

# Real Solutions for Fake News?

## Measuring the Effectiveness of General Warnings and Fact-Check Tags in Reducing Belief in False Stories on Social Media

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### Abstract

Social media has increasingly enabled “fake news” to circulate widely, most notably during the 2016 U.S. presidential campaign. These intentionally false or misleading stories threaten the democratic goal of a well-informed electorate. This study evaluates the effectiveness of strategies that could be used by Facebook and other social media to counter false stories. Results from a pre-registered experiment indicate that false headlines are perceived as less accurate when people receive a general warning about misleading information on social media or when specific headlines are accompanied by a “Disputed” or “Rated false” tag. Though the magnitudes of these effects are relatively modest, they generally do not vary by whether headlines were congenial to respondents’ political views. In addition, we find that adding a “Rated false” tag to an article headline lowers its perceived accuracy more than adding a “Disputed” tag (Facebook’s original approach) relative to a control condition. Finally, though exposure to the “Disputed” or “Rated false” tags did not affect the perceived accuracy of unlabeled false or true headlines, exposure to a general warning *decreased* belief in the accuracy of true headlines, suggesting the need for further research into how to most effectively counter false news without distorting belief in true information.

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Since the 2016 U.S. election, the effects of “fake news” have received considerable attention. Many Americans now worry about the effects of this factually dubious content that imitates the format of journalism but is produced with no regard for accuracy or fairness (Lazer et al. 2018). This type of content, which we will refer to as “false news” for expositional clarity, is most often created for profit by dubious websites.<sup>1</sup> Many people appear to believe false news stories, which circulated widely before the election and generally favored Donald Trump over Hillary Clinton (Silverman 2016; Silverman and Singer-Vine 2016). Although false news most likely did not change the election’s outcome (Allcott and Gentzkow 2017), its prevalence is still an important concern. False news promotes misperceptions among voters and can induce distrust of legitimate information. In this sense, it presents a serious threat to American democracy.

The public’s vulnerability to false information has grown as people have come to increasingly rely on social media as a source of news. According to a recent Pew survey, 62 percent of American adults get news from social media sites such as Facebook (Gottfried and Shearer 2017), which played an especially important role in the spread of false news during the 2016 presidential campaign. The most viral false news articles were shared more on Facebook in the months prior to the election than the most widely shared mainstream news stories (Silverman 2016). Online misinformation, both political and otherwise, has continued to be a challenge since the election. For example, false claims swirled around social media in the aftermath of Hurricane Harvey, including an article claiming that Black Lives Matter protesters blocked emergency responders from reaching hurricane victims (Schaedel 2017).

To address these concerns, Facebook began adding “Disputed” tags to stories in its News Feed that have been debunked by fact-checkers in December 2016 (Mosseri 2016). It used this approach for approximately one year before switching to providing fact-checks in a “Related Articles” format underneath suspect stories (Smith, Jackson, and Raj 2017). The company also promoted tips for

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<sup>1</sup>“Fake news” has many definitions and is frequently used in imprecise or confusing ways. Moreover, the debate over the meaning of the term and related concepts raises epistemological issues that are beyond the scope of this paper (e.g., speaker intent; see Wardle and Derakhshan 2017). We therefore employ “false news” as an alternative term throughout this paper, which define as described above (“factually dubious content that imitates the format of journalism but is produced with no regard for accuracy or fairness”; see Lazer et al. 2018). This approach is consistent with the practices of various news and social media sources (e.g., Oremus 2017) and is intended to avoid unnecessary confusion.

spotting false news at the top of News Feed in April 2017 and May 2018 (Constine 2017; Owen 2018). Both approaches presumably seek to reduce the probability that people will believe false news articles.

Research suggests that combating misinformation is a difficult challenge (for reviews, see, e.g., Flynn, Nyhan, and Reifler 2017; Lewandowsky et al. 2012). In particular, studies that focus specifically on exposure to false news on social media have found mixed results. Though “disputed” tags seem to modestly reduce belief in false news headlines, they may fail to counteract exposure effects over time (Pennycook, Cannon, and Rand 2017) and could create an “implied truth” effect in which unlabeled false headlines are seen as more accurate (Pennycook and Rand 2017). Similarly, Ecker, Lewandowsky, and Tang (2010) find that specific warnings are more effective than general warnings at reducing the continued influence of exposure to false information on beliefs, but neither approach eliminates this effect entirely. They argue that a specific warning (directly alerting readers about how misinformation can continue to influence them even after being debunked) reduces belief in false claims by helping people to tag misinformation, whereas a general warning (telling participants that the media sometimes does not check facts before publishing information that turns out to be inaccurate) promotes “nonspecifically induced alertness” (Ecker, Lewandowsky, and Tang 2010, p. 1096) that is less effective.

In this study, we investigate whether interventions like the ones used by Facebook can effectively reduce belief in false news. Specifically, we test the effects of both a general warning about false news and two types of specific warnings about individual articles questioned by fact-checkers. Our results indicate that exposure to a general warning about false news modestly reduces the perceived accuracy of false headlines. We also find that adding a “Rated false” or “Disputed” tag underneath headlines reduces their perceived accuracy somewhat more. In particular, the “Rated false” tag is most effective at reducing the perceived accuracy of false headlines, though neither tag measurably reduced the self-reported likelihood that headlines would be shared on social media. The effects of these tags did not vary consistently depending on whether participants had previously received a general warning. Similarly, there were not consistent differences between the effect of tags on

politically congenial versus non-congenial information. Finally, though we find no evidence that tagging headlines as “Rated false” or “Disputed” has large spillover effects to belief in other headlines, exposure to a general warning did reduce belief in the accuracy of true headlines as well as false ones, suggesting that efforts to promote greater skepticism toward false news can also increase distrust of legitimate news and information.

## **Theoretical expectations**

We specifically test the following hypotheses and research questions, which were pre-registered at EGAP prior to the administration of our study (<http://www.egap.org/registration/2516>).

First, though people’s initial belief in false information can be difficult to change (see Flynn, Nyhan, and Reifler 2017 for a review), some evidence suggests that warnings about false information can reduce belief in false claims or prevent the uptake of misinformation. Ecker, Lewandowsky, and Tang (2010) find that warnings about the limits of fact-checking in the media reduce belief in outdated facts and increased acceptance of correct information, but do not entirely eliminate the effect of misinformation. Similarly, Bolsen and Druckman (2015) find that warnings are more effective than corrections at countering directionally motivated reasoning about scientific claims.

We focus specifically on headlines, which are the dominant form of content in social media and can be misleading even to relatively attentive readers (e.g., Ecker et al. 2014). Our study tests the effectiveness of two approaches that have been used by Facebook to try to reduce belief in false news: general warnings to beware of misleading content and specific tags on article headlines that mark them as “Disputed.” We also test the effectiveness of specific tags that instead mark headlines as “Rated false.”

The first approach we test is a general warning. In April 2017 and May 2018, Facebook rolled out a warning of this sort to users, distributing a message at the top of News Feed that highlighted “tips for spotting fake news” (Mosseri 2017; Owen 2018). In this experiment, we use an analogous

warning message to test the following hypothesis:

H1: Exposure to a general warning about misleading articles will reduce the perceived accuracy of false headlines relative to a no-warning condition.

Our study also tests the effect of a specific warning by building on Pennycook, Cannon, and Rand (2017), who find that a Facebook-style “Disputed” tag under headlines reduces belief in the accuracy of false stories and reduces users’ intent to share them. Similarly, Bode and Vraga (2015) find that including corrective information in Facebook’s “related stories” function, which links articles to other articles that may correct false claims, effectively reduces misperceptions. These interventions warn users about misinformation or false news at the time when they are exposed to a headline and are intended to help readers notice false information as soon as they encounter it. We therefore propose the following hypothesis:

H2a: The presence of a Facebook-style “Disputed” tag under false headlines will reduce their perceived accuracy relative to a no-tag condition.

However, tags warning that a claim is “Disputed” may not be sufficiently direct. We therefore evaluate the effect of a specific warning directly stating that a false news headline is untrue in an additional condition. A warning of this nature, though not yet used by Facebook, might convey a stronger message than the inconclusive terminology of the “Disputed” warning. Communicating expert consensus has been found to increase belief in global warming and support for expert views on other environmental problems (Aklin and Urpelainen 2014; Bolsen and Druckman 2015; Corbett and Durfee 2004). Personal agreement with the existence of global warming rises consistently with communicated levels of scientific agreement (Chinn, Lane, and Hart 2018). Moreover, media coverage that does not take a side in an effort to appear “balanced” can distort public perceptions of expert consensus (Boykoff and Boykoff 2004; Koehler 2016). Finally, audiences make inferences about the communicators of information and the context that information comes from when they process new information (e.g., Wegner et al. 1981). In the context of false news on Facebook, “Disputed” tags may signal ambivalence about how strongly the platform endorses the fact-check.

Our alternate specific warning describes a false news headline as “Rated False by Snopes and Politifact.” This tag is specific enough to effectively reduce belief (Ecker, Lewandowsky, and Tang 2010) and more clearly conveys the consensus among fact-checking websites that the claim in the article is false. We therefore expect that the effects of the “Rated false” tag would be larger than the effects of the “Disputed” tag and propose the following hypothesis:

H2b: The presence of a “Rated false” tag under false headlines will reduce their perceived accuracy relative to a no-tag condition.

H2c: The presence of a “Rated false” tag under false headlines will reduce their perceived accuracy relative to a Facebook-style “Disputed” tag.

Ecker, Lewandowsky, and Tang (2010) do not test how general and specific warnings work in tandem with one another, though it is plausible that a general warning could increase alertness to subsequent specific warnings about false information. Indeed, van der Linden et al. (2017) find that presenting respondents with information on the scientific consensus about global warming, as well as a general or specific statement about the existence of climate change, was more effective at inoculating respondents against misinformation on climate change than either treatment alone. We therefore propose the following hypothesis about how exposure to a general warning will strengthen the effects of specific warnings:

H3: Exposure to a general warning about misleading articles will increase the negative effects of “Disputed” or “Rated false” tags on the perceived accuracy of false headlines.

Consistent with prior research (Flynn, Nyhan, and Reifler 2017; Kahan 2015), we also expect that people’s belief in false news depends on whether it aligns with their political identity and preferences. “Disputed” or “Rated false” tags could be less effective when a person is viewing a headline with which they are inclined to agree (e.g., Nyhan and Reifler 2010; Kahan et al. 2017). In this case, we evaluate how the effects of our experimental manipulations vary depending on participants’ approval of President Trump (each article concerns either President Trump and his

allies or his opponent, Hillary Clinton) and the slant of the articles in question. For example, the negative effect of a “Disputed” or “Rated false” tag on the perceived accuracy of a news story may be attenuated if the news story is politically congenial (e.g., a pro-Trump headline seen by a Trump supporter). We therefore hypothesize that the effect of specific warnings will be reduced when they accompany politically congenial information relative to uncongenial information:

H4: The effect of a Facebook-style “Disputed” tag on perceived accuracy (H4a) or “Rated false” tag (H4b) will be reduced for politically congenial information versus uncongenial information (versus a headline with no tag).

Finally, we also seek to answer three pre-registered research questions about which we had weaker theoretical expectations. Drawing on previous research identifying the importance of political preferences on belief in misinformation, we investigate whether the effect of warnings on the perceived accuracy of headlines varies between congenial information and uncongenial information (RQ1). We also investigate whether specific warnings on a headline will affect the perceived accuracy of untagged false (RQ2a)<sup>2</sup> or true (RQ2b) headlines, and whether a general warning about misleading articles will reduce the perceived accuracy of true information (RQ3). Practically, it would be difficult for social media platforms to fact-check and add a “Disputed” or “Rated false” tag to every false news headline. Because some false news headlines could inevitably fall through the cracks, we are interested in seeing how general and specific warnings influence respondents’ perceived accuracy of such news items.<sup>3</sup>

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<sup>2</sup>Pennycook and Rand (2017), which we had not seen at the time of pre-registration, also considers this question.

<sup>3</sup>We pre-registered an additional research question about the effects of exposure to a general warning and/or to a “Disputed” or “Rated false” tag on respondents’ self-reported likelihood of “liking” and sharing the headlines on Facebook. The results of this analysis are presented in Online Appendix B.

# Methods

## Participants

The study, which was approved by the Dartmouth College Committee for the Protection of Human Subjects (STUDY0003028), was conducted from May 8–9, 2017 among participants recruited from Amazon Mechanical Turk (MTurk). Although samples from MTurk are not nationally representative, results from studies conducted with participants from the site mirror those obtained from other samples (e.g., Berinsky, Huber, and Lenz 2012; Coppock 2016; Horton, Rand, and Zeckhauser 2011; Mullinix et al. 2015).<sup>4</sup> Non-U.S. residents, people under 18 years of age, and people who completed a prior pilot study were not allowed to participate.<sup>5</sup> We also exclude six respondents from the data who dropped out prior to the experimental manipulation. Our final sample is 2,994 participants.

Although our sample is diverse, it skews female (54% female), younger (median age group 25–34) and more educated (55% have a bachelor’s degree or greater) than the U.S. population. Our sample also overrepresents Democrats—32% identify as Republican or lean Republican, whereas 58% identified as Democrat or lean Democrat. Participants also approve of Trump (30%) and voted for him (30% of those who report voting) at lower rates than the U.S. population. A detailed comparison of the composition of our sample to population benchmarks is provided in Online Appendix C (Table C1).

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<sup>4</sup>A minority of studies conclude that MTurk samples are not externally valid (e.g., Krupnikov and Levine 2014). For example, participants on MTurk tend to skew liberal and young. Moreover, the underrepresentation of conservatives and older participants may suggest that these participants differ from other conservatives or older individuals in the general population. However, numerous studies find that experimental treatment effect estimates typically generalize from MTurk to national probability samples, suggesting these problems are rare (e.g., Berinsky, Huber, and Lenz 2012; Coppock 2016; Horton, Rand, and Zeckhauser 2011; Mullinix et al. 2015). Finally, our MTurk sample is externally valid in the sense that it is made up disproportionately of frequent users of the Internet—precisely the group who may be most likely to encounter false news (Pennycook and Rand 2018a). We thus conclude that respondents from MTurk constitute a valid sample for testing our hypotheses, though replication on representative samples would of course be desirable.

<sup>5</sup>The pilot study tested the effects of “Disputed” and “Rated false” tags only on perceived accuracy and likelihood of liking/sharing for six false news headlines. The results of this study were similar to our main analysis, and are available upon request.



## Experimental design and procedure

We focus on beliefs in headlines because of their primacy on social media (Gabelkov et al. 2016; Manjoo 2013). The initial judgments that people form when reading headlines are also likely to shape their subsequent beliefs and opinions (Thorson 2016).

The experiment used a  $2 \times 3$  between-subjects design that also includes a pure control group. Participants were randomly assigned with equal probability to a pure control group or to one of six experimental conditions (see Table 1). We manipulated whether participants were exposed to a general warning about misleading articles or not (middle column of Table 1). We also independently randomized non-controls into one of three headline conditions: a condition in which no fact-checking tags were presented (first two rows of Table 1), a specific warning condition that included tags labeling articles as “Disputed” (second two rows of Table 1), and a specific warning condition in which they were instead labeled as “Rated false” (last two rows of Table 1).

Table 1: Experimental conditions

Tag	General warning	N
None	No	469
None	Yes	424
“Disputed”	No	413
“Disputed”	Yes	429
“Rated false”	No	429
“Rated false”	Yes	397
Pure control		433

The study proceeded as follows. Once participants consented to participate, they answered a series of demographic questions, followed by questions about their use of social media, political preferences, voting behavior, and trust in fact-checking and the media. Participants were then asked to rate the accuracy of several real and fabricated political statements to test their predisposition

to hold political misperceptions.<sup>6</sup> Afterward, they answered a political knowledge battery, which provided a buffer between the misperception items and the experimental task.<sup>7</sup>

In the general warning condition, participants were shown a message warning them about misleading articles and providing advice for identifying false information (see Online Appendix A for exact wording and design). The design of the general warning was chosen to resemble Facebook's false news message to users (Mosseri 2017). Participants in the no-warning conditions were shown an identical image with innocuous instructions to eliminate any potential confounding effects. In the pure control group, respondents were exposed to no images, no articles, no general warning, no tags, and no headlines, and proceeded directly to the questions measuring the outcome variable (discussed in the next section).

Each participant who was not assigned to the pure control group was shown nine selected political headlines formatted as they would appear on Facebook in random order: three false pro-Trump headlines, three false anti-Trump headlines, and three true headlines (see Online Appendix A for the exact stimuli used). Each participant saw each of the nine headlines, with the appearance of the headlines (i.e., whether they included "Disputed" or "Rated false" tags) randomly varying based on treatment condition. We selected a balance of pro- and anti-Trump false news articles from Snopes and BuzzFeed, excluding those related to the 2016 election that had become less relevant.<sup>8</sup> True political headlines from mainstream media sources were also included so that the veracity of head-

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<sup>6</sup>As in most studies, we cannot know how much false news respondents were exposed to during the 2016 presidential election and its aftermath (e.g., Allcott and Gentzkow 2017). While it would be useful to measure this quantity, our main interest is the effect of warnings and tags on belief accuracy when they encounter false news. In addition, the auxiliary measure of misperception belief mentioned above does allow us to test whether individuals who are susceptible to believing false news respond differently to warnings and tags than those who are not. We find no consistent evidence of such heterogeneity in exploratory analyses reported in Online Appendix C. Scholars should collect data on individuals' exposure to false news and explore treatment effect heterogeneity by this variable directly in future research.

<sup>7</sup>A possible concern is that asking respondents to rate political statements for accuracy could have primed them to be particularly alert to clues that the treatment articles could be deceptive in nature. However, Pennycook, Cannon, and Rand (2017) and Pennycook and Rand (2017) did not ask respondents to rate any statements for accuracy before their experiment and also found that tagged false news headlines were rated as less accurate than untagged ones, suggesting that the tags reduce the perceived accuracy of false headlines independently of a possible priming effect.

<sup>8</sup>Some of these articles were originally used in Pennycook, Cannon, and Rand (2017), which examined the effect of prior exposure to false news headlines on the perceived accuracy of false news. Others were taken from Silverman (2016), a compilation of the most widely shared false news articles during the 2016 election. The original sources of the false news articles were dubious websites that had intentionally created them for profit.

Table 2: Headlines displayed in survey

Headline	Source	Type
Trump questions why U.S. Civil War had to happen	Reuters	True
Trump Orders Airstrikes in Syria After Chemical Attack	CBS New York	True
Neil Gorsuch Confirmed to Supreme Court	CNN	True
Trump on Revamping the Military: “We’re Bringing Back the Draft”	Real News Right Now	False, anti-Trump
Trump Plagiarized the Bee Movie for Inaugural Speech	Daily Kos	False, anti-Trump
FBI Discovers Kremlin is blackmailing Jason Chaffetz over Donald Trump and Russia	Palmer Report	False, anti-Trump
“Donald Trump Protester Speaks Out: ‘I was paid \$3,500 to protest Trump’s rally’”	ABCnews.com.co	False, Pro-Trump
Donald Trump Sent His Own Plane to Transport 200 Stranded Marines	Top Rated Viral	False, pro-Trump
FBI Agent Suspected in Hillary Email Leaks Found Dead in Apparent Murder-Suicide	Alexander Higgins	False, pro-Trump

In the study, the Chaffetz headline identified him as a “Republican Congressman.”

lines was not uniform.<sup>9</sup> Finally, though Pennycook and Rand (2018b) find that news sources do not significantly affect belief in the perceived accuracy of false news headlines (see also Clayton et al. 2018), we purposefully omitted news sources (and authors) to minimize potentially confounding variables and isolate the effects of warnings and tags on belief in false news headlines.

In the disputed condition, two randomly chosen pro-Trump and two anti-Trump false news headlines were tagged as “Disputed by Snopes.com and PolitiFact” (see Online Appendix A for headline format). Similarly, in the false condition, two randomly chosen pro-Trump and two anti-Trump false news headlines were tagged as “Rated false by Snopes.com and PolitiFact.” The wording and format of these tags were chosen to resemble warnings implemented by Facebook. Tags were distributed evenly to pro-Trump and anti-Trump headlines (i.e., two of each). Finally, the two remaining false headlines (one pro-Trump and one anti-Trump) were not tagged. This distribution allows us to test the effects of political congeniality while also simulating a typical news feed in which not all false news stories will be fact-checked.

<sup>9</sup>The true headlines that were tested were taken from actual mainstream news sources and were not intended to be explicitly pro- or anti-Trump, though respondent interpretations of them may differ.

After each headline was displayed, participants were asked to evaluate the accuracy of the headline and to self-report how likely they would be to “like” and share the story on Facebook (see Online Appendix A for wording).

## Measures

To test the perceived accuracy of the claims in false news headlines, participants were asked to evaluate the accuracy of each claim on a four-point Likert scale from “Not at all accurate” (1) to “Very accurate” (4).<sup>10</sup> This question format is a common approach in recent studies measuring participants’ belief in misinformation and false news (e.g., Clayton et al. 2018; Kuru, Pasek, and Traugott 2017; Pennycook, Cannon, and Rand 2017; Pennycook and Rand 2017, 2018a,b); employing it allows us to directly compare our results with the existing literature. We reasoned that a unipolar four-point scale allowed respondents to express a more nuanced assessment of a statement’s accuracy than, for example, a binary true/false question (e.g., by choosing “somewhat accurate” versus “very accurate”). This question format was also appropriate for the pure control group, which did not view a headline but could be asked to assess each claim’s general accuracy. Finally, we elected not to include a “don’t know” option and instead permitted respondents to skip questions in the survey.<sup>11</sup> Summary measures of respondent belief in each of the nine false news headlines included in our survey are provided in Online Appendix C (see Table C2).<sup>12</sup>

We also asked participants who indicated they use Facebook in a pre-treatment measure about

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<sup>10</sup>A potential concern is that highly attentive MTurk respondents saw these accuracy questions as an attention check rather than a measure of sincere belief and responded accordingly. However, previous research has found that the effect of corrections to misinformation were almost identical among samples of MTurk workers and Morning Consult poll respondents (Nyhan et al. 2017) and provides limited and inconsistent evidence of demand effects in survey experiments (Mummolo and Peterson 2018).

<sup>11</sup>de Leeuw, Hox, and Boevé (2015) find that excluding “don’t know” options but allowing respondents to skip questions in online surveys (as we did) reduces missing data and increases reliability in online surveys relative to the inclusion of a “don’t know” option, and suggest using “don’t know” options only when there is a theoretical reason to do so. We also opt to exclude the “don’t know” option to encourage compatibility between our study and others in the field that examine belief in false news and other forms of political misinformation (e.g., Pennycook, Cannon, and Rand 2017; Pennycook and Rand 2017).

<sup>12</sup>Our preregistration did not offer hypotheses about the correlates of false news belief, but see Pennycook and Rand (2018b), which finds that individuals who have a tendency to ascribe profundity to randomly generated sentences and who overstate their level of knowledge are more likely to perceive false news as accurate. Those who engage in analytic thinking are less susceptible.

their willingness to “like” or share a given headline on Facebook on a scale from “Not at all likely” (1) to “Very likely” (4) (see Online Appendix B for results). Finally, because our headlines included a mixture of pro- and anti- Trump stories, we measured respondents’ approval of President Trump prior to the manipulation on a scale from “Strongly disapprove” (1) to “Strongly approve” (4) and classified those who “strongly” or “somewhat” approve of him as approvers and those who “strongly” or “somewhat” disapprove as disapprovers.

## Results

We analyze the effects of our experiment using OLS with robust standard errors.<sup>13</sup> All analyses were pre-registered in advance on EGAP unless otherwise specified. Replication data and code are available on the *Political Behavior* Dataverse (<https://dataverse.harvard.edu/dataverse/polbehavior>).

Our primary outcome measure is the perceived accuracy of false headlines, which we pooled across the false news headlines that respondents evaluated. The statistical analyses below include standard errors clustered by respondent and question fixed effects as well as indicators for exposure to a general warning about false news and whether the respondent saw a “Disputed” or “Rated false” tag under the headline in question. Importantly, our pre-registered specifications exclude responses to untagged headlines by respondents in the disputed or false conditions. We thus compare responses to tagged headlines in the disputed and false conditions to responses to headlines in the condition in which no tags were shown. All experimental treatment effects were estimated as intent to treat effects.<sup>14</sup>

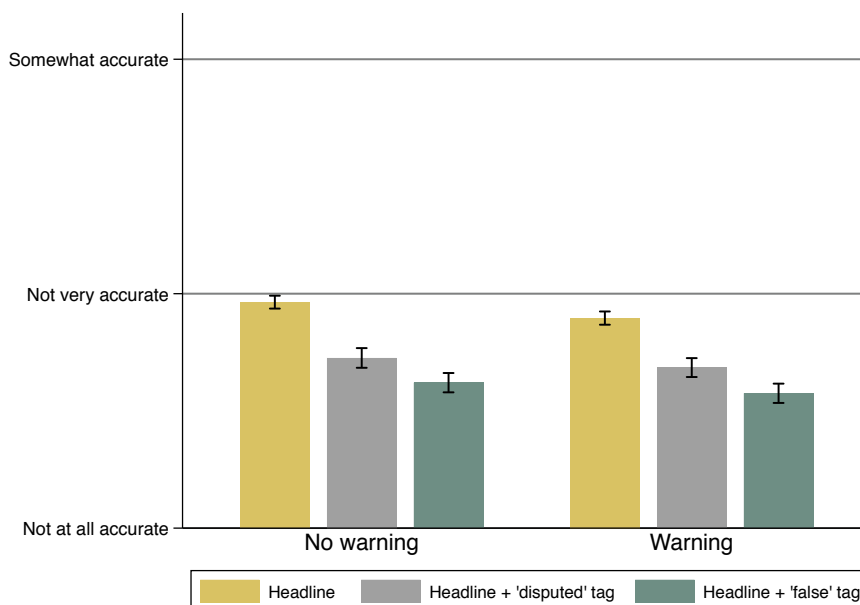
Since we could not estimate respondents’ likelihood to “like” or share headlines (results in Online Appendix B) in our pure control condition, we deviate from our pre-registration for expositional reasons to present estimates below that exclude the pure control group. As a result, we focus on the

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<sup>13</sup>All results are virtually identical when estimated using ordered probit instead. See Online Appendix C.

<sup>14</sup>We do not include respondent fixed effects, which were incorrectly specified in the pre-registration (they cannot be estimated due to multicollinearity). However, we show in Online Appendix C that our primary results are consistent when estimated in a model that includes random effects by respondent.

Figure 1: Effects of general and specific warnings on the perceived accuracy of false headlines



Mean belief that false headlines were accurate on a four-point Likert scale from “Not at all accurate” (1) to “Very accurate” (4). See Online Appendix A for question wording and stimulus materials.

sample of 2,561 respondents assigned to our experimental conditions. Accordingly, the baseline in our statistical models is the group that did not receive a general warning about false news or any “Disputed” or “Rated false” tags on headlines. Our estimates are substantively identical when the pure control group is used as the baseline condition instead (see Online Appendix C).

### Effects of general and specific warnings

Figure 1 summarizes the mean perceived accuracy of the false news headlines by whether respondents received a general warning about misleading articles and/or whether the headline was identified as “Disputed” or “Rated false” by fact-checkers. As the figure indicates, a general warning only slightly decreased the perceived accuracy of false headlines in the untagged headlines condition, reducing it from 1.96 to 1.90 on our four-point Likert scale.<sup>15</sup> The perceived accuracy of

<sup>15</sup>The estimates reported here refer to the effects of each treatment alone independent of any moderators, with all other manipulations set at 0. We estimate models that include interactive terms below.

false headlines declined more when specific warnings were provided, decreasing from 1.96 to 1.73 when a “Disputed” tag appeared and 1.62 when a “Rated false” tag appeared. This effect was nearly identical when a general warning was previously provided. In substantive terms, the “Disputed” tag reduced the mean proportion of respondents who accept a headline as “Somewhat accurate” or “Very accurate” when no general warning was provided from 29% in the baseline condition to 19%, a ten-percentage point decline (95% CI in a regression using the binary accuracy measure described above as the outcome:  $-7$  to  $-13$  percentage points). Similarly, the “Rated false” tag reduced the proportion of respondents who accepted the headline as accurate to 16%, a 13-percentage point decline from the baseline condition (95% CI:  $-11$  to  $-17$  percentage points). These effects are larger than those reported in Pennycook and Rand (2017), who find that a “Disputed” tag reduces perceived accuracy by 3.7 percentage points (a point estimate that is outside our 95% confidence interval).

We test our hypotheses and research questions more formally in Table 3, which shows the results of our pooled regression models for the perceived accuracy of false news headlines. Our results largely support the hypotheses that warnings reduce belief in false information. Consistent with H1, average belief in false headlines was slightly lower for participants who saw a general warning before seeing headlines than for participants who saw headlines with no warning ( $-0.08$ ;  $p < .05$ ). However, the substantive magnitude of this reduction in perceived belief accuracy is small (Cohen’s  $d = 0.08$ ).<sup>16</sup>

The negative effect of tags on perceived accuracy was stronger, however. Average perceived accuracy for participants who saw a headline with a “Disputed” tag was 0.24 points lower on our four-point scale than for participants who saw no tag ( $p < .01$ ; Cohen’s  $d = 0.26$ ) and 0.34 points lower for those who saw a “Rated false” tag than for those who saw no tag ( $p < .01$ ; Cohen’s  $d = 0.38$ ), supporting H2a and H2b. Most notably, “Rated false” tags were significantly more effective than “Disputed” tags at reducing belief in false information relative to a no tag condition ( $-0.11$ ,  $p < .01$ ), supporting H2c and suggesting that the effect of specific warnings is greater when they

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<sup>16</sup>The effects on perceived accuracy reported in Tables 3–5 are consistent when non-Facebook users are excluded from the sample in exploratory analyses (see Online Appendix C).

Table 3: Experimental effects on perceived accuracy of false headlines

	Accuracy
General warning	-0.08** (0.03)
“Disputed” tag	-0.24*** (0.04)
“Disputed” × warning	0.04 (0.05)
“Rated false” tag	-0.34*** (0.04)
“Rated false” × warning	0.03 (0.05)
Constant	1.85*** (0.03)
Question fixed effects	Yes
N (responses)	11962
Respondents	2554

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

clearly indicate that a headline is false.

We find no support for H3. Our results do not allow us to reject the null hypothesis of no difference in the effect of the “Disputed” and “Rated false” tags when a general warning is present compared to when it was not. The marginal effects of the tags remain negative and statistically significant when a general warning was previously provided, however (disputed:  $-0.20$ ,  $p < .01$ ; false:  $-0.31$ ,  $p < .01$ ). We therefore conclude that a general warning did not augment the effect of tags warning about specific misleading articles. Participants view specific warning tags immediately before answering questions about a headline’s accuracy, so the tag’s effect may be more immediate and take precedence over that of the general warning shown earlier.



## Differences by article slant

We next test whether these effects vary depending on whether the general or specific warnings provided to participants are politically congenial or uncongenial. To do so, we add a directional preference measure (an indicator for Trump approval) and corresponding interaction terms to our previous statistical model predicting the perceived accuracy of false news headlines. The coefficients of interest are presented in Table 4, which separately estimates results for false pro- and anti-Trump articles. These results are estimated only among respondents who approve or disapprove of Trump.<sup>17</sup>

Though Trump job approval is strongly associated with baseline levels of belief in the false headlines, our findings do not support our hypothesis that people would resist warnings about politically congenial false news (H4). We also find no difference in the effect of a general warning, a research question for which we had weaker expectations because it does not challenge a specific story like a fact-checking tag (RQ1). For pro-Trump stories, we are unable to reject the null hypothesis that the effects of the “Disputed” or “Rated false” tags do not differ between Trump approvers and disapprovers. We also find no measurable difference in the effect of the “Disputed” tags by Trump approval for anti-Trump stories. One possible explanation for these findings is that the headlines we tested were likely unfamiliar to many respondents. As a result, they may have been less connected to respondents’ personal beliefs and thus more easily debunked than well-known misperceptions.<sup>18</sup>

In one case, the effect of a “Rated false” tag did significantly differ between Trump approvers and disapprovers for anti-Trump stories, but the sign of the effect was the opposite of the hypothesized direction. The presence of a “Rated false” tag reduced the perceived accuracy of anti-Trump headlines significantly *more* among Trump disapprovers for whom the headlines were politically

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<sup>17</sup>A typo in the pre-registration statement to this effect instead mistakenly stated we would exclude “pure independents.” The results below again exclude pure controls but equivalent results including those respondents are provided in Online Appendix C. We do not include respondents with no opinion of Trump in that model because there were so few ( $n = 4$ ).

<sup>18</sup>Pennycook and Rand (2018a) similarly find that “the correlation between CRT [Cognitive Reflection Test scores] and perceived accuracy is unrelated to how closely the headline aligns with the participant’s ideology... Our findings therefore suggest that susceptibility to fake news is driven more by lazy thinking than it is by partisan bias per se.” Similarly, Porter, Wood, and Kirby (2018) find minimal differences between ideological groups in their willingness to accept false news headlines.

Table 4: Experimental effects on perceived accuracy of false headlines by article slant

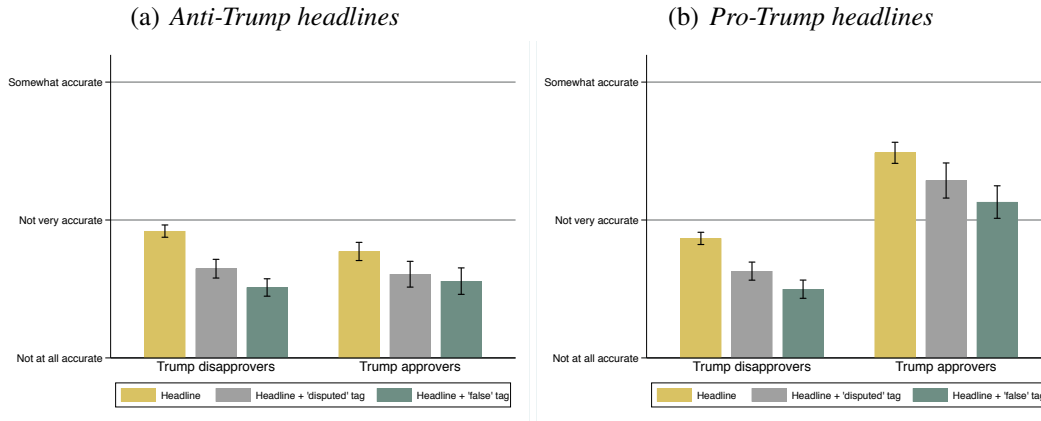
	Anti-Trump	Pro-Trump
General warning	-0.11** (0.05)	-0.06 (0.05)
“Disputed” tag	-0.28*** (0.05)	-0.24*** (0.05)
“Rated false” tag	-0.41*** (0.05)	-0.37*** (0.05)
Trump approval	-0.18*** (0.07)	0.64*** (0.07)
Warning × Trump approval	-0.04 (0.09)	0.06 (0.10)
“Disputed” × warning	0.10 (0.07)	0.04 (0.07)
“Disputed” × Trump approval	0.14 (0.09)	0.03 (0.10)
“Disputed” × warning × Trump approval	-0.05 (0.13)	-0.12 (0.15)
“Rated false” × warning	0.12* (0.07)	0.01 (0.07)
“Rated false” × Trump approval	0.23** (0.10)	-0.01 (0.10)
“Rated false” × warning × Trump approval	-0.06 (0.13)	-0.07 (0.15)
Constant	1.91*** (0.04)	1.99*** (0.04)
Question fixed effects	Yes	Yes
N (responses)	5972	5972
Respondents	2550	2548

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

congenial than among Trump approvers for whom they were uncongenial (0.20,  $p < .05$ ). This counterintuitive finding may be the result of differing levels of trust in fact-checking and false news susceptibility. An exploratory analysis using pre-treatment measures shows that Trump approvers were much more likely to believe in election 2016 false news (an average of 0.26 points higher on a four-point accuracy scale,  $p < .01$ ) and to distrust fact-checkers (an average of 0.53 points lower on a four-point trust scale,  $p < .01$ ) than disapprovers.

These results can be observed in Figure 2, which displays the mean level of perceived accuracy

Figure 2: Specific warning effects by political congeniality of false headlines



Mean belief that false headlines were accurate on a four-point Likert scale from “Not at all accurate” (1) to “Very accurate” (4). See Online Appendix A for question wording and stimulus materials. Excludes respondents assigned to receive a general warning as well as those in the pure control condition.

by condition for both anti- and pro-Trump headlines among respondents who did not receive a general warning. In general, the reduction in belief is similar regardless of whether the headline is politically congenial. The exception was anti-Trump headlines, which Trump disapprovers rate as more accurate than approvers when no tag is present, but view as less accurate when a “Rated false” tag accompanies the headline.<sup>19</sup>

## Spillover effects

Finally, we consider our last two research questions, which concern possible unintended spillover effects from general or specific warnings. We first test whether the presence of “Disputed” or “Rated false” tags affects belief in untagged false or true news headlines (RQ3). In this case, the presence of tags could cause a contrast effect that leads participants to infer that other stories are accurate.

Alternatively, the presence of these tags could make participants more skeptical about all news. We

<sup>19</sup>We conducted an additional exploratory analysis to test whether the effects of political congeniality were altered by a participant’s political knowledge. Consistent with previous research, we found that high political knowledge was associated with a lower belief in false news stories regardless of the article’s slant. However, we did not find convincing evidence that high political knowledge meaningfully changed a specific warning’s effect on belief in false news headlines. Results for this exploratory analysis are included in Online Appendix C (Table C16).

Table 5: Experimental tests for spillover effects of warnings on perceived accuracy

	Untagged false headlines	True news headlines
General warning	-0.08** (0.03)	-0.12*** (0.04)
“Disputed” condition	0.02 (0.05)	0.06 (0.04)
“Rated false” condition	0.04 (0.04)	0.03 (0.04)
“Disputed” × warning	0.06 (0.06)	0.09 (0.06)
“Rated false” × warning	0.03 (0.06)	0.14** (0.06)
Constant	1.87*** (0.03)	2.87*** (0.03)
Question fixed effects	Yes	Yes
N (responses)	7968	6585
Respondents	2436	2502

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a “Disputed” or “Rated false” tag for the specific headline in question.

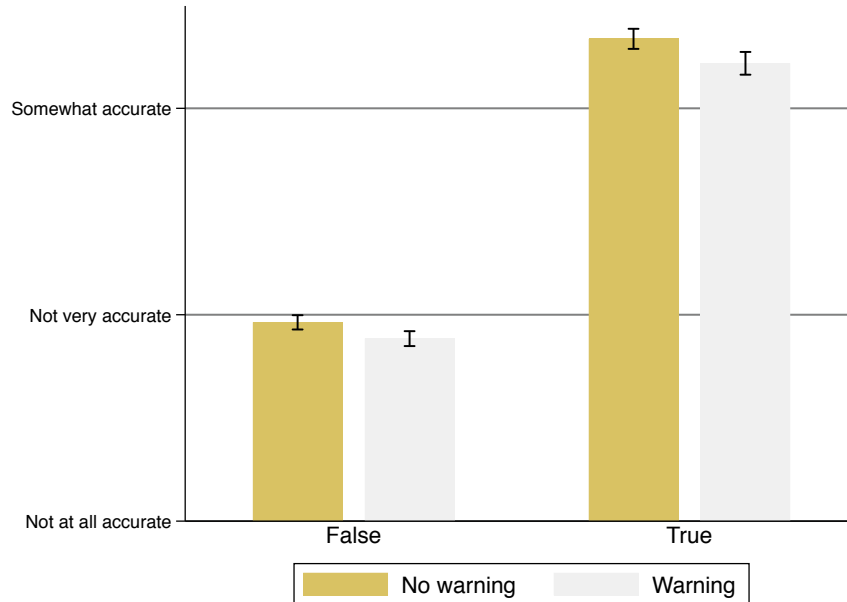
therefore test whether a general warning about false news affects the perceived accuracy of true news headlines (RQ3).

These research questions are evaluated in Table 5, which separately tests these effects on the perceived accuracy of both types of headlines. The models reported in the first column exclude headlines in which respondents in the “Disputed” or “Rated false” conditions saw a tag; they thus test the effect of assignment to those conditions on the perceived accuracy of *other* headlines.<sup>20</sup>

We find no statistically measurable effect on the perceived accuracy of untagged false articles when other false articles had a “Disputed” or “Rated false” tag (RQ2a). However, our point estimates for “Disputed” (0.02) and “Rated false” (0.04) are similar to the “implied truth” effect found by Pennycook and Rand (2017) (0.03); we simply lack the precision to detect an effect of that mag-

<sup>20</sup>Headlines viewed by respondents in the “Disputed” or “Rated false” conditions before exposure to the first tag are also excluded (spillover is impossible for participants who are not yet treated).

Figure 3: General warning effects on belief in true and false news articles



Mean belief that true and false news headlines were accurate on a four-point Likert scale from “Not at all accurate” (1) to “Very accurate” (4). See Online Appendix A for question wording and stimulus materials.

nitide (95% CI for “Disputed”: -0.07, 0.11; 95% CI for “Rated false”: -0.05, 0.16). The perceived accuracy of true articles was also unaffected when other false articles included these tags (RQ2b).

However, the general warning had an unintended spillover effect on the perceived accuracy of true headlines ( $-0.12, p < .01$ ), though the substantive magnitude of this effect was small (Cohen’s  $d = 0.12$ ). While overall levels of belief in false news stories were much lower than for true ones, the decrease in perceived accuracy from the general warning was greater with true stories ( $-0.12$ ) than false ones ( $-0.08$ ), though an exploratory analysis pooling evaluations of true and false headlines shows we cannot reject the null hypothesis of no difference in perceived accuracy between them. The findings suggest that the specific warnings were more effective because they reduced belief solely for false headlines and did not create spillover effects on perceived accuracy of true news.

## Conclusion

This study provides several important new findings about how to most effectively counter false information on social media. First, both “Disputed” and “Rated false” tags modestly reduce belief in false news. Notably, we find larger accuracy effects for the “Disputed” tags than Pennycook and Rand 2017.<sup>21</sup> However, our results demonstrate that “Rated false” tags, which specifically tell users when claims made in headlines are untrue, are more effective at reducing belief in misinformation than the “Disputed” tags previously used by Facebook. Encouragingly, we find no consistent evidence that the effects of these tags varies by the political congeniality of the headlines or that exposure to the tags increases the perceived accuracy of unlabeled false headlines (though our study lacks the precision necessary to detect the small “implied truth” effect that Pennycook and Rand 2017 identify).

By contrast, though general warnings about false news also appear to decrease belief in false headlines, the effect of a general warning is small compared to either type of tag. Moreover, general warnings also reduce belief in real news and do not enhance the effects of the “Rated false” and “Disputed” tags, suggesting that they are a less effective approach.

Our results provide support for prior studies finding a negative effect of general warnings on belief in misinformation (Bolsen and Druckman 2015; Ecker, Lewandowsky, and Tang 2010; van der Linden et al. 2017), but our finding that these warnings also reduce the perceived accuracy of true headlines suggest that they pose a potential hazard. False news may already increase distrust in legitimate information; unintended spillover effects from general warnings or related proposals to fight false information by increasing media literacy (e.g., Atkins 2017) could exacerbate this problem. Our “Disputed” and “Rated false” tags, which more effectively reduce the perceived accuracy of false headlines without causing these spillover effects, may be a safer way to reduce belief in misinformation.

Further research is needed to evaluate this finding and better understand the mechanism for the

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<sup>21</sup>This difference in effect size could be partially attributable to respondents being aware that their ability to discern true from false headlines was under scrutiny, since they had previously been asked to rate political statements as true or false at the beginning of our survey.

spillover effect we observe. One potential explanation is that warnings about false news prime people to think about misleading information online, making them less likely to trust any articles they see on social media. Another possible interpretation is that we are observing a “tainted truth” effect in the context of political misinformation. In social cognition research, such an effect occurs when eyewitnesses who are warned about the influence of misinformation overcorrect for this threat and identify fewer true items than eyewitnesses who are not warned (Echterhoff et al. 2007; Szpitalak and Polczyk 2010). Our results suggest that the tainted truth effect could apply to prospective warnings about political misinformation, but further research is necessary to test if this finding holds in other contexts and designs.

Our study has several important limitations that should be addressed in future research. First, our sample was more educated and politically active than the general population and leaned liberal; future studies should be conducted with nationally representative sample. A second limitation is that our study examined the effect of warnings and tags on pro- and anti-Trump headlines in the aftermath of a presidential election — a substantively important but specific context. While Trump-related headlines remain timely and salient, further research should seek to determine whether our results hold in other contexts and for other types of false headlines. Third, we do not examine over-time effects. Future studies should evaluate long-term belief in false news after the initial exposure and how our manipulations affect those beliefs. Fourth, as noted above, our design does not allow us to identify the causal mechanisms responsible for the effects we observed — a challenge facing nearly all experimental studies (Bullock, Green, and Ha 2010). In particular, future research should employ designs that provide more leverage for understanding the effects of warnings on belief in false news. Finally, as in any experimental study, we cannot fully rule out the possibility of demand effects. Any survey that asks about the perceived accuracy of political statements and the effects of interventions on those self-reports is susceptible to these effects, though research suggests they are rare (Mummolo and Peterson 2018).

Our study also made a number of design choices that should be revisited in future research. First, we focused on belief in headlines because they are prominently displayed on social media, but future

studies should also measure the effects of warnings and tags on belief when people actually read the articles in question. Second, false news articles typically appear on a user’s timeline because a friend liked or shared the article, but we chose not to test the effect of social endorsements or other contextual cues on belief in false news articles. Future research should use field experiments or conduct studies in other settings to evaluate the extent to which our findings generalize to real-world contexts. Third, our headlines omitted article sources to allow us to isolate the effect of our treatments, but these sources are likely to play a role in how individuals evaluate Facebook posts. Exploring the interactions between source credibility and warnings on belief in misinformation is another important avenue for future research (though see Pennycook and Rand 2018b and Clayton et al. 2018, who both find that source cues may play a limited role in credibility assessments of true and false news). Finally, we chose to use the same two fact-checking sources throughout the study, PolitiFact and Snopes. However, people differ in how much they trust the most prominent national fact-checking organizations (e.g., Nyhan and Reifler N.d.). Future studies should vary the source of fact-checks in order to determine whether the fact-checking source influences individuals’ perceptions of true and false headlines.

Despite these limitations, this study provides important insights into how efforts to prevent belief in misinformation on social media could be more effective and suggests that online false news can be countered with some degree of success.

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## Online Appendix A

This study is being conducted by Brendan Nyhan, a professor at Dartmouth College. We ask for your attention for a few minutes and we thank you for your attention and your responses. Your participation is voluntary and you may decline the survey or withdraw at any time. No information that identifies you will be collected or retained by the researchers. However, any online interaction carries some risk of being accessed. Do you consent to participate in the survey?

-Yes

-No

[Demographics]

How old are you?

-Under 18

-18 - 24

-25 - 34

-35 - 44

-45 - 54

-55 - 64

-65 - 74

-75 - 84

-85 or older

In what state do you currently reside?

-Alabama

-Alaska

-Arizona

-Arkansas

-California

-Colorado

-Connecticut

-Delaware

-District of Columbia

-Florida

-Georgia

-Hawaii

-Idaho

-Illinois

-Indiana

-Iowa

-Kansas

-Kentucky

-Louisiana

-Maine

-Maryland

- Massachusetts
- Michigan
- Minnesota
- Mississippi
- Missouri
- Montana
- Nebraska
- Nevada
- New Hampshire
- New Jersey
- New Mexico
- New York
- North Carolina
- North Dakota
- Ohio
- Oklahoma
- Oregon
- Pennsylvania
- Rhode Island
- South Carolina
- South Dakota
- Tennessee
- Texas
- Utah
- Vermont
- Virginia
- Washington
- West Virginia
- Wisconsin
- Wyoming
- Puerto Rico
- I do not reside in the United States

What is your gender?

- Male
- Female
- Other

When it comes to politics, would you describe yourself as liberal, conservative, or neither liberal nor conservative?

- Very conservative
- Somewhat conservative
- Slightly conservative
- Moderate; middle of the road
- Slightly liberal

- Somewhat liberal
- Very liberal

Generally speaking, do you usually think of yourself as a Republican, a Democrat, an Independent, or something else?

- Republican
- Democrat
- Independent
- Something else

[If Democrat is selected]

Would you call yourself a strong Democrat or a not very strong Democrat?

- Strong Democrat
- Not very strong Democrat

[If Republican is selected]

Would you call yourself a strong Republican or not a very strong Republican?

- Strong Republican
- Not very strong Republican

[If Independent or Something else is selected]

Do you think of yourself as closer to the Republican Party or to the Democratic Party?

- Closer to the Republican Party
- Closer to the Democratic Party
- Neither

What is the highest degree or level of school you have completed?

- Did not graduate from high school
- High school diploma or the equivalent (GED)
- Some college
- Associate's degree
- Bachelor's degree
- Master's degree
- Professional or doctorate degree

Please check one or more categories below to indicate what race(s) you consider yourself to be.

- White
- Black or African American
- American Indian or Alaska Native
- Asian/Pacific Islander
- Multi-racial
- Other

Are you of Spanish or Hispanic origin or descent?

- Yes
- No

Generally, how interested are you in politics?

- Extremely interested
- Very interested
- Somewhat interested
- Not very interested
- Not at all interested

How frequently do you use Facebook?

- Daily
- A few times a week
- Once a week
- A few times a month
- Once a month
- Less frequently than once a month
- Never

[If Never is not selected]

How frequently do you share political news stories on Facebook?

- Daily
- A few times a week
- Once a week
- A few times a month
- Once a month
- Less frequently than once a month
- Never

In talking to people about elections, we often find that a lot of people were not able to vote because they weren't registered, they were sick, or they just didn't have time. Which of the following statements best describes you?

- I did not vote in the election this November
- I thought about voting this time, but didn't
- I usually vote, but didn't this time
- I am sure I voted

[If I am sure I voted is selected]

Who did you vote for in the presidential election?

- Hillary Clinton
- Donald Trump
- Gary Johnson
- Jill Stein
- Other candidate

Do you approve or disapprove of the way Donald Trump is handling his job as President?

- Strongly approve (4)
- Somewhat approve (3)

- Somewhat disapprove (2)
- Strongly disapprove (1)

When fact-checking organizations like Snopes and PolitiFact evaluate the accuracy of claims made online, how much do you trust the information they provide?

- A great deal (4)
- A moderate amount (3)
- Not much (2)
- Not at all (1)

In general, how much trust and confidence do you have in the mass media - such as newspapers, TV, and radio - when it comes to reporting the news fully, accurately, and fairly?

- A great deal (4)
- A moderate amount (3)
- Not much (2)
- Not at all (1)

[Political knowledge and prior misperceptions]

The next set of questions helps us learn what types of information are commonly known to the public. Please answer these questions on your own without asking anyone or looking up the answers. Many people don't know the answers to these questions, but we'd be grateful if you would please answer every question even if you're not sure what the right answer is. It is important to us that you do NOT use outside sources like the Internet to search for the correct answer. Will you answer the following questions without help from outside sources?

- Yes
- No

To the best of your knowledge, how accurate are the following statements?

Pope Francis endorsed Donald Trump for president.

- Not at all accurate
- Not very accurate
- Somewhat accurate
- Very accurate

As Secretary of State, Hillary Clinton arranged for the United States to sell weapons to the jihadist militant group known as ISIS.

- Not at all accurate
- Not very accurate
- Somewhat accurate
- Very accurate

In May 2016, Ireland announced that it was officially accepting Americans requesting political asylum from a Donald Trump presidency.

- Not at all accurate

- Not very accurate
- Somewhat accurate
- Very accurate

FBI Director James Comey put a Trump sign on his front lawn.

- Not at all accurate
- Not very accurate
- Somewhat accurate
- Very accurate

Hillary Clinton won the popular vote in the 2016 election.

- Not at all accurate
- Not very accurate
- Somewhat accurate
- Very accurate

The FBI concluded that Hillary Clinton and her associates in the Department of State were "extremely careless in their handling of very sensitive, highly classified information."

- Not at all accurate
- Not very accurate
- Somewhat accurate
- Very accurate

Supreme Court Justice Antonin Scalia died in 2016.

- Not at all accurate
- Not very accurate
- Somewhat accurate
- Very accurate

The website WikiLeaks published numerous hacked emails from Hillary Clinton's campaign.

- Not at all accurate
- Not very accurate
- Somewhat accurate
- Very accurate

For how many years is a United States Senator elected - that is, how many years are there in one full term of office for a U.S. Senator?

- Two years
- Four years
- Six years
- Eight years
- None of these
- Don't know

How many times can an individual be elected President of the United States under current laws?

- Once

- Twice
- Four times
- Unlimited number of terms
- Don't know

How many U.S. Senators are there from each state?

- One
- Two
- Four
- Depends on which state
- Don't know

Who is currently the Prime Minister of the United Kingdom?

- Richard Branson
- Nick Clegg
- David Cameron
- Theresa May
- Margaret Thatcher
- Don't know

For how many years is a member of the United States House of Representatives elected - that is, how many years are there in one full term of office for a U.S. House member?

- Two years
- Four years
- Six years
- Eight years
- For life
- Don't know

[Pure control outcome measures (order randomized)]

To the best of your knowledge, how accurate is the claim that Trump questioned why the U.S. Civil War had to happen?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

To the best of your knowledge, how accurate is the claim that Trump ordered airstrikes in Syria after a chemical attack?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

To the best of your knowledge, how accurate is the claim that Neil Gorsuch was confirmed to the

Supreme Court?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

To the best of your knowledge, how accurate is the claim that Trump is bringing back the draft?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

To the best of your knowledge, how accurate is the claim that Trump plagiarized the Bee Movie for his inaugural speech?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

To the best of your knowledge, how accurate is the claim that Republican Congressman Jason Chafetz was blackmailed by the Kremlin?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

To the best of your knowledge, how accurate is the claim that a Donald Trump protester was paid \$3,500 to protest Trump's rally?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

To the best of your knowledge, how accurate is the claim that Donald Trump sent his own plane to transport 200 stranded Marines?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

To the best of your knowledge, how accurate is the claim that an FBI agent suspected in Hillary Clinton's email leaks was found dead in an apparent murder-suicide?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)



[warning manipulation (randomized; not shown to pure controls)]



### **How to spot misleading articles**

You will be asked to evaluate the accuracy of some news headlines shared on social media.

Although some of these stories may be true, others may be misleading. Some news stories can be politically motivated and use misleading tactics and statements to try to convince the public that they are true. It is important to remain skeptical when reading headlines and to think carefully about whether a story could be true in order to help stop the spread of misleading articles.

[no-warning image (randomized; not shown to pure controls)]



You will be asked to evaluate the accuracy of some news headlines shared on social media.

Please read the following headlines and answer the questions that follow.

[tag conditions (headline order randomized; not shown to pure controls):

-no tags: no tags were shown on any headlines

-disputed: random subset of four false headlines were shown with “Disputed” tag

-false: random subset of four of six false headlines were shown with “Rated false” tag]




Trump on Revamping the Military: We're Bringing Back the Draft

Trump unveiled his plan to 'make the military great again,' saying he intends to reinstate the draft as part of a larger effort to bolster the armed forces.



Trump on Revamping the Military: We're Bringing Back the Draft

Trump unveiled his plan to 'make the military great again,' saying he intends to reinstate the draft as part of a larger effort to bolster the armed forces.

 Disputed by Snopes.com and PolitiFact



Trump on Revamping the Military: We're Bringing Back the Draft

Trump unveiled his plan to 'make the military great again,' saying he intends to reinstate the draft as part of a larger effort to bolster the armed forces.

 Rated False by Snopes.com and PolitiFact

To the best of your knowledge, how accurate is the claim that Trump is bringing back the draft?

-Not at all accurate (1)

-Not very accurate (2)

-Somewhat accurate (3)

-Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to “like” this story on Facebook?

-Very likely (4)

-Somewhat likely (3)

-Not very likely (2)

-Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

-Very likely (4)

-Somewhat likely (3)

-Not very likely (2)

-Not at all likely (1)

Donald Trump Inaugural Address (2017)

Barry B. Benson, Bee Movie (2007)

We are one nation -- and their pain is our pain. Their dreams are our dreams;

We are one colony -- and their pain is our pain. Their dreams are our dreams;

**Trump Plagiarized The Bee Movie for Inaugural Speech**

Back during the convention when Melania plagiarized Michelle Obama, the RNC tried to say her inspiration was not the First Lady but rather My Little Pony So...

Donald Trump Inaugural Address (2017)

Barry B. Benson, Bee Movie (2007)

We are one nation -- and their pain is our pain. Their dreams are our dreams;

We are one colony -- and their pain is our pain. Their dreams are our dreams;

**Trump Plagiarised The Bee Movie for Inaugural Speech**

Back during the convention when Melania plagiarized Michelle Obama, the RNC tried to say her inspiration was not the First Lady but rather My Little Pony So...

**Disputed by Snopes.com and PolitiFact**

Donald Trump Inaugural Address (2017)

Barry B. Benson, Bee Movie (2007)

We are one nation -- and their pain is our pain. Their dreams are our dreams;

We are one colony -- and their pain is our pain. Their dreams are our dreams;

**Trump Plagiarized The Bee Movie for Inaugural Speech**

Back during the convention when Melania plagiarized Michelle Obama, the RNC tried to say her inspiration was not the First Lady but rather My Little Pony So...

**Rated False by Snopes.com and PolitiFact**

To the best of your knowledge, how accurate is the claim that Trump plagiarized the Bee Movie for his inaugural speech?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to “like” this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)



FBI discovers Kremlin is blackmailing Republican Congressman over Donald Trump and Russia  
This would explain why Chaffetz has been protecting Trump over Russia



FBI discovers Kremlin is blackmailing Republican Congressman over Donald Trump and Russia  
This would explain why Chaffetz has been protecting Trump over Russia

Disputed by Snopes.com and PolitiFact



FBI discovers Kremlin is blackmailing Republican Congressman over Donald Trump and Russia  
This would explain why Chaffetz has been protecting Trump over Russia

Rated False by Snopes.com and PolitiFact

To the best of your knowledge, how accurate is the claim that Republican Congressman Jason Chaffetz was blackmailed by the Kremlin?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to “like” this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)



Donald Trump Protester Speaks Out: "I Was Paid \$3,500 To Protest Trump's Rally"

PHOENIX A.Z. — For months now, rumors have circulated the Internet that individuals were being paid to protest at rallies held by presidential hopeful Donald Trump...



Donald Trump Protester Speaks Out: "I Was Paid \$3,500 To Protest Trump's Rally"


PHOENIX A.Z. — For months now, rumors have circulated the Internet that individuals were being paid to protest at rallies held by presidential hopeful Donald Trump...

 Disputed by Snopes.com and PolitiFact



Donald Trump Protester Speaks Out: "I Was Paid \$3,500 To Protest Trump's Rally"

PHOENIX A.Z. — For months now, rumors have circulated the Internet that individuals were being paid to protest at rallies held by presidential hopeful Donald Trump...

 Rated False by Snopes.com and PolitiFact

To the best of your knowledge, how accurate is the claim that a Donald Trump protester was paid \$3,500 to protest Trump's rally?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to "like" this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)




**Donald Trump Sent His Own Plane To Transport 200 Stranded Marines**

After hearing about 200 Marines left stranded after returning home from Operation Desert Storm back in 1991, Donald J. Trump came to the aid of those Marines by...



**Donald Trump Sent His Own Plane To Transport 200 Stranded Marines**

After hearing about 200 Marines left stranded after returning home from Operation Desert Storm back in 1991, Donald J. Trump came to the aid of those Marines by...

 Disputed by Snopes.com and PolitiFact



**Donald Trump Sent His Own Plane To Transport 200 Stranded Marines**

After hearing about 200 Marines left stranded after returning home from Operation Desert Storm back in 1991, Donald J. Trump came to the aid of those Marines by...

 Rated False by Snopes.com and PolitiFact

To the best of your knowledge, how accurate is the claim that Donald Trump sent his own plane to transport 200 stranded Marines?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to “like” this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)



FBI Agent Behind Hillary Email Leaks Found Dead In Murder Suicide

FBI Agent Suspected in Hillary Email Leaks Found Dead in Apparent Murder-Suicide. Archived here as this story will likely be removed or altered.



FBI Agent Behind Hillary Email Leaks Found Dead In Murder Suicide

FBI Agent Suspected in Hillary Email Leaks Found Dead in Apparent Murder-Suicide. Archived here as this story will likely be removed or altered.

 Disputed by Snopes.com and PolitiFact



FBI Agent Behind Hillary Email Leaks Found Dead In Murder Suicide

FBI Agent Suspected in Hillary Email Leaks Found Dead in Apparent Murder-Suicide. Archived here as this story will likely be removed or altered.

 Rated False by Snopes.com and PolitiFact

To the best of your knowledge, how accurate is the claim that an FBI agent suspected in Hillary Clinton's email leaks was found dead in an apparent murder-suicide?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to "like" this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)



### Trump questions why U.S. Civil War had to happen

Donald Trump has shown a fascination with populist 19th-century U.S. president ...

To the best of your knowledge, how accurate is the claim that Trump questioned why the U.S. Civil War had to happen?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to “like” this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)





### Trump Orders Military Airstrike Against Syria After Deadly Chemical Attack

Trump said that there is “no dispute” that Syria used banned chemical weapons...

To the best of your knowledge, how accurate is the claim that Trump ordered airstrikes in Syria after a chemical attack?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to “like” this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)



### Neil Gorsuch confirmed to the Supreme Court

The Senate Friday confirmed Neil Gorsuch, a 49-year-old federal judge who could help cement a conservative majority on the bench for decades, to the Supreme Court.

To the best of your knowledge, how accurate is the claim that Neil Gorsuch was confirmed to the Supreme Court?

- Not at all accurate (1)
- Not very accurate (2)
- Somewhat accurate (3)
- Very accurate (4)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to “like” this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

[If How frequently do you use Facebook? Never Is Not Selected]

How likely would you be to share this story on Facebook?

- Very likely (4)
- Somewhat likely (3)
- Not very likely (2)
- Not at all likely (1)

It is essential for the validity of this study that we know whether participants looked up any information online during the study. Did you make an effort to look up information during the study? Please be honest; you will still be paid and you will not be penalized in any way if you did.

-Yes, I looked up information

-No, I did not look up information

Do you have any comments on the survey? Please let us know about any problems you had or aspects of the survey that were confusing.

[textbox]

Thank you for answering these questions. The purpose of this study was to examine the believability of fake news, as well as the effectiveness of different forms of fact-checking. During this survey, participants were asked a series of questions about general information and their political opinions. After answering these questions, participants viewed a variety of headlines — both true and “fake” — in different formats and with various levels of fact-checking. The “fake” articles, while not rooted in fact, were all obtained from existing websites. Thank you again for your participation. Please do not share any information about the nature of this study with other potential participants. This research is not intended to support or oppose any political candidate or office. The research has no affiliation with any political candidate or campaign and has received no financial support from any political candidate or campaign. Should you have any questions about this study, please contact Prof. Brendan Nyhan at [brendan.j.nyhan@dartmouth.edu](mailto:brendan.j.nyhan@dartmouth.edu).

## Online Appendix B

In addition to examining the effects of warnings on perceived accuracy of news headlines, we also investigated the effect of a warning about misleading articles or the presence of a “Disputed” or “Rated false” tag on liking or sharing behavior (RQ1). Prior research has found that the perceived truthfulness of information, based on collective opinion data including number of other shares, impacts the likelihood that it is shared (Li and Sakamoto 2014). Tagging articles as disputed or false explicitly does this, yet this explicit tag has not yet been tested. Moreover, previous research has found that social endorsements predict individuals’ likelihood to like and share articles on Facebook (Messing and Westwood 2014).

Our results show that participants reported being somewhat less likely to “like” false headlines that were tagged as “Rated false” but not to share them. We found that the presence of a “Rated false” tag causes a marginally significant reduction in the probability that a respondent would “like” a story on Facebook ( $-0.06$ ,  $p < .10$ ) but has no measurable effect on the self-reported likelihood of sharing it ( $-0.04$ , *ns*). Moreover, we cannot conclude that exposure to a warning or a “Disputed” tag reduces self-reported willingness to “like” or share false stories. We note that our results for this research question may be the result of participants’ general unwillingness to “like” or share the articles they saw. For example, 76% of responses indicated that participants were “not at all likely” to “like” a given article and 80% said they were “not at all likely” to share it. It is also worth noting that our decision to exclude non-users of Facebook from this analysis created a uniformly-distributed sampling bias that may have impacted the results. Finally, we note that regression models that exclude responses to untagged false headlines (Table C3), that include random effects (Table C4), and that employ ordered probit (Table C1) are included in Online Appendix C. Across all of these models, the fact-checking tags did not appear to have strong or consistent effects on respondents’ self-reported willingness to “like” or share false news headlines.

Table B1: Experimental effects on social endorsements of false headlines

	“Like”	Share
General warning	0.01 (0.04)	0.01 (0.04)
“Disputed” tag	-0.03 (0.04)	-0.03 (0.04)
“Disputed” × warning	-0.04 (0.05)	-0.05 (0.05)
“Rated false” tag	-0.06* (0.04)	-0.04 (0.04)
“Rated false” × warning	-0.04 (0.05)	-0.07 (0.05)
Constant	1.27*** (0.03)	1.29*** (0.03)
Question fixed effects	Yes	Yes
N (responses)	11127	11119
Respondents	2377	2377

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. “Like” measures the likelihood that Facebook users would “like” a headline on a four-point scale from “Not at all likely” (1) to “Very likely” (4). “Likelihood to share” measures the likelihood that Facebook users would share a headline on a four-point scale from “Not at all likely” (1) to “Very likely” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

## Online Appendix C

Table C1: Comparison of survey sample with population benchmarks

Characteristic	Sample	Census	Gallup	FEC
<i>Education</i>				
Less than high school	0.3%	13.7%	-	-
High school or equivalent	8.4%	26.0%	-	-
Some college	24.8%	16.8%	-	-
Associate's degree	11.7%	10.1%	-	-
Bachelor's degree	38.8%	20.8%	-	-
Master's degree	12.4%	9.3%	-	-
Professional or doctorate degree	3.6%	3.4%	-	-
<i>Age</i>				
18–24	13.9%	13.1%	-	-
25–44	64.8%	35.0%	-	-
45–64	18.7%	34.7%	-	-
65 and older	2.6%	17.2%	-	-
<i>Gender</i>				
Male	45.3%	49.2%	-	-
Female	54.3%	50.8%	-	-
<i>Trump approval</i>				
Disapprove	70.0%	-	53.0%	-
Approve	30.0%	-	40.0%	-
<i>2016 presidential vote</i>				
Clinton	56.3%	-	-	48.2%
Trump	30.1%	-	-	46.1%
Other	13.6%	-	-	5.7%

Sources for population benchmarks: education (United States Census Bureau 2017), age and gender (Howden and Meyer 2011), Trump approval (Gallup News 2017), 2016 election results (Federal Election Commission 2017).

Table C2: Distribution of accuracy beliefs

Headline	Not at all accurate	Not very accurate	Somewhat accurate	Very accurate
Draft reinstated (F)	52.1%	28.0%	15.8%	4.1%
Bee movie (F)	54.6%	28.7%	12.1%	4.5%
Chaffetz blackmail (F)	42.6%	38.1%	17.4%	2.0%
Protesters paid (F)	35.9%	30.6%	24.6%	9.0%
Marines in Trump plane (F)	41.8%	29.6%	21.0%	7.6%
FBI agent dead (F)	50.7%	27.2%	16.3%	5.8%
Trump/Civil war (T)	14.6%	19.9%	29.3%	36.2%
Syria airstrikes (T)	2.3%	5.2%	18.4%	74.1%
Gorsuch confirmed (T)	3.9%	9.1%	23.4%	63.7%

See Online Appendix A for question wording. “T” indicates true and “F” indicates false news headlines.

Table C3: Excludes responses to untagged false headlines in “Disputed”/“Rated false” conditions

	Accuracy	“Like”	Share
General warning	-0.08** (0.03)	0.01 (0.04)	0.01 (0.04)
“Disputed” tag	-0.24*** (0.04)	-0.03 (0.04)	-0.03 (0.04)
“Disputed” × warning	0.04 (0.05)	-0.04 (0.05)	-0.05 (0.05)
“Rated false” tag	-0.34*** (0.04)	-0.06* (0.04)	-0.04 (0.04)
“Rated false” × warning	0.03 (0.05)	-0.04 (0.05)	-0.07 (0.05)
Constant	1.85*** (0.03)	1.27*** (0.03)	1.29*** (0.03)
Question fixed effects	Yes	Yes	Yes
N (responses)	11962	11127	11119
Respondents	2554	2377	2377

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). “Like” measures the likelihood that Facebook users would “like” a headline on a four-point scale from “Not at all likely” (1) to “Very likely” (4). “Likelihood to share” measures the likelihood that Facebook users would share a headline on a four-point scale from “Not at all likely” (1) to “Very likely” (4). Respondents in the pure control condition are excluded.

Table C4: Accuracy and social endorsement effects models (random effects)

	Accuracy	“Like”	Share
General warning	-0.08** (0.03)	0.01 (0.04)	0.01 (0.04)
“Disputed” tag	-0.24*** (0.04)	-0.03 (0.04)	-0.03 (0.04)
“Disputed” × warning	0.04 (0.05)	-0.04 (0.05)	-0.05 (0.05)
“Rated false” tag	-0.34*** (0.04)	-0.06* (0.04)	-0.04 (0.04)
“Rated false” × warning	0.04 (0.05)	-0.04 (0.05)	-0.07 (0.05)
Constant	1.86*** (0.03)	1.28*** (0.03)	1.30*** (0.03)
Question fixed effects	Yes	Yes	Yes
Respondent random effects	Yes	Yes	Yes
N (responses)	11962	11127	11119
Respondents	2554	2377	2377

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent and respondent random effects. “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). “Like” measures the likelihood that Facebook users would “like” a headline on a four-point scale from “Not at all likely” (1) to “Very likely” (4). “Likelihood to share” measures the likelihood that Facebook users would share a headline on a four-point scale from “Not at all likely” (1) to “Very likely” (4). Respondents in the pure control condition and responses with an untagged false headline are excluded.



Table C5: Accuracy effects models (includes pure controls)

	Accuracy
No correction	0.05 (0.03)
General warning (no correction)	-0.08** (0.03)
“Disputed” tag	-0.19*** (0.04)
“Disputed” × warning	0.04 (0.05)
“Rated false” tag	-0.30*** (0.04)
“Rated false” × warning	0.03 (0.05)
Constant	1.80*** (0.03)
Question fixed effects	Yes
N (responses)	14550
Respondents	2986

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Social endorsement effects are not included because they could not be measured in respondents who had not seen a headline (i.e. pure controls). Responses with an untagged false headline are excluded. (Note: The effect of the general warning in the pure control condition was misspecified in the pre-registration. It is calculated as the coefficient for the general warning term, not the difference between the general warning and no correction coefficients.)

Table C6: Perceived accuracy effects by article slant (includes pure controls)

	Anti-Trump	Pro-Trump
No correction	0.05 (0.05)	0.00 (0.05)
General warning	-0.11** (0.05)	-0.06 (0.05)
“Disputed” tag	-0.23*** (0.05)	-0.24*** (0.05)
“Rated false” tag	-0.37*** (0.05)	-0.37*** (0.05)
Trump approval	-0.24*** (0.06)	0.56*** (0.08)
No correction × Trump approval	0.06 (0.09)	0.07 (0.10)
Warning × Trump approval	-0.04 (0.09)	0.06 (0.10)
“Disputed” × warning	0.10 (0.07)	0.04 (0.07)
“Disputed” × Trump approval	0.20** (0.09)	0.10 (0.11)
“Disputed” × warning × Trump approval	-0.05 (0.13)	-0.12 (0.15)
“Rated false” × warning	0.12* (0.07)	0.01 (0.07)
“Rated false” × Trump approval	0.28*** (0.09)	0.06 (0.11)
“Rated false” × warning × Trump approval	-0.06 (0.13)	-0.07 (0.15)
Constant	1.86*** (0.04)	1.99*** (0.04)
Question fixed effects	Yes	Yes
N (responses)	7266	7266
Respondents	2982	2980

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

Table C7: Ordered probit model of accuracy and social endorsement effects

	Accuracy	“Like”	Share
General warning	-0.09** (0.04)	0.02 (0.06)	0.04 (0.07)
“Disputed” tag	-0.29*** (0.05)	-0.06 (0.06)	-0.03 (0.07)
“Rated false” tag	-0.45*** (0.05)	-0.10 (0.07)	-0.05 (0.07)
“Disputed” × warning	0.04 (0.07)	-0.08 (0.09)	-0.12 (0.10)
“Rated false” × warning	0.03 (0.07)	-0.09 (0.10)	-0.19* (0.11)
Question fixed effects	Yes	Yes	Yes
N (responses)	11962	11127	11119
Respondents	2554	2377	2377

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). Ordered probit model with robust standard errors clustered by respondent (cutpoints omitted). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

Table C8: Ordered probit model of effects on perceived accuracy of false headlines by article slant

	Anti-Trump	Pro-Trump
General warning	-0.13** (0.06)	-0.08 (0.06)
“Disputed” tag	-0.37*** (0.07)	-0.31*** (0.07)
“Rated false” tag	-0.57*** (0.07)	-0.53*** (0.07)
Trump approval	-0.23*** (0.08)	0.69*** (0.07)
Warning × Trump approval	-0.08 (0.12)	0.09 (0.11)
“Disputed” × warning	0.11 (0.10)	0.05 (0.10)
“Disputed” × Trump approval	0.19 (0.13)	0.09 (0.12)
“Disputed” × warning × Trump approval	-0.07 (0.19)	-0.15 (0.17)
“Rated false” × warning	0.14 (0.11)	0.01 (0.11)
“Rated false” × Trump approval	0.28** (0.14)	0.14 (0.12)
“Rated false” × warning × Trump approval	-0.06 (0.20)	-0.10 (0.18)
Question fixed effects	Yes	Yes
N (responses)	5972	5972
Respondents	2550	2548

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). Ordered probit model with robust standard errors clustered by respondent (cutpoints omitted). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

Table C9: Ordered probit model of spillover effects of warnings on perceived accuracy

	Untagged false headlines	Real headlines
General warning	-0.10** (0.05)	-0.15** (0.07)
“Disputed” condition	0.05 (0.07)	0.12 (0.07)
“Rated false” condition	-0.02 (0.06)	0.04 (0.08)
“Disputed” × warning	0.05 (0.09)	0.09 (0.11)
“Rated false” × warning	0.13 (0.09)	0.15 (0.11)
Question fixed effects	Yes	Yes
N (responses)	7953	6575
Respondents	2432	2498

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). Ordered probit models with robust standard errors clustered by respondent (cutpoints omitted). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a “Disputed” or “Rated false” tag for the specific headline in question.

Table C10: Experimental effects on perceived accuracy (Facebook users)

	Accuracy
General warning	-0.08** (0.04)
“Disputed” tag	-0.24*** (0.04)
“Disputed” × warning	0.04 (0.05)
“Rated false” tag	-0.34*** (0.04)
“Rated false” × warning	0.03 (0.05)
Constant	1.86*** (0.03)
Question fixed effects	Yes
N (responses)	11118
Respondents	2376

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent (excludes non-Facebook users). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

Table C11: Experimental effects on perceived accuracy by article slant (Facebook users)

	Anti-Trump	Pro-Trump
General warning	-0.10* (0.05)	-0.08 (0.05)
“Disputed” tag	-0.27*** (0.05)	-0.25*** (0.05)
“Rated false” tag	-0.41*** (0.05)	-0.38*** (0.05)
Trump approval	-0.19*** (0.07)	0.64*** (0.07)
Warning × Trump approval	-0.05 (0.09)	0.07 (0.10)
“Disputed” × warning	0.08 (0.08)	0.07 (0.07)
“Disputed” × Trump approval	0.15 (0.10)	0.04 (0.11)
“Disputed” × warning × Trump approval	-0.03 (0.14)	-0.14 (0.15)
“Rated false” × warning	0.11 (0.08)	0.02 (0.07)
“Rated false” × Trump approval	0.22** (0.10)	0.00 (0.10)
“Rated false” × warning × Trump approval	-0.04 (0.14)	-0.08 (0.15)
Constant	1.92*** (0.04)	2.00*** (0.04)
Question fixed effects	Yes	Yes
N (responses)	5550	5550
Respondents	2372	2370

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent (excludes non-Facebook users). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

Table C12: Experimental tests for spillover effects of warnings (Facebook users)

	Untagged false headlines	Real headlines
General warning	-0.08** (0.04)	-0.15*** (0.04)
“Disputed” condition	0.02 (0.05)	0.05 (0.04)
“Rated false” condition	0.05 (0.05)	0.01 (0.04)
“Disputed” × warning	0.07 (0.06)	0.10* (0.06)
“Rated false” × warning	0.02 (0.07)	0.18*** (0.06)
Constant	1.88*** (0.03)	2.87*** (0.03)
Question fixed effects	Yes	Yes
N (responses)	7409	6134
Respondents	2275	2326

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent (excludes non-Facebook users). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a “Disputed” or “Rated false” tag for the specific headline in question.



Table C13: Experimental effects on perceived accuracy by pre-treatment false news belief tercile

	Accuracy
General warning	-0.09* (0.05)
“Disputed” tag	-0.23*** (0.05)
“Disputed” × warning	0.03 (0.07)
“Rated false” tag	-0.33*** (0.05)
“Rated false” × warning	0.03 (0.07)
Medium false news belief tercile	0.26*** (0.05)
High false news belief tercile	0.59*** (0.06)
“Disputed” × medium tercile	0.02 (0.07)
“Disputed” × high tercile	-0.10 (0.09)
Warning × medium tercile	0.03 (0.07)
Warning × high tercile	-0.07 (0.08)
“Disputed” × warning × medium tercile	-0.03 (0.11)
“Disputed” × warning × high tercile	0.15 (0.13)
“Rated false” × medium tercile	-0.03 (0.08)
“Rated false” × high tercile	-0.04 (0.09)
“Rated false” × warning × medium tercile	0.04 (0.11)
“Rated false” × warning × high tercile	0.02 (0.13)
Constant	1.63*** (0.04)
Question fixed effects	Yes
N (responses)	11118
Respondents	2376

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. False news belief terciles are calculated using a pre-treatment measure of average level of belief in four false news claims (the low tercile is the excluded category above; see Online Appendix A for item wording). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

Table C14: Experimental effects on perceived accuracy by pre-treatment false news belief tercile

	Anti-Trump		Pro-Trump	
	Disapprove	Approve	Disapprove	Approve
General warning	-0.03 (0.06)	-0.22** (0.10)	-0.13** (0.06)	0.03 (0.14)
“Disputed” tag	-0.22*** (0.07)	-0.10 (0.11)	-0.26*** (0.07)	-0.28** (0.14)
“Disputed” × warning	-0.00 (0.09)	0.09 (0.15)	0.08 (0.09)	-0.05 (0.22)
“Rated false” tag	-0.30*** (0.06)	-0.16 (0.13)	-0.37*** (0.06)	-0.49*** (0.16)
“Rated false” × warning	0.04 (0.09)	0.06 (0.17)	0.03 (0.08)	-0.05 (0.24)
Medium false news belief tercile	0.42*** (0.07)	0.21* (0.11)	0.15** (0.07)	0.14 (0.13)
High false news belief tercile	0.70*** (0.09)	0.56*** (0.14)	0.51*** (0.10)	0.36*** (0.14)
“Disputed” × medium tercile	-0.03 (0.11)	-0.05 (0.17)	0.16 (0.10)	-0.14 (0.21)
“Disputed” × high tercile	-0.34** (0.14)	-0.10 (0.18)	-0.23 (0.15)	0.31 (0.20)
Warning × medium tercile	-0.13 (0.10)	0.12 (0.15)	0.18* (0.10)	-0.06 (0.20)
Warning × high tercile	-0.20* (0.12)	0.02 (0.18)	0.03 (0.14)	-0.08 (0.20)
“Disputed” × warning × medium tercile	0.01 (0.16)	0.03 (0.23)	-0.21 (0.15)	0.33 (0.32)
“Disputed” × warning × high tercile	0.58*** (0.19)	-0.07 (0.25)	0.25 (0.20)	-0.30 (0.30)
“Rated false” × medium tercile	-0.15 (0.11)	-0.11 (0.18)	0.12 (0.11)	-0.06 (0.20)
“Rated false” × high tercile	-0.29** (0.14)	-0.06 (0.20)	-0.12 (0.13)	0.24 (0.22)
“Rated false” × warning × medium tercile	0.09 (0.15)	0.15 (0.24)	-0.11 (0.15)	0.28 (0.32)
“Rated false” × warning × high tercile	0.27 (0.21)	-0.06 (0.26)	0.09 (0.20)	-0.21 (0.32)
Constant	1.61*** (0.05)	1.55*** (0.09)	1.81*** (0.05)	2.57*** (0.10)
Question fixed effects	Yes	Yes	Yes	Yes
N (responses)	4165	1807	4162	1810
Respondents	1777	773	1775	773

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent (excludes non-Facebook users). False news belief terciles are calculated using a pre-treatment measure of average level of belief in four false news claims (the low tercile is the excluded category above; see Online Appendix A for item wording). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a false headline that did not include a “Disputed” or “Rated false” tag.

Table C15: Experimental tests for warning spillover effects by pre-treatment false news belief tercile

	Untagged false headlines	Real headlines
General warning	-0.09* (0.05)	-0.07 (0.06)
“Disputed” condition	0.05 (0.07)	0.12** (0.06)
“Disputed” × warning	0.03 (0.09)	0.05 (0.08)
“Rated false” condition	0.01 (0.06)	0.09 (0.06)
“Rated false” × warning	0.11 (0.09)	0.05 (0.09)
Medium false news belief tercile	0.26*** (0.05)	-0.11* (0.06)
High false news belief tercile	0.59*** (0.06)	-0.43*** (0.07)
“Disputed” × medium tercile	0.01 (0.10)	-0.17** (0.09)
“Disputed” × high tercile	-0.16 (0.12)	0.05 (0.11)
Warning × medium tercile	0.03 (0.07)	-0.15 (0.10)
Warning × high tercile	-0.07 (0.08)	0.06 (0.10)
“Disputed” × warning × medium tercile	0.06 (0.14)	0.14 (0.13)
“Disputed” × warning × high tercile	0.13 (0.15)	-0.13 (0.15)
“Rated false” × medium tercile	0.09 (0.09)	-0.16* (0.09)
“Rated false” × high tercile	-0.04 (0.11)	-0.01 (0.10)
“Rated false” × warning × medium tercile	-0.16 (0.13)	0.24* (0.14)
“Rated false” × warning × high tercile	-0.06 (0.15)	0.06 (0.15)
Constant	1.64*** (0.04)	3.01*** (0.04)
Question fixed effects	Yes	Yes
N (responses)	7968	6585
Respondents	2436	2502

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent (excludes non-Facebook users). False news belief terciles are calculated using a pre-treatment measure of average level of belief in four false news claims (the low tercile is the excluded category above; see Online Appendix A for item wording). “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition are excluded, as are responses from participants in the disputed and false conditions who saw a “Disputed” or “Rated false” tag for the specific headline in question.

Table C16: Experimental effects on perceived accuracy by political knowledge

	Trump disapprovers		Trump approvers	
	Belief-consistent	Belief-inconsistent	Belief-consistent	Belief-inconsistent
“Disputed” tag	-0.26*** (0.07)	-0.23*** (0.07)	-0.25** (0.12)	-0.18* (0.10)
“Rated false” tag	-0.51*** (0.07)	-0.41*** (0.07)	-0.43*** (0.12)	-0.25** (0.11)
High knowledge	-0.34*** (0.07)	-0.19*** (0.07)	-0.05 (0.12)	-0.45*** (0.10)
“Disputed” × high knowledge	-0.05 (0.10)	-0.02 (0.10)	0.12 (0.18)	0.10 (0.15)
“Rated false” × high knowledge	0.24** (0.10)	0.10 (0.10)	0.12 (0.18)	0.18 (0.15)
Constant	2.02*** (0.05)	2.05*** (0.05)	2.74*** (0.08)	1.98*** (0.08)
Question fixed effects	Yes	Yes	Yes	Yes
N (responses)	2141	2143	927	924

\*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < .01$  (two-sided). OLS models with robust standard errors clustered by respondent. “Accuracy” measures belief that a headline is accurate from “Not at all accurate” (1) to “Very accurate” (4). Respondents in the pure control condition and responses with an untagged false headline are excluded. Belief-consistent/inconsistent articles are those which match the assumed partisanship of a participant based on his or her approval of President Trump. If a participant approved of the President, then belief-consistent articles would be pro-Trump and belief-inconsistent articles would be anti-Trump.