Trends in the Diffusion of Misinformation on Social Media

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Abstract

We measure trends in the diffusion of misinformation on Facebook and Twitter between January 2015 and July 2018. We focus on stories from 570 sites that have been identified as producers of false stories. Interactions with these sites on both Facebook and Twitter rose steadily through the end of 2016. Interactions then fell sharply on Facebook while they continued to rise on Twitter, with the ratio of Facebook engagements to Twitter shares falling by approximately 60 percent. We see no similar pattern for other news, business, or culture sites, where interactions have been relatively stable over time and have followed similar trends on the two platforms both before and after the election.

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1 Introduction

Misinformation on social media has caused widespread alarm in recent years. A substantial number of U.S. adults were exposed to false news stories prior to the 2016 election, and post-election surveys suggest that many people who read such stories believed them to be true (Silverman and Singer-Vine 2016; Allcott and Gentzkow 2017; Guess et al. 2018). Many argue that false news stories played a major role in the 2016 election (for example, Olson 2016; Parkinson 2016; Read 2016; Gunther et al. 2018), and in the ongoing political divisions and crises that have followed it (for example, Spohr 2017; Azzimonti and Fernandes 2018; Tharoor 2018). Numerous efforts have been made to respond to the threat of false news stories, including educational and other initiatives by civil society organizations, hearings and legal action by regulators, and a range of algorithmic, design, and policy changes made by Facebook and other social media companies.

Evidence on whether these efforts have been effective—or how the scale of the misinformation problem is evolving more broadly—remains limited. A recent study argues that false stories remain a problem on Facebook even after changes to its news feed algorithm in early 2018 (Newswhip 2018). The study reports that the 26th and 38th most engaging stories on Facebook in the two months after the changes were from fake news websites. Many articles that have been rated as false by major fact-checking organizations have not been flagged in Facebook’s system, and two major fake news sites have seen little or no decline in Facebook engagements since early 2016 (Funke 2018). Facebook’s now-discontinued strategy of flagging inaccurate stories as “Disputed” can modestly lower the perceived accuracy of flagged headlines (Blair et al. 2017), though some research suggests that the presence of warnings can cause untagged false stories to be seen as more accurate (Pennycook and Rand 2017). Media commentators have argued that efforts to fight misinformation through fact-checking are “not working” (Levin 2017) and that misinformation overall is “becoming unstoppable” (Ghosh and Scott 2018).

In this paper, we present new evidence on the volume of misinformation circulated on social media from January 2015 to July 2018. We assemble a list of 570 sites identified as sources of false stories in a set of five previous studies and online lists. We refer to these collectively as fake news sites. We measure the volume of Facebook engagements and Twitter shares for all stories on these sites by month. As points of comparison, we also measure the same outcomes for stories on (i) a set of major news sites; (ii) a set of small news sites not identified as producing misinformation; and (iii) a set of sites covering business and culture topics.

The results show that interactions with the fake news sites in our database rose steadily on both Facebook and Twitter from early 2015 to the months just after the 2016 election. Interactions then declined by more than half on Facebook, while they continued to rise on Twitter. The ratio of Facebook engagements to Twitter shares was roughly steady at around 40:1 from the beginning of our period to late 2016, then fell to roughly 15:1 by the end of our sample period. In contrast,
interactions with major news sites, small news sites, and business and culture sites have all re-
mained relatively stable over time, and have followed similar trends on Facebook and Twitter both
before and after the 2016 election. While this evidence is far from definitive, we see it as consist-
tent with the view that the overall magnitude of the misinformation problem may have declined, at
least temporarily, and that efforts by Facebook following the 2016 election to limit the diffusion of
misinformation may have had a meaningful impact

The results also show that the absolute level of interaction with misinformation remains high,
and that Facebook continues to play a particularly important role in its diffusion. In the period
around the election, fake news sites received almost as many Facebook engagements as the 38
major news sites in our sample. Even after the sharp drop following the election, Facebook en-
gagements of fake news sites still average roughly 70 million per month.

Our evidence is subject to many important caveats and must be interpreted with caution. This is
particularly true for the raw trends in interactions. While we have attempted to make our database
of false stories as comprehensive as possible, it is likely far from complete, and many factors could
generate selection biases that vary over time. The raw decline in Facebook engagements may
partly reflect the under-sampling of sites that could have entered or gained popularity later in our
sample period, as well as efforts by producers of misinformation to evade detection on Facebook by
changing their domain names. It may also reflect changes over time in demand for highly partisan
political content that would have existed absent efforts to fight misinformation, and could reverse
in the future, for example in the run-up to future elections.

We see the comparison of Facebook engagements to Twitter shares as potentially more infor-
mative. If the design of these platforms and the behavior of their users were stable over time, we
might expect sample selection biases or demand changes to have similar proportional effects, and
thus leave the ratio of Facebook engagements to Twitter shares roughly unchanged. For example,
we might expect producers changing domain names to evade detection to produce similar declines
in our measured interactions on both platforms. The fact that Facebook engagements and Twit-
ter shares follow similar trends prior to late 2016 and for the non-fake-news sites in our data, but
diverge sharply for fake news sites following the election, suggests that some factor has slowed
the relative diffusion of misinformation on Facebook. The suite of policy and algorithmic changes
made by Facebook following the election seems like a plausible candidate.

However, even the relative comparison of the platforms is only suggestive. Both Facebook
and Twitter have made changes to their platforms, and so at best this measure captures the relative
effect of the former compared to the latter. Engagements on Facebook affect sharing on Twitter
and vice versa. The selection of stories into our database could for various reasons differentially
favor the kinds of stories likely to be shared on one platform or the other, and this selection could
vary over time. Demand changes need not have the same proportional effect on the two platforms.
Some of these factors would tend to attenuate changes in the Facebook-Twitter ratio, leading our
results to be conservative, but others could produce a spurious decrease over time.

In the appendix, we show that our qualitative results survive a set of robustness checks intended to partially address potential sample selection biases. These checks include: (i) focusing on sites identified as fake in multiple lists; (ii) excluding sites from each of our five lists in turn, (iii) looking at sites that were active in different periods; (iv) excluding potential outliers and looking at sites of different sizes; and (v) looking at sites with different likelihoods to publish misinformation.

2 Background

Both Facebook and Twitter have taken steps to reduce the circulation of misinformation on their platforms. In the appendix, we list twelve such announcements by Facebook and five by Twitter since the 2016 election. Broadly, the platforms have taken three types of actions to limit misinformation. First, they have limited its supply, by blocking ads from pages that repeatedly share false stories and removing accounts that violate community standards. Second, they have introduced features such as “disputed” flags or “related articles” that provide corrective information related to a false story. Third, they have changed their algorithms to de-prioritize false stories in favor of news from trustworthy publications and posts from friends and family.

Legislators are also taking action. For example, Connecticut, New Mexico, Rhode Island, and Washington passed laws in 2017 encouraging media literacy and digital citizenship (Zubrzycki 2017). Executives from Facebook, Google, and Twitter have been asked to testify before various congressional committees about their efforts to combat misinformation (Shaban et al. 2017; Popken 2018). Although there has been no major national legislation, this testimony may have raised public awareness.

Finally, civil society organizations also play an important role. For example, the News Literacy Project provides non-partisan educational materials to help teachers educate students to evaluate the credibility of information; demand for its materials has grown substantially in the past few years (Strauss 2018). In 2017, the newly established News Integrity Initiative (NII) made ten grants totaling $1.8 million to help build trust between newsrooms and the public, make newsrooms more diverse and inclusive, and make public conversations less polarized (Owen 2017).

3 Data

We compile a list of sites producing false news stories by combining five previous lists: (i) a research project by Grinberg et al. (2018, 490 sites); (ii) PolitiFact’s article titled “PolitiFact’s guide to fake news websites and what they peddle” (Gillin 2017, 325 sites); (iii) three articles by BuzzFeed on fake news (Silverman 2016; Silverman et al. 2017a; Silverman et al. 2017b; 223 sites); (iv) a research project by Guess et al. (2018, 92 sites); and (v) FactCheck’s article titled “Websites
that post fake and satirical stories” (Schaedel 2017, 61 sites). Politifact and FactCheck are independent journalistic fact-checking websites, while BuzzFeed similarly applies journalistic standards to evaluating whether articles are true or false. The two lists from research projects originally derive from subsets of the other three, plus Snopes.com, another independent fact-checking site, and lists assembled by blogger Brayton (2016) and media studies scholar Zimdars (2016). The union of these five lists is our set of fake news sites.

PolitiFact and FactCheck work directly with Facebook to evaluate the veracity of stories flagged by Facebook users as potentially false. Thus, these lists comprise fake news sites that Facebook is likely to be aware are fake. As a result, our results may be weighted toward diffusion of misinformation that Facebook is aware of, and may not fully capture trends in misinformation that Facebook is not aware of. It is difficult to assess how large this latter group might be. Our list almost certainly includes the most important providers of false stories, as Facebook users can flag any and all questionable articles for review. On the other hand, the list might exclude a large tail of small sites producing false stories.

Combining these five lists yields a total of 673 unique sites. We have data for 570 of them. We report in the appendix the names and original lists of 50 largest sites in terms of total Facebook engagements plus Twitter shares during the sample period. In our robustness checks, we consider alternative rules for selecting the set of sites.

Our sets of comparison sites are defined based on category-level web traffic rankings from Alexa.1 Alexa measures web traffic using its global traffic panel, a sample of millions of Internet users who have installed browser extensions allowing their browsing data to be recorded, plus data from websites that use Alexa to measure their traffic. It then ranks sites based on a combined measure of unique visitors and pageviews. We define major news sites to be the top 100 sites in Alexa’s News category. We define small news sites to be the sites ranked 401-500 in the News category. We define business and culture sites to be the top 50 sites in each of the Arts, Business, Health, Recreation, and Sports categories. For each of these groups, we omit from our sample government websites, databases, sites that do not mainly produce news or similar content, international sites whose audiences are primarily outside the U.S., and sites that are included in our list of fake news sites. Our final sample includes 38 major news sites, 78 small news sites, and 54 business and culture sites.

We gather monthly Facebook engagements and Twitter shares of all articles published on these sites from January 2015 to July 2018 from BuzzSumo (www.buzzsumo.com). BuzzSumo is a commercial content database that tracks the volume of user interactions with internet content on Facebook, Twitter, and other social media platforms, using data available from the platforms’ application programming interfaces (APIs). We use BuzzSumo’s data on total Facebook engagements and total Twitter shares by originating website and month. Facebook engagements are defined

1 © 2018, Alexa Internet (www.alexa.com)
as the sum of shares, comments, and reactions such as "likes." We have data for 570 out of 673 fake news sites in our list and all sites in the comparison groups. We sum the monthly Facebook engagements and Twitter shares of articles from all sites in each category and then average by quarter.

4 Results

Figure 1 shows trends in the number of Facebook engagements and Twitter shares of stories from each category of site. Interactions for major news sites, small news sites, and business and culture sites have remained relatively stable during the past two years, and follow similar trends on Facebook and Twitter. Both platforms show a modest upward trend for major news and small news sites, and a modest downward trend for business and culture sites. In contrast, interactions with fake news have changed more dramatically over time, and these changes are very different on the two platforms. Fake news interactions increased steadily on both platforms from the beginning of 2015 up to the 2016 election. Following the election, however, Facebook engagements fell sharply (declining by more than 50 percent), while shares on Twitter continued to increase.

Figure 2 shows our main result: trends in the ratio of Facebook engagements to Twitter shares. The ratios have been relatively stable for major news, small news, and business and culture sites. For fake news sites, however, the ratio has declined sharply, from around 45:1 during the election to around 15:1 two years later.

While these results suggest that the circulation of fake news on Facebook has declined in both absolute and relative terms, it is important to emphasize that the absolute quantity of fake news interactions on both platforms remains large, and that Facebook in particular has played an outsized role in its diffusion. Figure 1 shows that Facebook engagements fell from a peak of roughly 200 million per month at the end of 2016 to roughly 70 million per month at the end of our sample period. As a point of comparison, the 38 major news sites in the top left panel—including the New York Times, Wall Street Journal, CNN, Fox News, etc.—typically garner about 200-250 million Facebook engagements per month. On Twitter, fake news shares have been in the 4-6 million per month range since the end of 2016, compared to roughly 20 million per month for the major news sites.

We report a number of robustness checks in the appendix designed to address concerns about selection into our sample of sites. First, we restrict to sites that are identified as fake news sites by at least two or three of our original five lists, which leaves 116 and 19 sites, respectively. Second, given that people might disagree with any one particular study’s list of fake news sites, we run five additional analyses, each excluding fake news sites identified exclusively by one of our five lists. Third, we focus on sites that started active operations after November 2016, sites that were still in active operation as of July 2018, and sites that were in active operation from August 2015
to July 2018, which leaves 226, 215, and 82 sites respectively. (Active operation is defined to
be a global traffic rank reported by Alexa of at least one million.) Fourth, we exclude the five
largest sites in terms of total interactions to ensure the trend is not driven solely by outliers. We
also look at sites in the first decile and sites in the bottom nine deciles separately to see if the
trend holds for both large sites and small sites. Fifth, Grinberg et al. (2018) provide three lists of
sites classified by different likelihoods to publish misinformation. We look at each of these lists
separately. Our main qualitative conclusions remain consistent across these checks, though the
exact size and shape of the trends vary. Finally, we present an alternative comparison group: a
small set of politically focused sites such as Politico and The Hill. These sites do see a decline in
engagements on Facebook relative to Twitter, but it mainly occurred in late-2015.
Figure 1: Engagement on Facebook and Twitter

Panel A: Facebook Engagements

Panel B: Twitter Shares

Notes: This figure shows monthly Facebook engagements and Twitter shares of all articles published on sites in different categories averaged by quarter. Data comes from BuzzSumo. Major News Sites include 38 sites selected from the top 100 sites in Alexa’s News category. Small News Sites include 78 sites selected from the sites ranking 401-500 in the News category. Business and Culture Sites include 54 sites selected from the top 50 sites in each of the Arts, Business, Health, Recreation, and Sports categories. Fake News Sites include 570 sites assembled from five lists. The complete lists can be found in the appendix.
Figure 2: Relative Engagement on Facebook

Notes: This figure shows the ratio of monthly Facebook engagements over Twitter shares of all articles published on sites in different categories averaged by quarter. Data comes from BuzzSumo. Major News Sites include 38 sites selected from the top 100 sites in Alexa’s News category. Small News Sites include 78 sites selected from the sites ranking 401-500 in the News category. Business and Culture Sites include 54 sites selected from the top 50 sites in each of the Arts, Business, Health, Recreation, and Sports categories. Fake News Sites include 570 sites assembled from five lists. The complete lists can be found in the appendix.
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