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# Political Endorsements by Scientific Organizations Reduce Trust in Scientific Expertise During COVID-19

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## Abstract

High-profile political endorsements by scientific organizations and publications have been common in recent years, which raises concerns about political backlash against the endorsing organization in particular and scientific expertise in general. In a pre-registered large-sample online controlled experiment, I randomly assign subjects to receive information about *Nature*'s endorsement of Joe Biden for the 2020 U.S. presidential election in the context of the COVID-19 pandemic. Seeing the endorsement message leads to large reductions in self-reported trust in *Nature* among Trump supporters. This distrust translates into lower demand for COVID-related information provided by *Nature*, as evidenced by substantially reduced requests for *Nature* articles on vaccine efficacy and emerging COVID variants when offered. The endorsement also reduces Trump supporters' stated trust in scientists in general, suggesting "reputational externalities" on the scientific community. Treatment effect estimates for Biden supporters are positive but small and often statistically insignificant across outcomes. Finally, I find little evidence that the endorsement is successful at changing subjects' views about Biden and Trump. These results show political endorsement by scientific organizations can undermine public confidence in the endorsing organization and in the scientific community.

**Keywords:** Polarization, trust in science, endorsement, COVID-19

# 1 Introduction

2 Scientific organizations and publications have become increasingly involved in  
3 electoral politics. In the run-up to the 2020 U.S. Presidential Election, numer-  
4 ous influential scientific publications, including *Nature*, *Scientific American*,  
5 the *Lancet*, the *New England Journal of Medicine*, and *Science*, published  
6 editorial pieces criticising then-President Donald Trump’s mishandling of the  
7 COVID-19 pandemic and his antagonistic attitudes toward science. Most of  
8 these journals urged voters to replace Trump. Among them *Nature*, *Scientific*  
9 *American*, and the *Lancet* explicitly endorsed his challenger Joe Biden[1].  
10 It marked the first time *Scientific American* or the *Lancet* made political  
11 endorsement. These publications were also joined by 81 American Nobel  
12 laureates in endorsing Biden’s candidacy[2].

13 The increased political engagement by scientists raises concerns that their  
14 endorsements cause right-wing backlash[3]. Trust in the scientific community  
15 has been on the decline in the U.S. for decades, with the most pronounced  
16 decline among those on the political right[4][5]. During the COVID-19 pan-  
17 demic, such skepticism toward scientific expertise reduced compliance with  
18 public health intervention[6] and may explain the partisan difference in  
19 compliance[7][8], with important implications for public health outcomes[9].  
20 By endorsing a Democratic candidate in a polarizing presidential election  
21 during the pandemic, scientists risk intensifying existing distrust from a large  
22 segment of the population, particularly because these endorsements were  
23 widely reported by conservative media outlets[10][11].

24 The possibility of a right-wing backlash is consistent with the literature  
25 on affective political polarization in the U.S.[12], which often finds that  
26 associating individuals or entities with a political party increases out-party  
27 animosity toward them, and Bayesian models of information economics and  
28 decision theory, which predicts that an agent uncertain about the quality  
29 of an information source may judge its quality by the degree to which its  
30 messages conform to her prior[13][14][15]. On the other hand, it would not be  
31 unreasonable to expect no backlash. Given the prestige of publications such  
32 *Nature*, it is unclear if their credibility would be judged on the basis of a  
33 political endorsement - in Bayesian terms, there may or may not be enough  
34 prior uncertainty with respect to source quality to trigger a substantively  
35 meaningful update. In addition, expressed partisan hostilities are sometimes  
36 “cheerleading”[16] that does not translate into behaviours when there are  
37 stakes such as health risks. Finally, research show that priming Americans  
38 about COVID reduces affective polarization[17]. Whether these endorsements  
39 have any effects on trust and behaviours is thus an empirical question.

40 This paper presents findings from a pre-registered online experiment exam-  
41 ining the effects of *Nature*’s 2020 endorsement of Joe Biden for U.S. President  
42 amid the COVID-19 pandemic. Conducted in late July and early August  
43 2021, the experiment randomly assigned subjects to receive information about  
44 *Nature*’s endorsement, while the control group received irrelevant information.

45 The results show that the endorsement reduced stated trust in *Nature*’s

46 competence and impartiality among Trump supporters by 0.633 to 0.845 (full-  
47 sample) standard deviations. Treated Trump and Biden supporters become  
48 two to four times more polarized than control subjects on these stated mea-  
49 sures of trust in *Nature*. The effect is greater for subjects who did not expect  
50 *Nature* to make an endorsement *ex ante*. The endorsement also significantly  
51 dampened Trump supporters' demand for COVID-related information pro-  
52 vided by *Nature*. When prompted to acquire information about emerging  
53 COVID-19 variants and vaccine efficacy, treated Trump supporters were 38%  
54 less likely than control Trump supporters to request stories from *Nature*'s  
55 website, indicating the decrease in stated trust has behavioural consequences.

56 The endorsement also significantly reduced trust in scientists in general  
57 among Trump supporters, creating a "reputational externality" on the entire  
58 scientific community.

59 Point estimates for Biden supporters suggest the endorsement may have  
60 slightly increased Biden supporters' confidence in *Nature* and scientists, but  
61 the estimated effects are substantively small and often statistically insignifi-  
62 cant.

63 Finally, there is little evidence that the endorsement changed subjects'  
64 opinions about the two presidential candidates with respect to the issues that  
65 the endorsement piece highlights.

66 In addition to pointing to adverse consequences of well-intentioned  
67 political activism in the scientific community, this paper contributes to the  
68 literature on scientific communication, the social sciences of trust in scientific  
69 expertise[18][19][4][20][21], on political endorsement[22][23][24][25], non-  
70 political consequences of political polarization in the U.S.[7][26][27][28][29][30],  
71 and the social and behavioral sciences of COVID-19 response[31]. To the  
72 best of my knowledge, this paper presents the first experimental study of the  
73 effects scientists' political activities have on trust in scientists. (For an obser-  
74 vational study on public opinion effects of the 2017 March for Science rallies,  
75 see [20]). In addition, though political endorsements have been extensively  
76 studied, no previous work has examined the effect of endorsements on the  
77 endorser or public perceptions thereof. This study fills this gap.

78

## 79 **2 Results**

### 80 **2.1 The experiment**

81 The experiment took the form of an online survey with randomized compo-  
82 nents. At the beginning of the survey, subjects were screened for attention  
83 and asked about their political beliefs. The experimental sample consists  
84 of 4,260 individuals and is representative of the U.S. adult population  
85 along most demographic dimensions (Supplementary Table 1). However, the  
86 sample is skewed toward Biden supporters, as indicated by responses to a  
87 (pre-treatment) question asking subjects their preference between Biden and

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88 Trump. 55.14% of the subjects preferred Biden, while 35.06% favoured Trump.

89

#### Scientific journal *Nature* endorsed Joe Biden

- Weeks before the 2020 presidential election, *Nature*'s editorial board officially endorsed Joe Biden, citing:
  - Donald Trump's pandemic response had been "disastrous";
  - Biden, unlike Trump, would listen to science, and thus;
  - Biden would handle the COVID-19 pandemic better.



(Reference: The full endorsement can be found here: <https://www.nature.com/articles/d41586-020-02852-x>)

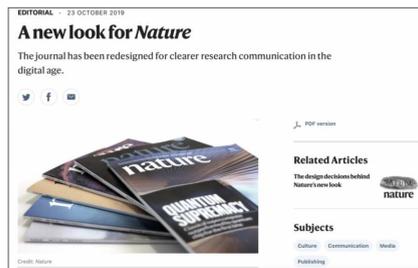
- *Nature* is one of the most-cited and most prestigious peer-reviewed scientific journals in the world.

# nature

Fig. 1 Treatment message

#### Scientific journal *Nature* unveiled its new design

- In October 2019, *Nature*'s editorial board unveiled a new design for the scientific journal, including:
  - New print and digital designs:
  - A new logo.
- The editors hope the redesign will help the journal to communicate research with style and greater clarity.



(Reference: The full announcement can be found here: <https://www.nature.com/articles/d41586-019-03167-2>)

- *Nature* is one of the most-cited and most prestigious peer-reviewed scientific journals in the world.

# nature

Fig. 2 Control message

90 To examine the effect of *Nature*'s endorsement, I randomly assigned half of  
91 the subjects to read a short message summarizing *Nature*'s endorsement piece  
92 for Biden (Figure 1). The message highlights *Nature*'s criticism of Trump's  
93 mishandling of COVID-19 and the expectation that Biden would do better. To  
94 make sure the message is credible, the summary is followed by a screenshot of  
95 the endorsement piece's title, lead paragraph, and cover picture from *Nature*'s  
96 official website, as well as a link to the piece. Finally, subjects are reminded  
97 that *Nature* is "one of the most cited and most prestigious peer-reviewed sci-  
98 entific journals in the world".

99 Control subjects are assigned to read a message about *Nature*'s announce-  
100 ment of its new visual designs for its website and print copies, instead of the  
101 endorsement (Figure 2). The message is presented in the same format as the  
102 endorsement message that treatment subjects see. The text is also followed by  
103 a screenshot and a link to *Nature*'s announcement of the new design, as well  
104 as the same reminder of *Nature*'s scientific prominence. Nowhere in the survey  
105 are the control subjects informed of the endorsement.

106 After the treatment or control message, subjects were shown batteries of  
107 questions and messages eliciting measures of (1) trust in *Nature*, (2) assess-  
108 ments of Biden and Trump, (3) demand for COVID-19 information provided  
109 by *Nature*, (4) persuasiveness of a climate change-related message attributed  
110 to *Nature*, and (5) trust in scientists in general.

111 Analyses are conducted on subjects who indicate some degrees of supports  
112 for either Biden or Trump in the aforementioned pre-treatment candidate pref-  
113 erence question. This excludes a small minority (8.80%) of subjects who stated  
114 that they support "someone else".

## 115 2.2 Stated trust in *Nature*

116 First examined is the effect of seeing the endorsement on stated trust in  
117 *Nature*. The survey elicits the subject's confidence in the journal along two  
118 dimensions, namely its competence and impartiality. These outcomes are cap-  
119 tured by two questions asking subjects to report how much they trust *Nature*  
120 for (1) being "*informed*" when it comes to providing advice on science-related  
121 issues facing the society, and (2) giving their "*unbiased*" opinions to the  
122 public, to the best of their knowledge, when contentious issues are concerned.  
123 The questions and the distributions of responses by political alignment and  
124 treatment status are presented in Figures 3 and 4.

125 Table 1 reports the results from regression analyses of the treatment  
126 effects. The responses to the two questions are mapped into five-point  
127 scales and, as are all other outcomes analyzed in Table 1, standardized  
128 as  $z$ -scores with mean 0 and standard deviation 1. The endorsement  
129 has large negative effects on Trump supporters' trust in *Nature*'s com-  
130 petence ( $p < 0.001, 95\% CI = [-0.955, -0.752]$ ) and impartiality  
131 ( $p < 0.001, 95\% CI = [-0.730, -0.534]$ ). The effects on Biden support-  
132 ers are positive but small, significant only for the "*informed*" outcome  
133 ( $p < 0.001, 95\% CI = [0.047, 0.169]$ ). Taken together, the endorsement

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134 appears to have further polarized trust in *Nature*.

135

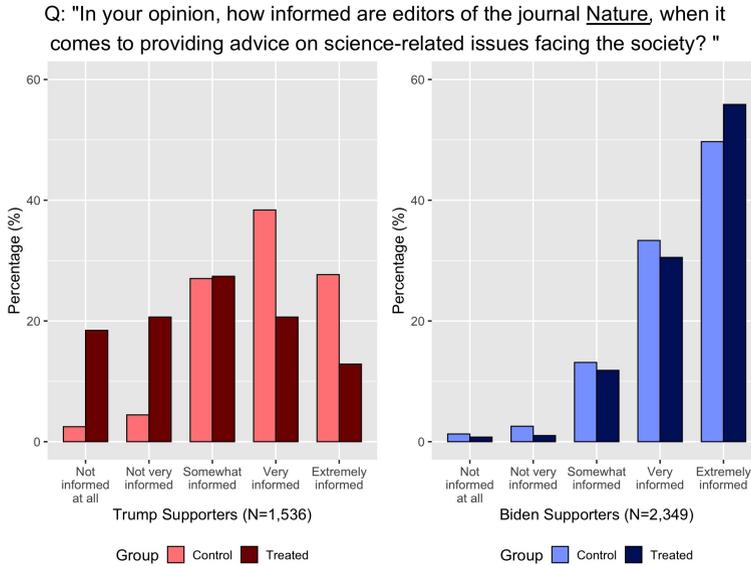


Fig. 3 Stated trust in *Nature* being “informed”

Q: "When contentious or divisive issues are concerned, how much confidence do you have in the editorial board of *Nature* to provide their unbiased opinions to the public, to the best of their knowledge? "

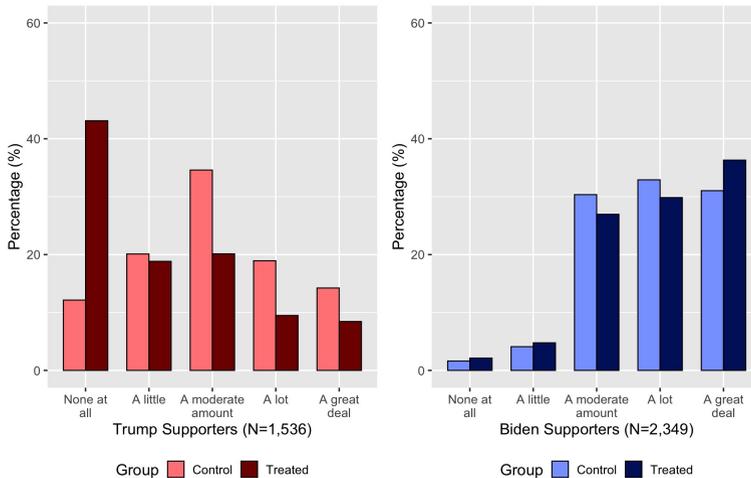


Fig. 4 Stated trust in *Nature* being “unbiased”

**Table 1** Experiment results

Outcome $z$ -score	$\widehat{CATE}$		
	Trump supporters (Robust s.e.)	Biden supporters (Robust s.e.)	Baseline difference (Robust s.e.)
<i>Nature</i> is informed when providing advice	-0.854*** (0.0518)	0.108*** (0.0309)	-0.387*** (0.0386)
<i>Nature</i> is unbiased on contentious issues	-0.633*** (0.0498)	0.0450 (0.0313)	-0.655*** (0.0399)
Biden would have handled COVID better than Trump had he been president in 2020	0.0556+ (0.0322)	0.0448+ (0.0238)	-1.642*** (0.0282)
Trump would have handled COVID better than Biden if he was still president now(2021)	-0.0241 (0.0329)	-0.0168 (0.0240)	1.628*** (0.0287)
Biden is better at making use of scientific knowledge for decision-making than Trump	0.0449 (0.0354)	0.0135 (0.0140)	-1.775*** (0.0267)
Subject requests <i>Nature</i> 's article for info about vaccine efficacy against new variants	-0.285*** (0.0463)	0.048 (0.0409)	-0.386*** (0.0455)
U.S. scientists are informed when providing advice	-0.130* (0.0532)	0.0485 (0.0335)	-0.756*** (0.0442)
U.S. scientists are unbiased on contentious issues	-0.161** (0.0516)	0.0161 (0.0310)	-0.937*** (0.0422)
" $\geq 90\%$ " of climate scientists agree human-caused climate change is real	-0.0461 (0.0558)	0.0265 (0.0350)	-0.590*** (0.0463)
Human-caused climate change is real	-0.0232 (0.0582)	0.0147 (0.0266)	-0.974*** (0.0456)

$N = 3,885$  " $\widehat{CATE}$  Trump (Biden) supporters" is the estimated treatment effect for Trump (Biden) supporters. "Baseline difference" is the mean difference between Trump supporters and Biden supporters within the control group. Sample includes 1,173 control Biden supporters, 1,176 treatment Biden supporters, 766 control Trump supporters, and 770 treatment supporters. Outcomes are  $z$ -scores with mean 0 and standard deviation 1. All estimates are from OLS estimation of a linear regression model, described in the Method section. Robust standard errors in parentheses. All null hypotheses are testing are two-sided  $t$  tests. +  $p < 0.10$ , \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*  $p < 0.001$

136 The estimates suggest that the polarizing effect of the endorsement is  
 137 greater than the baseline difference between Biden supporters and Trump  
 138 supporters. Untreated, Trump supporters' confidence in *Nature*'s competence  
 139 and impartiality is 0.387 ( $p < 0.001$ , 95%  $CI = [0.577, 0.734]$ ) and 0.655  
 140 ( $p < 0.001$ , 95%  $CI = [0.311, 0.463]$ ) standard deviations lower than Biden  
 141 supporters', respectively. The treatment pulls them apart by an additional  
 142 0.962 standard deviations ( $p < 0.001$ , 95%  $CI = [0.844, 1.080]$ ) and 0.678 stan-  
 143 dard deviations ( $p < 0.001$ , 95%  $CI = [0.562, 0.793]$ ), respectively. Depending

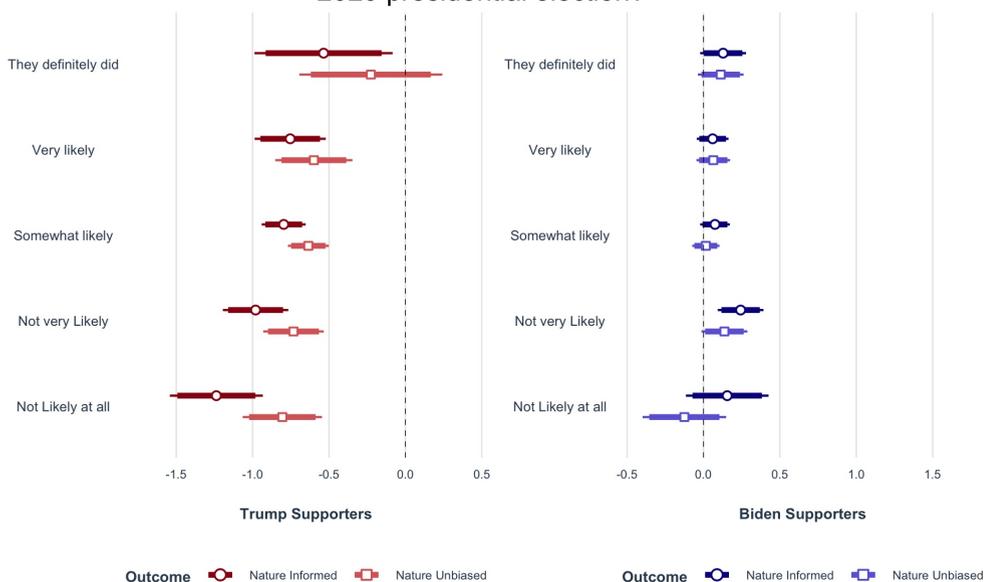
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144 on the measure looked at, the treatment increases the trust gap between the  
 145 two categories by a factor of 2 or 3.5.

146  
 147 Theoretically, the effect of the message should be greater for individu-  
 148 als who did not expect *Nature* to make political endorsements *ex ante*, as  
 149 the treatment induces in them larger updates of beliefs. I elicited subjects'  
 150 prior beliefs by asking how likely they thought it was that *Nature* made any  
 151 political endorsement in the 2020 Presidential Election, before showing the  
 152 treatment or control message. Figure 5 presents estimated effects on trust in  
 153 *Nature* for each prior belief level.

154 For Trump supporters, not expecting the endorsement is clearly associated  
 155 with larger decrease in trust when told that *Nature* did endorse Biden. The  
 156 linear interactions between prior belief and treatment are significant for both  
 157 trust in competence ( $p = 0.002$ ) and in impartiality ( $p = 0.025$ ). Among Biden  
 158 supporters, the size of the estimated increase in trust in *Nature*'s knowledge  
 159 also seems larger for those who thought endorsement was unlikely, but the  
 160 interaction is statistically insignificant. There is no discernible pattern in  
 161 effects on Biden supporters' trust in *Nature*'s impartiality, which is unsurpris-  
 162 ing given that the aggregate effect on this outcome is not significant for them.  
 163

Q: "Nature is one of the most-cited scientific journals in the world. How likely  
 do you think it is that Nature officially endorsed one of the candidates in the  
 2020 presidential election? "



Treatment effect of the endorsement message for each prior level, elicited before treatment, presented, with 90% and 95% confidence intervals.

**Fig. 5** Treatment effect heterogeneity by prior belief

## 2.3 Endorsement persuasiveness

I look into whether the endorsement was successful at changing people’s minds about Biden and Trump’s relative competence by focusing on the issues that the endorsement highlights, namely pandemic response and attitudes toward science. The next three rows of Table 1 report the findings. There is little evidence that the endorsement persuades subjects. In addition to the general absence of statistical significance, the upper bounds of the 95% CIs never exceed 0.12 standard deviation and are typically much smaller.

The construction of these questions is fairly conducive to finding persuasive effects. COVID-19 and science are the focuses of the endorsement message. If the message fails to change Americans’ views on these specific issues, it is even less likely that it shifts overall opinions about the race, let alone changing vote intentions.

## 2.4 Demand for information

Do the shifts in stated trust in *Nature* translate into changes in behaviours such as demand for COVID-related information provided by *Nature*? To study this question, I look at how subjects choose to acquire information from a menu of sources. I prompted subjects with a message reminding them of emerging COVID variants and encouraging them to “stay informed about vaccine efficacy against new COVID variants” (see Supplementary Information for the survey instrument). The topic, which was salient and affected everyone’s health and safety, was chosen to increase the stake of the choice. As the situation was rapidly changing, most people were unlikely to have kept track of it and likely to want to learn more.

The message offers links to “easy-to-read” articles about “new variants and how well available vaccines perform against them” from a variety of sources. Specifically, subjects can choose to read from *Nature*, the Mayo Clinic, (unspecified) “news media websites”, any combination of the three sources, or not to read about the topic at all.

The endorsement led to a statistically significant 14.2-percentage point reduction in the frequency at which Trump supporters request *Nature* articles ( $p < 0.001$ , 95%  $CI = [-18.750, -9.688]$ ) - a 38.3% decline relative to control Trump supporters, who requested *Nature* articles 37.1% of the time. Biden supporters, on the other hand, selected *Nature* 56.4% of the time under the control condition. The estimated treatment effect for Biden supporters is positive but small and statistically insignificant. The  $z$ -score versions of these results are presented in Table 1.

It is worth noting that the experiment was conducted early during the Delta variant-induced surge of U.S. cases in summer 2021, when the variant first became salient in Americans’ consciousness. Google Trends[32] shows that U.S. search frequencies for keywords “Delta” and “variant” was at their all-time peak when the experiment was conducted (July 28 - August 10). The

207 search frequency for “vaccine” also reached a local peak. Given the thirst for  
 208 information on this topic at the time, it is all the more telling that Trump  
 209 supporters became substantially less likely to want to read about it from  
 210 *Nature* after exposure to the endorsement information.

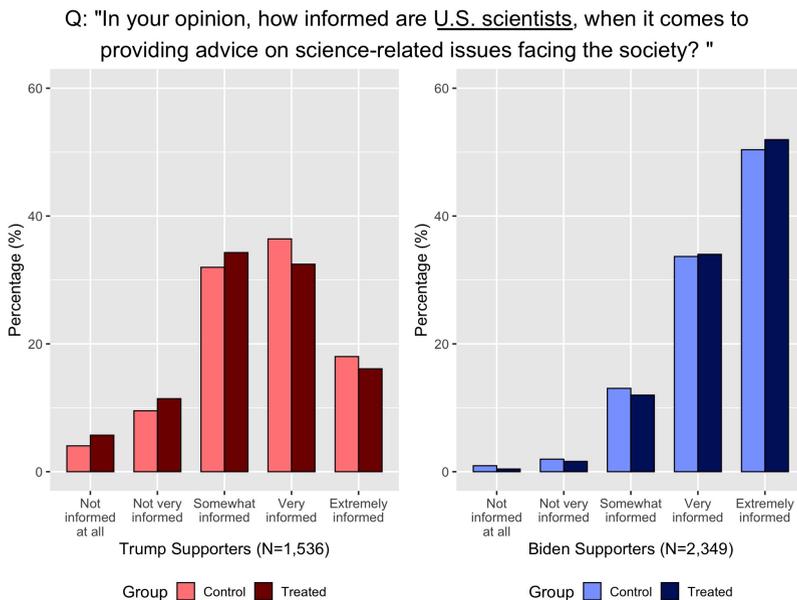
211

## 212 2.5 Trust in scientists in general

213 Next, I consider possible reputational externalities on the scientific commu-  
 214 nity. Namely, the endorsement may affect trust not just in *Nature*, but also in  
 215 scientists generally. The next set of outcomes are based on a pair of questions  
 216 that closely parallel the two questions about trust in *Nature*’s knowledge and  
 217 impartiality, with “*Nature*” replaced by “U.S. scientists”.

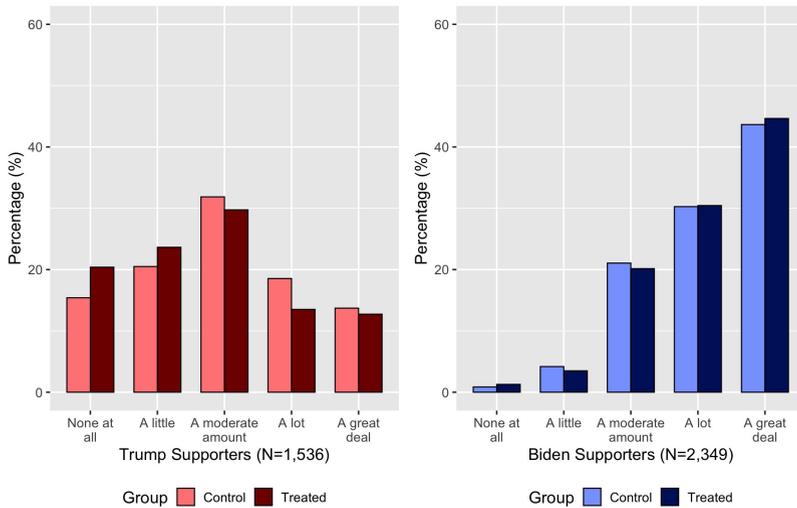
218 Figures 6 and 7 show distribution of responses by group. Under the  
 219 control condition, Trump supporters were less likely to report high lev-  
 220 els of trust toward U.S. scientists than Biden supporters, and the gap  
 221 is larger for treated subjects. Regression analyses reported in Table 1  
 222 show statistically significant negative treatment effects for Trump suppor-  
 223 ters ( $\beta^{informed} = -0.130, p = 0.015, 95\% CI = [-0.234, -0.026]$ ;  
 224  $\beta^{unbiased} = -0.161, p = 0.002, 95\% CI = [-0.262, -0.060]$ ), albeit consid-  
 225 erably smaller than the corresponding effects on trust in *Nature*. The estimates  
 226 for Biden supporters are positive but small and statistically insignificant.

227



**Fig. 6** Stated trust in **U.S. scientists** being “informed”

Q: "When contentious or divisive issues are concerned, how much confidence do you have in U.S. scientists to provide their unbiased opinions to the public, to the best of their knowledge? "



**Fig. 7** Stated trust in **U.S. scientists** being “unbiased”

## 2.6 Climate Change

I also examine whether the endorsement affects *Nature*'s credibility when it communicates scientific consensus on other domains. The questionnaire displays a quote from an editorial piece by *Nature Climate Change*, stating that 97% of climate scientists agree that climate change is real and caused by human activities. (On the questionnaire, the quote was mistakenly attributed to *Nature* instead of *Nature Climate Change*, due to the author's misunderstanding of the distinction at the time. ) The treatment group were again reminded that *Nature* endorsed Biden. Subjects were then asked if they believe the statement about climate scientists' consensus and if they believe in human-caused climate change. The signs of all estimates are consistent with results for trust in *Nature*, but the effects are small and insignificant (Table 1).

## 3 Discussion

This study shows that electoral endorsements by scientific organizations reduce public trust in the endorser, particularly among supporters of the out-party candidate. The effect has negative impacts on trust in the scientific community as a whole and information acquisition behaviours with respect to critical public health issues. Positive effects among supporters of the endorsed candidate are null or small, and they do not offset the negative effects among the opposite camp. This likely results in a lower overall level of

public confidence and more polarization. There is little evidence that seeing the endorsement message changes opinions about the candidates, even if one focuses narrowly on the science-related issues that the message emphasizes. These findings suggest, if the objective is shaping public opinions and the political environment in a way that is conducive to scientific endeavours and evidence-based policies, scientific organizations giving political endorsements may have substantial downsides and little upsides.

The findings are not a results of Biden’s relative unpopularity since late 2021. The experiment was conducted in late July and early August 2021, when Biden’s approval rating was above 50% and only slightly below its peak[33]. Further, a pilot study I conducted found qualitatively similar effects in May 2021, when Biden’s overall popularity was still at its “honeymoon” level, with 71% of Americans, including 47% of Republicans, approved his handling of COVID[34].

The findings have implications beyond this specific context. Education has been playing an increasingly important role in political alignment across Western democracies over the past several decades[35]. In such settings, scientists and science skeptics are likely, more than ever, represented by opposing sides of the partisan divides. Dynamics similar to the one studied here may thus play out in other times and places[36][37].

## 4 Methods

### 4.1 Pre-Registration

The experiment is double-registered as AEARCTR-0007007 (American Economics Association RCT Registry) and 10.17605/OSF.IO/SJUPQ (Open Science Foundation Registry). The former was pre-registered before the experiment launched; the latter was pre-registered during the response collection period and before I first accessed or analyzed any portion of the data. The contents of the two registrations do not have any substantive differences.

The only deviation from the pre-registered protocol is the sample size. The pre-registration states that 4,000 responses would be collected, that is, after attention screening but before excluding participants who are neither “Biden supporters” nor “Trump supporters” from analyses. Although I explicitly requested only 4,000 responses from Lucid Theorem, by the time I downloaded and started analyzing the data, 4,260 responses had already been collected. All analyses presented are based on this larger-than-expected sample. This deviation is small in magnitude, and restricting my analysis to the first 4,000 responses collected does not alter the findings in any substantively meaningful way.

Two additional analyses specified in the pre-analysis plan, namely multiple testing adjustments and lasso regression adjustments with covariates, are presented in the Appendices A and B to keep the main text concise, as the findings are not meaningfully different from what is in the main texts.

## 4.2 Sample

The experiment took place between July 28 and August 10, 2021 and was conducted using a national (U.S.) sample of 4,260 individuals who had been screened for attention. Subjects were invited to share their opinions “in a survey about current events” and participated voluntarily by signing a consent document. The sample was recruited from Lucid, which helps researchers build samples representative of the U.S. adult population along targeted demographic dimensions, with two caveats: First, the distribution of characteristics *not* targeted by Lucid Theorem may not reflect that of the population. For example, Democrats tend to be over-represented in Lucid Theorem samples. Second, the attention check I placed, which screens out inattentive subjects, could potentially alter the distribution if attention is correlated with other subject characteristics.

In practice, however, the second caveat does not seem to pose a meaningful challenge to the representativeness of my sample. Supplementary Table 1 shows post-attention check sample breakdowns by key demographic and geographical categories, which broadly mirrors that of the U.S. adult population. However, as Supplementary Table 2 shows, the sample does not seem representative with respect to political leaning. Specifically, Biden supporters are likely over-represented, if the 2020 presidential election result is taken as the benchmark. It is worth pointing out that this skewed distribution is not the result of screening for attention, but rather due to Democrats being regularly over-represented in Lucid Theorem samples.

## 4.3 Subject attention

The sample described above is the result of screening out inattentive respondents. A grid of questions are placed before the treatment and consists of three substantive questions and three attention checks. To pass the check, a participant must answer all three attention checks “correctly”. The attention check asks if subjects agree or disagree with the following statements (1) “Vermont is LESS populous than California.” (passable answers: “Strongly Agree”, “Agree”, “Neither Agree Nor Disagree”, or “Don’t Know”); (2) “Please select ‘Disagree’” (passable answer: “Disagree”); and (3) “Two plus one equals three”, (passable answers: ‘Strongly Agree’ or “Agree”). To make the attention screening less apparent to inattentive subjects, the survey mixes these questions with three substantive questions about subjects’ interests in science and current events. The check had a passing rate of 72.29% - of the 6,170 respondents who entered the survey, 1,710 failed the attention checks and were screened out from the study. Of the 4,460 subjects who passed the attention heck, 200 did not complete the survey and were excluded from the analyses.

## 4.4 Randomization

The treatment was randomized on the individual level and blocked on prior political position (Supplementary Table 2) to ensure finite sample balance.

334 The randomization was implemented with Qualtrics randomizers. There are  
335 five blocks. One half of the subjects in each block received the treatment.

## 336 4.5 Analysis

337 Throughout the analyses, I focus on heterogeneity by subjects' baseline  
338 political opinion, particularly by whether they support Biden or Trump. This  
339 opinion is elicited by a pre-treatment question about their "hypothetical vote  
340 intention" asking who they would vote for if they "were to choose again"  
341 between Biden and Trump (since the study took place many months after  
342 the election). The distribution of answers among subjects who passed the  
343 attention check is reported in Supplementary Table 2. A majority 55.14%  
344 of the subjects responded either "Definitely Biden" or "Probably Biden". I  
345 label them "Biden supporters". 35.06% favoured Trump and are thus labelled  
346 "Trump supporters". 8.80% answered "Someone else". In my analysis, I drop  
347 the 8.80% who stated that they would vote for "someone else", since there is  
348 no clear interpretation for their pre-treatment political alignment.

### 350 4.5.1 Statistical information

The experimental data is analyzed by estimating the following linear regression model:

$$Y_i = \alpha + \beta D_i \times TS_i + \gamma D_i \times (1 - TS_i) + \delta TS_i + \epsilon_i \quad (1)$$

351 where  $Y_i$  is the outcome for subject  $i$ ;  $D_i$  is a dummy variable indicating  
352 subject  $i$  is in the treatment group;  $TS_i$  is a dummy variable indicating  $i$  is a  
353 Trump supporter;  $\epsilon_i$  is a heteroskedastic error term.

354 This specification is chosen, among various equivalent models, for its  
355 ease of interpretation. In particular, since I drop subjects who are coded as  
356 neither "Trump supporter" nor "Biden supporters", the omitted category is  
357 control Biden supporters.  $\beta$  and  $\gamma$  thus represent treatment effects for Trump  
358 supporters and for Biden supporters respectively.  $\delta$  is the baseline difference  
359 between the two categories. In Table 1, estimates of  $\beta$ ,  $\gamma$  and  $\delta$  are presented  
360 in columns 1, 2, and 3, respectively. The analyses reported in Table 1 and the  
361 main text do not use any covariates in additions to the variables included in  
362 Equation 1.

## 364 4.6 Manipulation check

365 The following manipulation check is placed after the last outcome question,  
366 near the end of the survey, to examine if the treatment is successful at  
367 informing subjects of *Nature's* endorsement.

368 Did the scientific journal *Nature* make any explicit political statements in support  
369 of any candidates in the run-up to the 2020 presidential election?

- 370 • No, they did not.

- 371 • Yes, they endorsed Joe Biden.
- 372 • Yes, they endorsed Donald Trump.
- 373 • Don't know.

374 Experimental data shows that the treatment produced a large and statisti-  
375 cally highly significant shift in the rate at which subjects answer “Yes,  
376 they endorsed Joe Biden.”. Among Biden supporters, 68.79% treated subjects  
377 correctly answered that *Nature* endorsed Biden - a 43.47-percentage point  
378 shift ( $p < 0.001$ , two-tail t test) relative to control Biden supporters, who  
379 answered correctly 25.32% of the time. The magnitude of the shift is remark-  
380 ably similar among Trump supporters, albeit from a lower baseline. Treated  
381 Trump supporters answered correctly 58.57% of the time, 45 percentage points  
382 ( $p < 0.001$ ) more frequent than the 13.58% baseline among control Trump  
383 supporters. Despite starting from different baselines, Biden supporters and  
384 Trump supporters were similarly receptive to the message - there is no statisti-  
385 cally significant difference in the magnitude of the shift between Biden and  
386 Trump supporters ( $p = 0.594$ ). Thus, any heterogeneity in treatment effects  
387 on outcomes is not due to differential compliance to the treatment.

388 **Supplementary information.** Supplementary information is provided in  
389 the form of a pdf file, which contains supplementary tables reporting sample  
390 summary statistics and additional analyses (lasso regression adjustment and  
391 multiple comparison correction, see Appendices A and B for descriptions) and  
392 a copy of the survey instrument.

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399 **Author contribution.** Floyd Jiuyun Zhang conceived, conducted, ana-  
400 lyzed, and interpreted this study, as well as wrote this manuscript.

## 401 Declarations

402 The author is not aware of any competing interests.

403 The ethics of the study protocol was approved by Stanford University  
404 Research Compliance Office (Protocol number: 60462).

## 406 Appendix A Lasso regression adjustment

407 During the experiment, a rich set of covariates were collected, including  
408 information on the subject's political ideology, age, gender, U.S. state of res-  
409 idence, race/ethnicity, education attainment, subject of post-secondary study

410 (STEM or not), religion, residential setting (urban/suburb/rural), interest in  
 411 current events, and interest in popular science reading. For simplicity and  
 412 transparency, analyses presented in the main text do not make use of these  
 413 information. In this appendix, I present treatment effect estimates with regres-  
 414 sion adjustments using these information to reduce estimator variance.

415 To efficiently select relevant covariates for each outcome, I use the lasso  
 416 regression[38] to regularize the coefficients, in accordance to my pre-analyzed  
 417 plan. Specifically, I employ the lasso regression adjustment method proposed  
 418 by Wager *et al*[39] to separately estimate the treatment effects on Biden sup-  
 419 porters and on Trump supporters. The results are presented in Supplementary  
 420 Table 3. The adjustments reduce standard errors modestly and do not mean-  
 421 ingfully alter the point estimates. The significance levels of the treatment  
 422 effects are mostly unchanged, with two exceptions. Without adjustments, the  
 423 estimated effects on trust in *Nature*'s impartiality and U.S. scientists' com-  
 424 petence for Biden supporters are statistically insignificant at the conventional  
 425 level ( $p = 0.151$  and  $p = 0.148$  respectively). With lasso regression adjust-  
 426 ments, they become significant at the 0.05 ( $p = 0.032$ ) and 0.10 ( $p = 0.067$ )  
 427 levels, respectively.

## 428 **Appendix B Multiple Testings**

429 Supplementary Table 4 reports results of adjusting analyses in the main text  
 430 for multiple testing using sharpened False Discovery Rate (sharpened FDR)  
 431  $q$ -values[40][41].

432 As specified in my pre-analysis plan, I divide all ten outcomes into five  
 433 mutually exclusive and collectively exhaustive sets, namely "stated trust in  
 434 *Nature*", "endorsement persuasiveness", "demand for information", "climate  
 435 Change", and "trust in scientists in general" and, *within each set of out-*  
 436 *comes*, I adjust for multiple testing for each of the two following clusters of  
 437 hypotheses: "the treatment effect for Trump supporters is zero" and "the  
 438 treatment effect for Biden supporters is zero". This groups 20 hypotheses  
 439 about treatment effects into 10 clusters, of which two clusters, namely Trump  
 440 supporters' and Biden supporters' demand for information (from *Nature*),  
 441 are singletons containing only one hypothesis each and are thus *not* adjusted.  
 442 Unlike the treatment effects, baseline differences between Biden supporters  
 443 and Trump supporters are not parts of my pre-registered hypotheses, but  
 444 analogous sharpened FDR  $q$ -values are also reported for completeness. As  
 445 Supplementary Table 4 makes clear, the corrections do not alter change the  
 446 results in any meaningful way.

## 448 **Codes availability statement**

449 Data necessary for interpreting and replicating this study will be made publicly  
 450 available by the final submission (if accepted). It will also be made available  
 451 earlier during the review process at editors' and/or reviewers' request.

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Codes necessary for interpreting and replicating this study will be made publicly available by the final submission (if accepted). It will also be made available earlier during the review process at editors' and/or reviewers' request.

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