior | Temporal Analysis

The upset election victory of Donald Trump in 2016 and the concurrent rise of social media platforms as modes of political campaigning and engagement have fundamentally reshaped the American political landscape. These parallel phenomena are often viewed as deeply intertwined, with platforms like Twitter and Facebook seen as instrumental to Trump's electoral success. Digital media platforms changed how citizens encounter political information(1, 2), how populist candidates like Mr. Trump built trust(3) among voters, and how they used social media for agenda setting in the mainstream press(4). This transformation has spurred an extensive line of research investigating the link between digital engagement and real-world political behaviors, including both turnout and vote choice. However, while it is widely believed that social media played a crucial role in the 2016 and 2020 elections, the precise nature of this relationship—how specific online actions translate into offline ballot-box decisions—remains underexplored.

A foundational premise in political science is that engagement with political information and discourse is predictive of greater participation (5). Social media platforms dramatically lower the costs of such engagement(6, 7), creating a stream of political content delivered directly into users' personal networks. This environment facilitates not just passive consumption but also a spectrum of interactive behaviors, from low-cost "likes" to more active sharing and commenting to posting original political content. Early research in the social media era demonstrated the potential of these platforms to mobilize citizens, with large-scale experiments showing that social cues on Facebook could tangibly increase real-world voting (8).

Past research has also reported on the relationship between engagement modalities like liking, forwarding, commenting, and posting political content and offline political behavior. The easiest form of engagement is reading or viewing content, which has been reported to have had only a weak association with voting, if any(9, 10). The next level of engagement involves liking or commenting on political content, and is known to have a relatively stronger association(7, 8) with political behavior offline compared to merely seeing content.

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Moving further in this hierarchy of engagement levels, we see users who share, forward, or repost content, thereby amplifying its reach. Researchers(7, 11?) have identified several mechanisms by which such activities shape offline political outcomes, such as gateway effects, echo chambers, and increased affective polarization. Finally, the highest tier of engagement on social media platforms is to post original political content(9), and such users are already highly politically engaged in the offline space(?) .

Yet, as the digital ecosystem has matured and become more polarized, the optimistic view of social media as a purely mobilizing force has been complicated. Scholars now grapple with whether these platforms primarily activate existing partisans or if they can persuade and mobilize a wider, less-engaged segment of the population (7). The role of social media-driven misinformation(12) in helping populist candidates bypass filters of the moderated mainstream press, and use emotive appeals(?) to aggravate affective polarization, have invited critical evaluations.

This debate is particularly salient in the context of Donald Trump, whose candidacy and first presidency were characterized by a unique use of social media to bypass traditional media gatekeepers and build a powerful, direct channel to his supporters (13). His use of platforms like Twitter was not just a communication strategy. It became a central feature of his political brand(1, 3), presenting amateurism as authenticity (?). Past research also shows how the Trump 2016 campaign substituted traditional mobilization and campaign apparatus of the RNC with a social media-centered strategy. Eventually, Mr. Trump moved to Truth Social in 2022 after social media bans from other platforms, yet maintained a strong presence on the platform. Trump's disproportionate emphasis on social media for campaigning and mobilization makes him of particular interest to our study.

Much of the past research on social media's impact on politics has focused on the dynamics of a single election cycle. This overlooks long term temporal changes in relationships between social media activities and offline political activities. Moreover, different modes of engagement—passively reading content versus actively reacting to or amplifying it—may have different relationships with political behavior behavior (6?). But no prior study, to our knowledge, captures this relationship with electoral support for Mr. Trump. Furthermore, prior work has shown that media effects on support for Trump are heterogeneous across racial and ethnic identities, as well as by gender, immigrant background, and educational attainment. These findings, however, also show yet another limitation of extant literature — they rarely include accurate information about race or ethnicity and therefore obscure analysis of racial heterogeneity in social media effects on voting behavior. Moreover, the use of non-probability convenience samples, often drawn from among social media users limits the generalizability of their findings.

This study addresses these critical gaps by (1) quantifying the relationship between distinct modalities of political engagement on social media and voting for Donald Trump, (2) measuring the temporal change in this relationship across three election cycles (2016-2024), and (3) revealing racial heterogeneity in this relationship and its temporal evolution. Using nationally representative pooled cross-sectional data from the Cooperative Election Study (CES) for the 2016,

2020 and 2024 US presidential elections, we investigate whether a durable association exists between Americans' engagement with political material on social media and who they vote for. We move beyond generic measures of social media use to examine specific engagement behaviors such as liking, following, forwarding, commenting, and posting political content. Our central hypothesis is that engaging with political content on social media is positively associated with the likelihood of voting for Trump. We then probe this relationship further through a key research question: How does this association vary across racial identities? By analyzing online behaviors and voter demographics across three contentious election cycles, this paper enhances our understanding of social media's role in contemporary American politics. In doing so, we provide critical insights into the dynamics of support for populists like Trump in an increasingly fragmented and digitally mediated American public sphere.

Background

Researchers have scrutinized the relationship between social media and the rise Donald Trump in the US. In this section, we enumerate studies that link political engagement with political behavior, how social media-based engagement may alter such behavior, and the key contributions and limitations of the prevailing literature.

Social Media and Donald Trump. Past work(14, 15) has documented how social media facilitated the rise of populism around the world, including the victory of Mr. Trump in the US. Several mechanisms have been proposed to explain why the affordances of social media platforms aid populist candidates like him. In particular, platforms like Twitter turned amateurism and authenticity into political currency in ways that elite-moderated media platforms did not (1). This helped candidates like Trump build trust among voters(3), easing his path to electoral viability(1, 3). Trump used Twitter effectively to build his own brand as an anti-establishment candidate(16).

Other studies have considered the agenda-setting effect of social media, whereby Trump effectively used Twitter to divert media discourse away from issues that are presumed to be disadvantageous to him toward more favorable issues (4, 17?). For instance, Lewandowsky et al. (2020) find that Trump increased tweeting on unrelated or advantageous issues following the release of negative news about the Russia investigation, and that this surge was followed by a measurable decline in subsequent coverage of the unfavorable story by major outlets such as The New York Times and ABC News (17). Likewise, Chen et al. (2022) show that increases in adverse pandemic indicators coincided with Trump putting greater emphasis on favorable themes such as the economy and jobs, effectively shifting public and media discourse (4). Together, these findings indicate that Trump effectively used Twitter as an agenda-setting tool to shift media and public discourse toward issues more advantageous to his political standing.

Finally, prior research has examined how presidential candidates have used social media to mobilize voters, often contrasting Barack Obama's integration of digital tools with traditional field operations against Donald Trump's far more

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platform-centric strategy. For example, Mork (2020) shows that whereas Obama's campaigns used social media primarily to complement offline volunteer networks, Trump's 2016 campaign relied heavily on platforms such as Facebook and Twitter as the primary infrastructure for mobilization. Trump's campaign emphasized follower growth, rapid content amplification, and platform-native calls to action, generating high engagement despite comparatively weaker on-the-ground organization. This evidence highlights the dominance of social media platforms in Trump's mobilization efforts (18).

Political Engagement and Political Behavior. A foundational premise in the study of political behavior is that political engagement—whether measured as information seeking, expressive activity, or interaction—tends to be predictive of political participation, particularly voting(8). Classic mobilization theories posit that more frequent exposure to political messages and opportunities for interaction should lower the cost of political activity and facilitate participation(9?). Social media dramatically lowers such costs (2) by delivering political content directly to users' handheld devices, making it easier than ever to view, react to, and spread political information.

Empirical research prior to the rise of social media consistently found that higher political involvement and exposure correlated positively with turnout and mobilization(?). With the growth of platforms such as Facebook, Twitter/X, and YouTube, researchers observed that political behaviors—both online and offline—might be meaningfully shaped by online engagement. A landmark experimental study with a sample of over 60 million users on Facebook found that exposure to social mobilization messages, including seeing others' engagement (e.g., "I Voted" badges, friends' voting), increased real-world voting behavior(8). Moreover, in 2016, both presidential candidates – Trump and Hillary Clinton – used social media to mobilize voters by convincing co-partisans of their likelihood of winning(19). Given these two findings—that Mr. Trump benefited from campaigning on social media platforms and that political engagement on social media has tangible impact on voting behavior — we arrive at the following hypothesis:

Hypothesis 1: *Social media users who engaged with political content are more likely to vote for Trump than those who did not engage with political content.*

We also include the following research question in our study to examine racial heterogeneity in this relationship.

RQ1: *How does the relationship between voting for Trump and political engagement on social media vary by race?*

Subsequent studies have explored the nuances of political engagement. Researchers began to ask whether shallow behaviors (viewing or reading content), reactive behaviors (liking or reacting), or amplifying behaviors (sharing, reposting, posting) were differentially related to political attitudes or action. Critically, while these activities are more common among more politically interested users, they also channel social reinforcement, feedback, and algorithmic amplification. Such processes can also intensify mobilization on one hand and political polarization on the other(20?).

The next subsection enumerates past studies that considered these distinct modalities of political engagement and their relationships with offline political behavior.

Engagement Modalities and Political Behavior. The most common and basic form of political engagement on social media is viewing or reading political content. Unlike offline environments, major social media platforms expose individuals to political information not necessarily by deliberate choice, but as a byproduct of algorithmic feeds, peer activity, and targeted advertising(21). Social media users do not need to explicitly seek political information to encounter it in their feeds. Past research that links online engagement with offline political behavior has largely yielded null results, though. While observational research repeatedly finds that individuals who are more exposed to political content online are also more likely to report higher political knowledge and turnout(9), the direction of causality is challenging to establish because politically motivated users may seek out such content. Indeed, studies that used randomized controlled trials (RCTs) to manipulate content viewed by users on Facebook and Instagram did not produce immediate changes in their political attitudes(10, 22). Comparable experiments that altered partisan lean of content fed to users on YouTube did not yield significant changes in political attitudes or political participation in the short run(23?). While these works illustrate the lack of any immediate effect, our study is focused on long term changes in voting behavior over a period of eight years.

The next tier of engagement covers likes, reactions, and comments, typically considered weaker forms of engagement than sharing or posting, but more active than passive viewing. These actions provide immediate social signals to algorithms and networks, often shaping what is amplified or made visible, and reinforcing in-group/out-group bonds(24). Compared to reading or viewing content, engagement through liking or commenting has shown modest links with offline political participation. For instance, Bode and colleagues(?) find that users who 'like' or comment on political tweets report greater interest and are more active in sharing and participation. In the Facebook field experiment by Bond et al.(8), simply seeing friends' likes or comments on voting mobilization content significantly increased the likelihood of turnout, showing that social cues embedded in such weak-tie actions can have real effects. Other studies(?) that attempted to link reactive engagement with offline participation have yielded null results, suggesting the need for nuance in proclaiming reactive social media engagement as a predictor of voting behavior.

Sharing, forwarding, and reposting are more active amplification behaviors—turning audiences into broadcasters, not just consumers. Experimental and observational evidence provides a more consistent relationship between active sharing and political action, especially in highly salient moments like elections. Mosleh et al.(?) demonstrated experimentally that content designed to trigger emotion is more widely shared and can shape beliefs about political events, suggesting a pathway from content characteristics to behavior via social transmission. Moreover, users who share political content are generally more likely to participate in politics beyond social media, including voting(20?). However, here too, we see evidence for caution. Whereas large field experiments that sought to manipulate sharing features on Facebook found

373 no measurable change on individual attitudes, others have
374 reported short term effects vis-a-vis affective polarization(25)?
375 ?). Therefore, studies of the relationship between sharing
376 or reposting content and voting behavior, thus, remain
377 inconclusive, especially when we consider longer time periods.
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379 Perhaps the highest level of social media engagement
380 occurs when users create and post original political content.
381 Posters serve as both opinion leaders and content sources for
382 their networks, a crucial node in the online information ecosystem.
383 Research has consistently found that individuals who
384 post political content are among the most politically active
385 and expressive segment of users, both online and offline (26).
386 They are disproportionately engaged in offline participation,
387 campaign volunteering, and voting(9). However, this group is
388 unrepresentative as they are more partisan, ideological, older,
389 and more civic-minded than typical social media users.

390 These findings motivate our second research question:

391 **RQ2A:** *Which modes of political engagement have a
392 significant relationship with the probability of voting
393 for Trump?*

394 **Political Engagement and Long-term Change in Political
395 Behavior.** A particularly important question for this study
396 is whether long term exposure, engagement, or posting is
397 associated with voting behavior. This is especially important
398 in the context of studying support for Mr. Trump, who was
399 on the presidential ballot for three consecutive years.

400 Literature on long-term effects offer a mixed picture.
401 Panel data studies extending from several months to a year
402 observed stability in attitudes, with only marginal evidence
403 for cumulative increases in polarization or shifts in candidate
404 support(27). On the other hand, experiments that exposed
405 users to content from ideological opponents for a month
406 found that polarization increased after the intervention(28).
407 Observational studies leveraging Facebook ad data found that
408 targeted microtargeting campaigns may have tilted undecided
409 voters toward Trump and increased Republican turnout(29).
410 A quasi-experimental county-level study found that higher
411 Twitter penetration was associated with lower Republican
412 (and thus Trump) vote share in 2016 and 2020, suggesting
413 broader platform context, rather than simply engagement,
414 may matter most(?).

415 These findings motivate our third research question which
416 considers temporal change in the relationship between modes
417 of political engagement and support for Trump:

418 **RQ2B:** *How does the relationship between each
419 mode of political engagement and voting for Trump
420 change over time?*

421 **Limitations of Current Work.** Our literature review revealed
422 substantial results that link political engagement with offline
423 political behavior, especially voting behavior. Here, we list
424 the gaps in current literature. First, most studies track short
425 term changes. While some extend their study period to a
426 year, we did not find any study that tracks changes in voting
427 behavior with respect to political engagement on social media
428 over multiple election cycles. In this study, we seek to remedy
429 that by exploring this relationship over a period of eight years
430 (three US presidential election cycles).

431 Second, most studies used non-probability samples in their
432 analyses. A significant limitation is that most platform field

433 experiments recruit active users of the platform in question
434 and thus do not deploy nationally representative samples.
435 This creates challenges in terms of the representativeness of
436 the study populations. Noteworthy exceptions include na-
437 tionally representative panel studies linked to web traces(30)
438 and large surveys drawing from representative online panels
439 (?). However, the overwhelming majority of studies rely on
440 platform-based convenience samples. By using a nationally
441 representative dataset – the Cooperative Election Study – we
442 are attempting to address this limitation.

443 Third, reporting on race and ethnicity is rare in platform-
444 based studies, as such information may be hard to infer
445 from online profiles. However, there is ample evidence that
446 patterns of preferences in social media consumption vary
447 widely across racial and ethnic boundaries. Consequently,
448 our study emphasizes on racial heterogeneity in analysing
449 the relationship between political engagement and voting
450 behavior.

451 Finally, virtually all major studies in this domain focus on
452 the largest U.S. platforms—Facebook, Instagram, Twitter/X,
453 and YouTube. Some research has also examined cross-
454 platform effects or compared engagement effects across
455 networks, but the bulk of longitudinal analyses are platform-
456 specific(10, 22?, 23). In this case, we are unable to address
457 this lack of platform diversity in this study. The self-reported
458 measures on social media use in the CES surveys do not
459 query platform specific use, but do mention platforms like
460 Facebook, Twitter, and YouTube by name in their questions.

461 **Data & Methods**

462 In this section, we present the datasets employed in our
463 analysis along with the data manipulations we performed for
464 our analysis. Additionally, we detail the model specifications
465 for the two logistic regression models utilized in our study,
466 and the rationale for the same.

467 We used data from the Cooperative Election Study for
468 three consecutive presidential elections – 2016, 2020, and
469 2024. The Cooperative Election Study (CES), formerly
470 the Cooperative Congressional Election Study (CCES), is
471 the largest academic survey of U.S. elections, conducted
472 biennially since 2006. In 2016, the CCES surveyed about
473 64,600 respondents recruited by YouGov through matched
474 random sampling from the American Community Survey
475 (ACS). Interviews occurred in two waves (Sept. 28–Nov. 7
476 and Nov. 9–Dec. 14, 2016). In 2020, it was renamed as
477 the CES and surveyed 61,000 adults. The pre-election wave
478 ran Sept. 29–Nov. 2, and the post-election wave from Nov.
479 8–Dec. 14, 2020. Sampling again relied on ACS and voter
480 records, with YouGov constructing matched random samples.
481 Vote validation, released in August 2021, confirmed turnout
482 against state files, producing one of the most reliable election
483 datasets. Response rates ranged from 61–69% depending on
484 calculation. In 2024, CES surveyed 60,000 respondents across
485 60 teams. The pre-election wave ran Oct. 1–Nov. 4, and the
486 post-election wave from Nov. 6–Dec. 10, 2024. State-level
487 samples were large enough for precise estimates, including
488 4,000+ in California, Florida, and Texas.

489 **Dependent Variable.** For our dependent variable, we consider
490 the candidate preference in presidential voting recorded in the

497 **Table 1. Number of Respondents by Modalities of Political Engagement on Social Media by Race and Year** 559
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| Year | Race/Ethnicity | Group N | Read Image | Follow Event | Forward Content | Post Comment | Post Image |
|------|----------------|---------|------------|--------------|-----------------|--------------|------------|
| 2016 | Asian | 1,517 | 1,052 | 487 | 378 | 386 | 357 |
| | Black | 5,454 | 3,666 | 1,492 | 1,450 | 1,649 | 1,561 |
| | Hispanic | 3,850 | 2,776 | 1,184 | 1,088 | 1,245 | 1,253 |
| | White | 32,470 | 21,804 | 10,131 | 9,761 | 11,410 | 9,849 |
| | Other | 2,024 | 1,460 | 712 | 713 | 825 | 785 |
| | Total | 45,315 | 30,758 | 14,006 | 13,390 | 15,515 | 13,805 |
| 2020 | Asian | 1,450 | 924 | 343 | 355 | 257 | 251 |
| | Black | 5,057 | 2,638 | 907 | 1,149 | 1,155 | 1,163 |
| | Hispanic | 4,057 | 2,308 | 875 | 1,062 | 992 | 993 |
| | White | 33,715 | 20,066 | 7,973 | 8,967 | 10,270 | 8,194 |
| | Other | 2,195 | 1,370 | 576 | 672 | 753 | 626 |
| | Total | 46,474 | 27,306 | 10,674 | 12,205 | 13,427 | 11,227 |
| 2024 | Asian | 1,663 | 1,047 | 345 | 333 | 236 | 223 |
| | Black | 6,053 | 3,498 | 1,145 | 1,339 | 1,222 | 1,244 |
| | Hispanic | 4,289 | 2,575 | 838 | 1,060 | 827 | 846 |
| | White | 33,542 | 20,169 | 6,757 | 7,852 | 7,866 | 5,869 |
| | Other | 3,019 | 1,899 | 654 | 825 | 810 | 680 |
| | Total | 48,566 | 29,188 | 9,739 | 11,409 | 10,961 | 8,862 |

511 **Table 2. Unweighted Summary Statistics of Dependent and Independent Variables by Year** 579
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| Variable | 2016 | 2020 | 2024 |
|----------------------------------|--------|--------|--------|
| Trump Vote (1=Yes) | 19,227 | 18,977 | 19,765 |
| Trump Vote (0=No) | 43,775 | 29,970 | 30,631 |
| <i>Total</i> | 63,002 | 48,947 | 50,396 |
| No Social Media (SM) | 19,285 | 14,526 | 11,434 |
| No Political Activity on SM | 7,906 | 14,121 | 14,791 |
| Politically Active on SM | 37,409 | 32,353 | 33,775 |
| <i>Total</i> | 64,600 | 61,000 | 60,000 |
| Reads Political Image (Yes) | 30,758 | 27,306 | 29,188 |
| Reads Political Image (No) | 14,557 | 19,168 | 19,378 |
| <i>Total</i> | 45,315 | 46,474 | 48,566 |
| Follows Political Event (Yes) | 14,006 | 10,674 | 9,739 |
| Follows Political Event (No) | 31,309 | 35,800 | 38,827 |
| <i>Total</i> | 45,315 | 46,474 | 48,566 |
| Forwards Political Content (Yes) | 13,390 | 12,205 | 11,409 |
| Forwards Political Content (No) | 31,925 | 34,269 | 37,157 |
| <i>Total</i> | 45,315 | 46,474 | 48,566 |
| Posts Political Comment (Yes) | 15,515 | 13,427 | 10,961 |
| Posts Political Comment (No) | 29,800 | 33,047 | 37,605 |
| <i>Total</i> | 45,315 | 46,474 | 48,566 |
| Posts Political Image (Yes) | 13,805 | 11,227 | 8,862 |
| Posts Political Image (No) | 31,510 | 35,247 | 39,704 |
| <i>Total</i> | 45,315 | 46,474 | 48,566 |

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pre-election survey by the CES in each presidential election
from 2016 to 2024.551
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Independent Variables. Our variable of interest is political engagement on social media. We construct this variable using a battery of six different questions on social media use that CES included in each of the three election cycles. First, the CES inquired if the respondent had used social media in the preceding 24 hours. Second, for respondents that answered in the affirmative to this question, the survey559
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asked them the following five mutually independent questions (see table 2 for question wording): did the respondent post a story, photo, video or link about politics?, did they post a comment about politics?, did they read story or watch a video about politics?, did they follow a political event?, and did they forward a story, photo, video or link about politics to friends?. Each of these questions captures a different form of political engagement, with different implications for their relationship with presidential voting preference. For instance, passive engagement modes like reading or watching videos or following are much more common among users compared to active engagement like liking, sharing, and posting original content. The latter is more associated with politically expressive users, some of whom are affiliated with campaigns. We encapsulate all of these in a single variable called PESM (Political Engagement on Social Media)598
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Control Variables. Past research has identified significant relationships between demographic characteristics and voting behavior in presidential elections. First, political ideology - coded as a five-level variable ranging from very liberal to very conservative - has been strongly associated with voting preference (31, 32), including voting for Donald Trump(33). Additionally, partisanship—assessed across five levels—serves as a robust predictor of vote choice, as demonstrated by long term studies of partisan voting behavior(33, 34). Third, the relationship between voting behavior and gender is statistically significant(33, 35, 36), as is the relationship with sexuality(37, 38). Fourth, we find that not only was education a significant predictor of voting behavior in 2016(39), but also that educational polarization has increased during the Trump era(35). Finally, age and income have also been identified as key factors in the voting function(33). Finally, race too plays a role in vote choice(33, 35). We included each of these variables as controls in our analysis.617
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Statistical Modeling. To validate our hypotheses regarding the relationship between political engagement on social media and voting preference for Trump, we used two logistic regression

621 **Table 3. Political Engagement on Social Media by Race & Year**

| 622 Year | 623 Race/ 624 Ethnicity | 625 Group N | 626 No 627 Social Media Use | 628 No 629 Political Engage- ment on 630 Social Media | 631 Politically 632 En- gaged 633 on Social Media |
|----------|----------------------------|-------------|--------------------------------------|--|---|
| 634 2016 | Asian | 2,278 | 761 | 248 | 1,269 |
| | Black | 7,926 | 2,472 | 813 | 4,641 |
| | Hispanic | 5,238 | 1,388 | 483 | 3,367 |
| | White | 46,289 | 13,819 | 6,073 | 26,397 |
| | Other | 2,869 | 845 | 289 | 1,735 |
| | Total | 64,600 | 19,285 | 7,906 | 37,409 |
| 635 2020 | Asian | 1,831 | 381 | 404 | 1,046 |
| | Black | 6,952 | 1,895 | 1,704 | 3,353 |
| | Hispanic | 5,180 | 1,123 | 1,247 | 2,810 |
| | White | 44,128 | 10,413 | 10,166 | 23,549 |
| | Other | 2,909 | 714 | 600 | 1,595 |
| | Total | 61,000 | 14,526 | 14,121 | 32,353 |
| 640 2024 | Asian | 1,949 | 286 | 498 | 1,165 |
| | Black | 7,728 | 1,675 | 1,856 | 4,197 |
| | Hispanic | 5,150 | 861 | 1,247 | 3,042 |
| | White | 41,443 | 7,901 | 10,399 | 23,143 |
| | Other | 3,730 | 711 | 791 | 2,228 |
| | Total | 60,000 | 11,434 | 14,791 | 33,775 |

641 models. In both models, we used the preference for Trump
642 voting as our dependent variable, and include all control
643 variables. The models, run in Stata SE 18.5, differ in one
644 regard. Model 1 (see equation 2) includes the three-level
645 political engagement variable as an independent variable
646 built as per equation 1. The second model (see equation
647 3) includes the five different social media variables as five
648 independent variables. This formulation helps answer two
649 separate questions – is there a relationship between political
650 engagement on social media and voting For trump? And if
651 so, which modality of political engagement is most strongly
652 associated with it? For the latter, we consider five different
653 modalities of political engagement - reading, liking, following,
654 forwarding, and sharing.

$$655 \text{political_engagement} = \begin{cases} 656 0, & \text{if } \text{social_media} == 0 \\ 657 1, & \text{if } \text{social_media} == 1 \\ 658 2, & \text{if } (\text{reading} \vee \text{following} \vee \\ 659 \text{commenting} \vee \text{forwarding} \\ 660 \vee \text{posting}) == 1 \end{cases} \quad [1]$$

661 It is pertinent to discuss our reasons for using two different
662 models here. First, the aggregation of different types of
663 political engagement into one variable in model 1 allows us to
664 compare politically active voters on social media with those
665 who aren't engaging with political content on these platforms
666 and those who did not use social media. Second, the survey
667 asked the questions on modalities only to those who responded
668 affirmatively to using social media, reducing the size of the
669 sample in the dataset used in model 2. Furthermore, a three-
670 way interaction between engagement modalities, race (or age
671 or gender), and year in model 2 could potentially leave the
672 study underpowered. Therefore, we exclude interactions with

673 demographic variables in model 2, and rely on model 1 for
674 our intersectional analysis by race, age, and gender.

675 The specification for model 1 is as follows:

$$676 \text{Trump_Vote} \sim \text{ideology} + \text{party_identity} + \text{education} + \\ 677 \text{foreign_born} + \text{sexuality} + \text{gender} + \\ 678 \text{family_income} + \text{age} + \text{year} + \\ 679 \text{political_engagement} + \text{race} \\ 680 \text{political_engagement} \times \text{race} \times \text{year} \\ 681 \text{political_engagement} \times \text{age} \times \text{year} \\ 682 \text{political_engagement} \times \text{gender} \times \text{year}$$

[2]

683 The specification for model 2 is as follows:

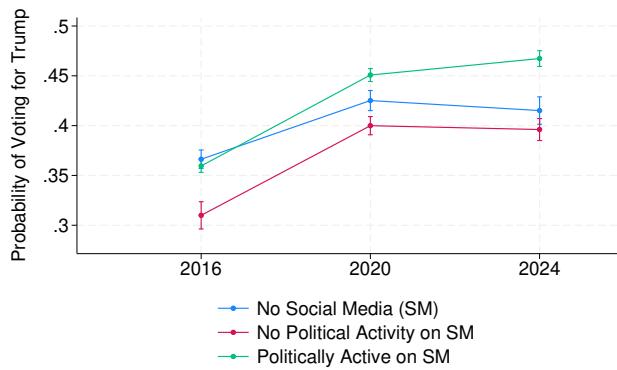
$$684 \text{Trump_Vote} \sim \text{ideology} + \text{party_identity} + \text{education} + \\ 685 \text{foreign_born} + \text{sexuality} + \text{gender} + \\ 686 \text{family_income} + \text{age} + \text{year} + \\ 687 (liking + reading + following \\ 688 + forwarding + positng) \times \text{year}$$

[3]

Results

689 Our results show a strong and robust association between
690 social media-based political engagement and support for
691 Donald Trump in the 2016 presidential election. Across
692 both model specifications, higher levels of online political
693 engagement are consistently linked to a greater likelihood
694 of voting for Trump, even after adjusting for standard
695 demographic, partisan, and ideological covariates. We
696 further find evidence of heterogeneity in these associations
697 across racial groups, indicating that the relationship between
698 online engagement and vote choice operates differently across
699 demographic contexts.

700 In this section, we enumerate the results of our statistical
701 modeling.



736 **Fig. 1.** Probability of voting for Donald Trump
737 by each level of the political engagement variable – those
738 who did not use social media, those who used it but did
739 not engage with political content, and those who engaged
740 with political content on social media – in 2016, 2020 and
741 2024.

742 Figure 1 shows the probability of voting for Donald Trump
743 by each level of the political engagement variable – those
744 who did not use social media, those who used it but did
745 not engage with political content, and those who engaged
746 with political content on social media – in 2016, 2020 and
747 2024.

2024. Across all respondents in the CES surveys, users
 745 who did not engage with political content on social media
 746 were less likely to vote for Trump in 2016 (31% chance of
 747 voting for Trump), compared to those who did not use social
 748 media (36%) or those who engaged with political content
 749 on social media (36.6%). The latter two categories of users
 750 were indistinguishable in their likelihood of supporting Mr.
 751 Trump in the 2016 presidential race. In 2020, however, we
 752 see a clear gradation among the three levels of social media
 753 engagement. Politically engaged social media users had the
 754 highest probability of voting for Trump (45.1%), followed by
 755 non-users of social media (42.5%), and then by users who did
 756 not engage with political content (40%). In 2024, we see yet
 757 another pattern. The gap between politically engaged users
 758 and non-users of social media increases further, whereas there
 759 is no statistically significant difference between politically
 760 inactive social media users and those who did not use social
 761 media at all. In the 2024 race, politically engaged users had
 762 46.73% probability of voting for Trump, those who did not
 763 engage with such content had a 39.6% chance of voting for
 764 Trump, while those did not use social media at all had a
 765 41.5% chance.

766
 767
 768 **Table 4. Difference in probability of voting for Trump by political**
 769 **engagement on social media, split by race and election year**

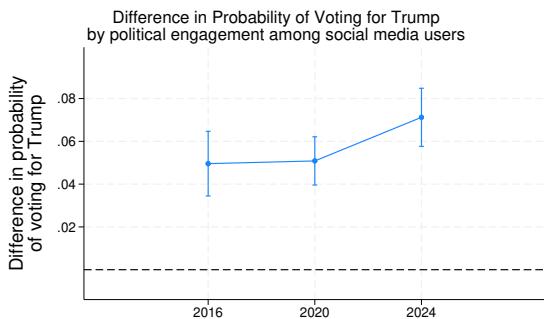
| Race | Year | No Political Engagement on Social Media | Politically Engaged on Social Media |
|-----------------|------|---|-------------------------------------|
| All Respondents | 2016 | -5.63*** (0.84) | -0.68 (0.58) |
| | 2020 | -2.52*** (0.69) | 2.56*** (0.61) |
| | 2024 | -1.90 (0.90) | 5.22*** (0.81) |
| Asian | 2016 | 0.88 (4.28) | 2.44 (3.17) |
| | 2020 | -3.94 (3.84) | 4.49 (3.28) |
| | 2024 | 1.20 (4.93) | 9.95** (4.14) |
| Black | 2016 | -9.12 (4.84) | -1.70 (3.14) |
| | 2020 | -5.61* (2.45) | 2.15 (2.27) |
| | 2024 | -7.24* (3.05) | 6.25* (2.66) |
| Hispanic | 2016 | -13.36*** (3.57) | -5.92* (2.52) |
| | 2020 | -2.13 (2.68) | 7.63** (2.25) |
| | 2024 | 1.10 (3.56) | 5.97 (3.17) |
| White | 2016 | -5.20*** (0.97) | -0.34 (0.67) |
| | 2020 | -2.51** (0.87) | 2.01** (0.76) |
| | 2024 | -1.83 (1.08) | 4.91*** (0.97) |

790 Notes: Baseline category is respondents who **do not use social**
 791 **media**.

792 Values are percentage point differences. Stars: * $p < .10$, **
 793 $p < .05$, *** $p < .01$ (p-values multiplied by 6 to account for
 794 Bonferroni correction).

795 Table 4 tracks these differences with the corresponding
 796 level of statistical significance indicated. We used a Bonferroni
 797 correction factor of 6 for each group of users listed in table
 798 4 – all respondents, Asian Americans, Black Americans,
 799 Hispanic Americans, and White Americans – as there were
 800 six comparisons for each of them. This includes two contrasts
 801 measured per year for a total of 3 years for every group of
 802 respondents.

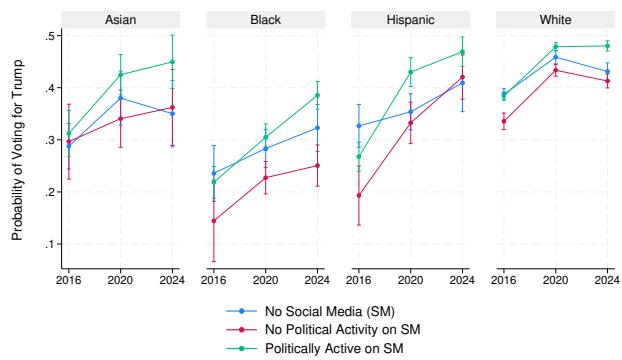
803 To validate hypothesis 1, we measure the difference in
 804 the probability of voting for Trump between social media
 805 users who engaged with political content and those who did
 806



807
 808 **Difference in Probability of Voting for Trump**
 809 **by political engagement among social media users**

810
 811 The validation of hypothesis 1 shows that politically engaged social media users
 812 were significantly more likely to vote for President Trump than those who did not
 813 engage with such content on social media. The gap was higher in 2024 than in 2016
 814 and 2020.

815
 816 not. Figure 2 shows the marginal contrast i.e. difference in probability of voting for Trump in each election year, depending on the user's political engagement on social media. We see that in 2016, politically engaged social media users were 4.96% points more likely to vote for Trump. In 2020, this difference was 5.08% points, whereas in 2024 it rose to 7.12% points. Given that the difference is statistically significant for each year (p -value < 0.001), we have validated hypothesis 1.



817
 818 **Fig. 3. Probability of voting for Trump by political engagement on social media, splits**
 819 **by race in 2016, 2020, and 2024 from Model 1**

820
 821 **Racial Heterogeneity in Voting Behavior.** Figure 3 shows how
 822 the relationship between political engagement on social media
 823 and voting for Trump varies among the four largest racial
 824 groups in the US – Asian, Black, Hispanic, and White
 825 Americans. We track the difference by level of political
 826 engagement for each election cycle. Table 4 shows the
 827 statistical significance of this relationship for each racial
 828 group in every election year. We observe that, among Asian
 829 Americans, political engagement had no significant relationship
 830 with voting for Trump in 2016 and 2020. However, in
 831 2024, social media users who engaged with political content
 832 were 9.95% points more likely to vote for Trump than those
 833 who did not use social media. Among Black respondents,
 834 compared to the baseline, social media users who did not
 835 engage with political content were less likely to vote for
 836 Trump in 2020 and 2024. In 2024, Black users of social
 837 media were 4.49% points less likely to vote for Trump than
 838 those who did not use social media. Among Hispanic
 839 Americans, political engagement was significantly related
 840 to voting for Trump in 2016, 2020, and 2024. In 2024,
 841 political engagement was 7.63% points more likely to vote
 842 for Trump than those who did not use social media. Among
 843 White Americans, political engagement was significantly
 844 related to voting for Trump in 2016, 2020, and 2024. In
 845 2024, political engagement was 4.91% points more likely
 846 to vote for Trump than those who did not use social media.

media who engaged with political content were 6.25% points more likely to vote for Trump than non-users of social media. Among Hispanic Americans, we see a different pattern. In 2016, Hispanic Americans who used social media were less likely to vote for Trump compared to those who used social media, regardless of political engagement. However, in 2020, politically engaged Hispanic American social media users had the highest probability of voting for Trump (7.63% points more than the baseline). Finally, among White Americans, political engagement had no significant relationship with voting for Trump in 2016, but was positively correlated with the same in 2020 (+2.01pp) and 2024 (+4.91pp).

Modes of Political Engagement and Voting Behavior. Research questions RQ2A and RQ2B explore the relationship between individual modes of political engagement on social media and voting behavior in the 2016, 2020, and 2024 presidential elections. Notably, this analysis only includes respondents who used social media. Figure 4 shows the difference in probability of voting for Trump based on each mode of engagement, measured using marginal contrasts from model 2. For instance, in 2024, social media users who read a post about politics were 3.6 percentage points more likely to support Trump than those who did not read a political post on social media.

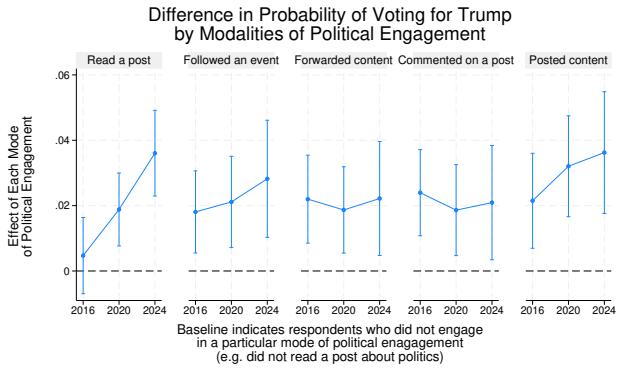


Fig. 4. Probability of voting for Trump by modes of political engagement on social media in 2016, 2020, and 2024 from Model 2

In 2016, reading a political post on social media was not associated with presidential voting behavior, but users who followed a political event were 1.5% points more likely to vote for Trump, and users who forwarded political content on social media were 2.2% points more likely to support Trump. Moreover, posting comments about politics was also positively correlated with Trump support (+2.40% points), as was posting content about politics (+2.15% points). In 2020, all modalities were positively correlated with support for Trump – reading a post (+1.88pp), following an event (+2.11pp), forwarding political content (+1.87pp), commenting about politics (+1.86pp), and posting original political content (+3.20pp). Likewise, in 2024, reading a political post (+3.6pp), following an event (+2.82pp), and posting original content (+3.62pp) had a statistically significant relationship with voting for Trump. However, forwarding a post or commenting about politics were not linked to voting for Trump. We used a Bonferroni correction factor of 15 for the 3 x 5 grid we used to measure contrasts for model 2.

Next, we consider the temporal change in the relationship between modes of political engagement and support for Mr. Trump. We used a difference-in-difference approach to measure the year-on-year change in this relationship, using the 2016 estimates as the baseline. We find that users who read political content on social media were significantly more likely to vote for Mr. Trump in 2024 than they were in 2016. No other mode of engagement shows a significant temporal change in its probability of Trump support.

Discussion

Our analysis reveals consistent and positive associations between various forms of political engagement on social media and the preference for voting for Donald Trump in 2016, 2020, and 2024. This finding, which holds even after controlling for demographic and partisan factors, supports our primary hypothesis and contributes to a growing body of evidence on the potent role of social media in shaping contemporary political behavior (40). Racial identities significantly moderate the relationship between online political engagement and voting behavior, with the impact varying across different groups and evolving over time. In this section, we elaborate on the theoretical mechanisms that can explain why these patterns emerged; we draw from the literature on political communication, social psychology, and network science (41) and propose that the observed associations result from a confluence of reinforcing mechanisms of social media's architecture and the nature of populist campaigns.

One primary mechanism of social media's influence on voting behavior is the creation of ideologically homogeneous online environments, often referred to as “echo chambers” or “filter bubbles” (11, 42) that reinforce attitudes and their intensity. Social media platforms, through their algorithmic content curation, tend to show users content that aligns with their previous engagement patterns (43). This, combined with individuals' natural tendency toward homophily—connecting with like-minded others—can insulate users from divergent viewpoints (21). Within these digital spaces, pro-Trump narratives, whether originating from the candidate himself, affiliated media, or fellow supporters, are amplified and reinforced (42). Constant exposure to ideologically congruent information can strengthen partisan identity, increase attitude intensity, and make the act of voting for the in-group's preferred candidate seem not only natural but necessary (44). Our finding that even passive forms of engagement, such as reading political posts, are associated with support for Trump is consistent with this explanation. Mere immersion in a pro-Trump information ecosystem can be a powerful force for persuasion and mobilization (26?).

When users were shown content that was ideologically opposed to their beliefs, Republicans were more likely to react negatively and further entrench their beliefs, while Democrats did not change their views significantly. This suggests that Mr. Trump's supporters were more susceptible to affective polarization — the tendency of individuals to feel more negatively toward the opposing political party (45). Affective polarization is a reinforcing spiral: individuals may have an initial susceptibility that is triggered by social media to produce a polarized outcome, which then functions as a mechanism to instigate voting behavior.

993 Furthermore, political content on social media, particularly
994 content related to a polarizing figure like Donald Trump,
995 is often emotionally charged, framing political opponents
996 not just as wrong, but as immoral or dangerous (?).
997 Engagement with emotionally charged content can intensify
998 feelings of animosity towards the out-group and strengthen
999 feelings of solidarity with the in-group (46). This heightened
1000 emotional state can be a powerful motivator for political
1001 action, including voting, as a means of defending one's group
1002 and defeating the opposition (47). The interactive nature
1003 of social media, where users can see their friends and social
1004 connections expressing similar outrage or enthusiasm, further
1005 validates and amplifies these affective responses (48).

1006 Another key mechanism is the 'gateway' function of social
1007 media engagement (7). Platforms lower the barrier to political
1008 participation by offering a spectrum of low-cost, expressive
1009 activities, such as liking, sharing, or commenting (6). While
1010 these actions may seem trivial in isolation, they serve as initial
1011 steps on a ladder of political engagement (?), easing the path
1012 to develop and perform a political identity (?). Engaging in
1013 these low-interaction activities can increase individuals' sense
1014 of political efficacy and make them more receptive to calls for
1015 higher-cost participation, such as attending a rally, donating
1016 money, or casting a vote (7). The habitual nature of social
1017 media use reinforces this process; daily, repeated engagement
1018 with political content keeps users mobilized and connected
1019 to the campaign's narrative (49). That such mechanisms
1020 are structurally beneficial to Mr. Trump, but not to the
1021 Democratic party, is an interesting finding.

1022 The last structural mechanism of social media we suggest
1023 is at play is political actors' ability to use social media
1024 to bypass traditional media gatekeepers and communicate
1025 directly with their supporters (14, 50). This dis-intermediated
1026 communication style is a hallmark of modern populism,
1027 enabling leaders to cultivate a sense of authenticity and a
1028 direct, personal connection with their base (13, 14). Through
1029 platforms like Twitter, Trump was able to rally his supporters
1030 and attack his opponents in real-time, creating a continuous,
1031 interactive campaign that traditional media struggled to keep
1032 pace with.

1033 This direct channel also facilitates the rapid spread of
1034 information and, critically, misinformation that is favorable
1035 to the campaign (9, 12). Research on the 2016 election,
1036 for example, found that false news was widely shared,
1037 disproportionately favored Donald Trump, and potentially
1038 influencing voting decisions (51). The engagement we observe
1039 in our data is, in part, engagement with this unique and
1040 powerful communication apparatus (?). Our study does not
1041 consider the content people engaged with on social media
1042 and cannot address a potential relationship between support
1043 for Mr. Trump with misinformation.

1044 TKTK

1045 Racial identities significantly moderate the relationship
1046 between online political engagement and voting behavior.
1047 The specific impacts impact vary across different groups and
1048 evolve over time. For instance, the relationship between online
1049 activity and voting for Trump among Asian Americans saw
1050 a dramatic shift between election cycles. In 2024, politically
1051 engaged Asian American users were more likely to vote for
1052 Trump than those who did not use social media at all. The
1053 racial heterogeneity observed suggests that social media's echo
1054

1055 chambers, dis-intermediated nature, and emotional impacts
1056 do not operate uniformly across all demographic groups
1057 (52). For instance, the experience of being in a political
1058 echo chamber, or the emotional impact of certain types
1059 of political messaging, may differ significantly for White,
1060 Black, or Hispanic voters (53). Future work should employ
1061 qualitative or mixed-methods approaches to explore the lived
1062 experience of political engagement on social media across
1063 different communities, and experimental designs to more
1064 precisely isolate the causal effects of these various mechanisms
1065 (?).

1066 In conclusion, our findings underscore the profound impact
1067 of social media on the American electorate (?). The positive
1068 association between online political engagement and support
1069 for Donald Trump is best understood through a multi-faceted
1070 theoretical lens that accounts for the technological affordances
1071 of platforms, the psychological dynamics of group identity,
1072 and the communication strategies of populist leaders (?).
1073 Understanding these mechanisms is crucial for navigating the
1074 challenges and opportunities of democratic life in the digital
1075 age (54).

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A. Contrast Tables

Table 5. Year-to-Year Contrasts in Social Media Political Engagement by Race

| Race | Social Media Use | 2020 | vs | 2024 | vs |
|----------|-------------------------------|--------------------|----|--------------------|----|
| | | 2016 | | 2020 | |
| Asian | No Social Media use | 0.092** (0.03) | | -0.030 (0.04) | |
| | No Political Engagement on SM | 0.044 (0.05) | | 0.022 (0.05) | |
| | Politically Engaged on SM | 0.113*** (0.03) | | 0.025 (0.03) | |
| Black | No Social Media use | 0.048 (0.03) | | 0.040 (0.03) | |
| | No Political Engagement on SM | 0.083 (0.04) | | 0.023 (0.03) | |
| | Politically Engaged on SM | 0.086*** (0.02) | | 0.081*** (0.02) | |
| Hispanic | No Social Media use | 0.027 (0.03) | | 0.056 (0.03) | |
| | No Political Engagement on SM | 0.139*** (0.04) | | 0.088** (0.03) | |
| | Politically Engaged on SM | 0.162*** (0.02) | | 0.039 (0.02) | |
| White | No Social Media use | 0.071*** (0.01) | | -0.027** (0.01) | |
| | No Political Engagement on SM | 0.098*** (0.01) | | -0.021* (0.01) | |
| | Politically Engaged on SM | 0.095*** (0.01) | | 0.002 (0.01) | |

Note: Entries are contrasts with standard errors in parentheses.

Significance levels (adjusted for multiple testing with Bonferroni, $p \times 6$): * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$.