Politics and science have become increasingly intertwined. Salient scientific issues such as climate change, evolution, and stem cell research become politicized, pitting partisans against one another. This creates a challenge of how to effectively communicate on such issues. Recent work emphasizes the need for tailored messages to specific groups. Here, we focus on whether generalized messages also can matter. We do so in the context of a highly polarized issue – extreme COVID-19 vaccine resistance. The results show that science-based, moral frame, and social norm messages move behavioral intentions, and do so by the same amount across the population (i.e., homogenous effects). Counter to common portrayals, the politicization of science does not preclude using broad messages that resonate with the entire population.

In 2010, the journal *Nature* published an editorial stating, “there is a growing anti-science streak...that could have tangible societal and political impacts” (2010: 133). About a decade later, the foreward to Naomi Oreskes book *Why Trust Science?* begins with “Science confronts a public crisis of trust” (Macedo 2019: 1). The widespread concern is that politics now overwhelms science. This makes the communication of science – on topics such as climate change, evolution, genetically modified foods, obesity, stem cell research, and more – a political challenge. COVID-19 reactions epitomized politicization with partisanship becoming such a driver of health decisions that it was “pernicious enough to threaten the health of citizens” (Gollwitzer et al. 2020: 1195). A burgeoning literature explores three distinct approaches to communication on scientific topics in politicized settings (Bayes et al. 2020), with each facing major political hurdles due to polarization.

First is research examining how scientific experts can influence opinions. While some have shown these effects (van der Linden et al. 2015), others question the influence of communications by scientists and/or experts (i.e., when such sources serve as *expert cues*) (Akin and Scheufele 2017, 25). The pessimism stems from an expectation that, on politicized topics, ideology and/or partisan identities lead people to reject science communication counter to their beliefs and identities (e.g., Kahan 2015). An alternative approach in light of these challenges is *framing* moral values, such as patriotism or harm prevention, to shift preferences among those who are sensitive to the value in question (Feinberg and Willer 2019). This work builds on moral foundations theory that suggests frames need to resonate with values that differ by party, such as appealing to Republicans’ concern for patriotism and Democrats’ concern for harm prevention (Wolsko, Ariceaga, and Seiden 2016). Finally, recognizing that people often emulate others, studies show that appealing to *descriptive norms* – what people in one’s surroundings do – can alter attitudes and behaviors (Cialdini 2007; Dwyer, Maki, and Rothman 2015; Moehring et al. 2021) even when complying with norms incurs personal costs (Pickup, Kimbrough, and de Rooij 2021). Even here, though, Raymond and colleagues (2021: 11) explain, “Intense partisanship can make it more important to emphasize the prevalence of a behavior or belief within the relevant political group.”
Scholars have followed suit, focusing on how targeted messages influence particular groups. As a 2017 National Academies of Sciences consensus study emphasizes, “Tailoring scientific messages for different audiences is one approach to avoiding a direct challenge to strongly held beliefs...” (56). Vaccine communication studies explore subgroup appeals based on factors such as partisan cues (Pink et al. 2021), values (Lunz Trujillo et al. 2020; Bokemper et al. 2021), and needle sensitivity (Lunz Trujillo et al. 2020). While such targeted appeals are effective, they present practical trade-offs. Targeted messaging requires the acquisition of curated information and the ability to reach the targeted audience. When targeted messages reach non-targeted audiences, they can backfire (Hersh and Schaffner 2013).

Here, we explore the extent to which generalized communication can be effective. We do so in the challenging domain of COVID-19 vaccine uptake. This is a difficult area given extreme differences in attitudes about COVID-19 vaccines – particularly at the initial stages of their distribution. Differences in baseline vaccine attitudes emerged based on age, education, income, race, and gender (Lazer et al. 2021). Perhaps most importantly, partisanship remained the most stable and sizable gap, exemplifying the extreme politicization that defined COVID-19 (Green et al. 2020; Clinton et al. 2021; Druckman et al. 2021).

Even so, Motta et al. (2021) and Palm, Bolsen, and Kingsland (2021) conducted foundational studies that suggest general messages can work, with a focus on COVID-19 vaccine uptake prior to FDA approval. These authors report a host of messages – such as safety, social norms, values, and health consequences – have effects across party lines. Our work builds on theirs in three distinct ways. First, we collected data immediately after FDA vaccine approval, when the decision was no longer hypothetical. At the same time, politicization was extremely high and trust in health institutions was falling (Hegland et al., Forthcoming). Second, we focus on the aforementioned three messaging approaches previously identified in the literature with ecologically realistic messages that overlap with but do not match the other studies. Third, we systematically test a wider range of potential moderators that coincide with the aforementioned gaps; we do so by using recently-developed machine learning methods. In sum, we offer a novel test of generalized messaging that includes three central science communication approaches and a large set of possible moderators, during a time of extreme politicization when individuals faced an immediate pending decision regarding the vaccine. Our central question, therefore, is whether non-targeted approaches, crafted for general audiences, can be effective in such politicized situations – even influencing partisans from both sides.

Before turning to our studies, we offer two caveats. First, vaccine intentions emerge from a complex set of psychological, social, and institutional factors (Brewer et al. 2017); we focus on the effects of communications, notwithstanding other influences. Second, we recognize targeted messages can play an essential role; we simply seek to explore whether generalized messages – in our case based on science cues, moral frames, or descriptive norms – can as well.

Data and Methods

We conducted an online survey experiment (N = 24,682) from December 16, 2020 to January 10, 2021. Thus, our data collection began after the FDA’s December 11 approval of the Pfizer vaccine and just before the December 18 approval of the Moderna vaccine. This makes the decision to get vaccinated no longer hypothetical (as in prior studies), but one that respondents faced. We recruited via PureSpectrum, an online survey panel aggregator, using quota sampling to approximate the race/ethnicity, age, and gender distributions within each state. A total of
24,682 respondents who passed two closed-form attention checks and one open-ended attention check and did not indicate that they had already been vaccinated against COVID-19 were retained for analysis.

We randomly assigned participants to read a treatment message that provided a rationale to get the vaccine, or to a control condition with no message. For the expert cue messages, we employed two distinct treatments. One suggested that most scientists recommended taking the vaccine. This is the type of message that those who emphasize a need for targeted approaches would not expect to be broadly effective since there is some skepticism at the efficacy of relying on scientific perspectives (National Academies of Sciences, Engineering, and Medicine 2017). The other provides cues from the respondent's physician, in line with work suggesting doctors – and in our case, a doctor close to the respondent – could play an important persuasive role (Uslu et al. 2021). The moral framing messages follow prior work by appealing either to one's duty to what is right for the country (patriotism) or to preventing harm to themselves and others. Prior research typically identifies conservatives as being more sensitive to patriotism and liberals as more sensitive to harm prevention (e.g., Feinberg and Willer 2013). We are interested in whether these values only work among particular ideological subgroups, or whether their effects are generally similar across the population. Finally, the fifth treatment appealed to descriptive norms by posing the hypothetical where most people the respondent knows have taken the vaccine (Jaeger and Schultz 2017). Additional details regarding the survey and exact question wordings used in the experiment are in the Appendix. Even though the aforementioned prior work suggests possible heterogeneous effects, we note that the messages were designed to be general.

All respondents then reported how likely they would be to take the vaccine on a 7-point scale from extremely unlikely to extremely likely. For our main analysis, we make this outcome binary, considering respondents to be vaccine “resistant” if they report that they are “extremely unlikely” to take the vaccine. Those who are vaccine resistant are likely to reject vaccination even as vaccination norms spread (Palm, Bolsen, and Kingsland 2021), posing challenges for achieving herd immunity. That said, we conduct a parallel analysis using the full 7-point scale as a continuous outcome in the Appendix, and we find the same substantive results. We estimate average treatment effects overall and in comparison with one another, adjusting for multiple comparisons (Bretz, Hothorn, and Westfall 2011). This establishes which messages are effective for reducing vaccine resistance, and which (if any) are significantly more effective than others on average.

We then test for the extent to which any of these treatment effects systematically vary over a wide array of demographic covariates including race, gender, age, education, household income, partisan and ideological identities, and the severity of the COVID-19 outbreak in respondents’ counties (see above discussion of vaccine gaps). See Appendix for full list of variables included in main and supplemental specifications (e.g., social media exposure). Rather than pre-specifying which of these covariates ought to be associated with treatment effects, we systematically explore the data for such effects using the generalized random forest (Athey, Tibshirani, and Wager 2019; Wager and Athey 2018). This is an extension of the commonly-used random forest algorithm (Breiman 2001) that, rather than maximizing the difference in outcomes at each split, maximizes the difference in average treatment effects at each split. The algorithm guards against overfitting by randomly partitioning the observations into a splitting subsample which is used to fit the tree in a given iteration, and an estimating subsample that is used to derive the predicted effects from that tree for that iteration. Put another way, a tree in the algorithm satisfies the “honesty condition” when each observation is used to either determine splits in the tree or estimate effects, but not both (Wager and Athey 2018). This means that any conditional average treatment effects that are
identified by the algorithm are those that were robust to confirmatory, held-out estimation. In addition, it allows for predicted treatment effects at the individual level.

Results

As shown in Figure 1, 22% of respondents exhibit vaccine resistance when receiving no message (in the control), but this percentage significantly decreases in every treatment condition except patriotism, which falls just short of our threshold for statistical significance (adjusted $p = .096$). For example, the percentage drops to 19% (adjusted $p < .01$) when respondents are given the harm reduction message and to 17% (adjusted $p < .001$) in the physician endorsement condition. Given the relatively low baseline levels of resistance in the control group, this five percentage point decrease corresponds to a 23% reduction in vaccine resistance.

Figure 1: Average treatment effects for vaccine resistance. Control estimate with 95% uncertainty interval shown with dashed line in shaded band; treatment condition estimates with 95% uncertainty intervals shown with point ranges. Effects significant at $p < .05$ (adjusted for multiple comparisons) darkened.

Also of particular note is that the messages that moved people the most were those that cued the respondent’s physician or the scientific community. As shown in Figure 2, these have significantly greater effects than the patriotism message, though they are not significantly different in effect relative to the messages concerning harm reduction and people the respondent knows (the next-largest difference is between messages concerning respondents’ personal physicians and people they know with an adjusted $p$-value of 0.075). This result cuts against the prevalent sentiment that “people rarely make decisions based only on scientific information” (e.g., National Academies of Sciences, Engineering, and Medicine 2017, 3). Rather, our findings suggest that circumstances involving high personal stakes, such as potentially severe health consequences, may prompt citizens to defer to subject matter experts. This is true even in politicized settings. The result also contrasts with work suggesting norms play a more powerful role in messaging on scientific issues (Bayes et al. 2020). In our results, descriptive norms do significantly reduce vaccine resistance, but the scientific and medical communities hold sway as well.
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Figure 2: Testing for differences between treatment effects. Cells represent differences in vaccine resistance between primary (y-axis) and comparison (x-axis) conditions. Differences significant at $p < .05$ (adjusted for multiple comparisons) darkened.

Table 1: Proportions of predicted individual-level effect types in each condition, vaccine resistance

<table>
<thead>
<tr>
<th>Condition</th>
<th>Negative</th>
<th>Positive</th>
<th>Null</th>
<th>Above Average</th>
<th>Below Average</th>
</tr>
</thead>
<tbody>
<tr>
<td>Physician Recommend</td>
<td>0.366</td>
<td>0</td>
<td>0.634</td>
<td>0.025</td>
<td>0.018</td>
</tr>
<tr>
<td>Scientists Recommend</td>
<td>0.359</td>
<td>0</td>
<td>0.641</td>
<td>0.025</td>
<td>0.026</td>
</tr>
<tr>
<td>People You Know</td>
<td>0.165</td>
<td>0</td>
<td>0.835</td>
<td>0.012</td>
<td>0.01</td>
</tr>
<tr>
<td>Preventing Harm</td>
<td>0.114</td>
<td>0</td>
<td>0.886</td>
<td>0.002</td>
<td>0.003</td>
</tr>
<tr>
<td>Patriotism</td>
<td>0.057</td>
<td>0.001</td>
<td>0.942</td>
<td>0.006</td>
<td>0.003</td>
</tr>
</tbody>
</table>

Next, we explore whether messages had differential effects on particular subgroups of our sample using the generalized random forest. We find very little evidence of heterogeneous effects. Table 1 shows, for each treatment-control comparison, the proportion of predicted individual-level treatment effects that are significantly distinguishable (or not) from zero, as well as the share that are statistically distinguishable from the average treatment effect. The table shows that treatment effects are substantively homogeneous at the individual level. Virtually everyone in the population is predicted to react similarly to each message.

This is apparent in two respects. First, predicted effects go in consistent directions. For essentially every respondent where treatment effects are predicted to be statistically distinct from zero, the effects are associated with reductions in vaccine resistance. Second, extremely few respondents are predicted to have treatment effects that significantly differ from the overall average effects. The highest such shares are in the resistance outcome for physician and scientists conditions, in which just under five percent of respondents are predicted to respond to the message either more or less strongly than the average respondent.

Neither political ideology or partisanship moderated the moral framing messages in a systematic manner, which is perhaps the greatest departure from expectations implied by previous literature (Feinberg and Willer 2019) and is discussed further in the Appendix. Instead, the characteristic that we find moderates effects to the most substantial degree is household income – and only in select conditions. In the maximally heterogeneous slices of these data – household
income moderating the effects of the physician and scientists treatments (shown in Figure 3a for the resistance outcome; corresponding tests on the likelihood outcome show substantively similar results) – household income is clearly associated with the rank-ordering of predicted effects (Figure 3b). However, as indicated by Table 1, scarcely any of these respondents are predicted to be significantly more or less sensitive to the given message than the average respondent.

Moreover, these differences are primarily attributable to a ceiling effect: fewer high-income respondents report vaccine resistance in the control condition, leaving less room for their counterparts in treatment conditions to exhibit differences. This is shown in Figure 3c, which shows control condition means and group average treatment effects by household income quartile. Those in the lowest (highest) quartile are predicted to see significantly higher (lower) than average reductions in vaccine resistance, but they are also starting from a higher (lower) than average baseline. Put simply, higher-income respondents have lower predicted reductions in vaccine resistance because they are already reporting lower rates of vaccine resistance in the control condition, not because they are less sensitive to pro-vaccine messaging.

**Discussion**

Our results have three significant implications. First, in politicized science environments, succinct messages crafted for general audiences can work, having homogenous effects across a range of respondent characteristics including partisanship and ideology. That is, while there are significant differences in baseline attitudes concerning COVID-19 vaccination by political identities, there are not significant differences in sensitivity to pro-vaccine messages. The findings reveal the types of communication that can be effective among entire populations (Druckman and Lupia 2017). This should not be taken as a criticism of targeted messages, which have a crucial role to play. From a normative perspective, our results reveal a downside insofar as the groups most resistant to vaccines (e.g., Republicans) are not affected more than those less resistant (e.g., Democrats). In that sense, homogenous effects are not always ideal and generalized messages need to be joined with targeted appeals. Understanding sources of heterogeneity matter (Callaghan et al. 2019). More generally, Callaghan et al. (2021) show that if marginalized populations do not have their concerns precisely addressed, then inequalities can be exacerbated. Our point is to clarify the practical usefulness of generalized approaches as part of a communication strategy.

Second, despite pessimism that direct messages from scientists or other expert cues can broadly persuade (National Academies of Sciences, Engineering, and Medicine 2017), we find that they can — in a highly politicized context and even among conservatives who tend to exhibit less trust in science (Gauchat 2012; although see Lee 2021). In fact, we find that cuing subject matter expertise prove most effective in the case of COVID-19 vaccines, reducing vaccine resistance by as much as 23%. To be clear, our findings reveal that expert cues matter relative to no message; they do not have a significantly stronger effect than messages from lay sources (also see Motta et al. 2021). The same homogeneity applies for the other message types, which as explained, many view as constrained to precise populations (also see Motta et al. 2021 and Palm et al. 2021). That said, it remains an open question whether these messages would stand up to competing communications, particularly those that introduce scientific uncertainty or attempt to undermine trust in expertise (Druckman 2017; Bolsen and Druckman 2015, 2018; Merkley 2020). Future work also would benefit from exploring whether the psychological processes underlying these distinct messaging approaches are similar to one another.

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1We thank an anonymous reviewer for this point.
Figure 3: Testing for treatment effect heterogeneity. Most important variables (a), sorted individual treatment effects (b), and group average treatment effects by household income quartile (c) for the resistance outcome.
Third, we began the paper by pointing out the possibility that partisan or ideological priors limit the potential for homogenous effects. In contrast, using recently-developed methods at the intersection of causal inference and machine learning, we find generalized messages exhibit homogenous effects not only with regard to partisanship and ideology but also a host of other variables. This echoes recent political science work (Coppock, Leeper, and Mullinix 2018; Motta et al. 2021; Palm, Bolsen, and Kingsland 2021) but using a distinct method and a large set of covariates. Surely heterogeneity exists with other message types and/or covariates, but it is not inevitable. In sum, rather than working to identify a variety of successful messages, each tailored to a particularly responsive subgroup, our results suggest it can be more efficient to identify the most persuasive messages overall and broadcast them widely.

Backmatter

Supplementary Material

Supporting information for this manuscript will be accessible online on the journal website upon publication.

Data Availability

All data and code necessary to reproduce the results reported in this manuscript and all supporting materials will be posted on the Harvard Dataverse upon publication.

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Competing Interests

Dr. Perlis has received consulting fees from Burrage Capital, Genomind, RID Ventures, and Takeda, outside of the present work; he holds equity in Psy Therapeutics and Outermost Therapeutics, outside of the present work. No other authors have potentially competing interests to declare.

Ethical Standards

This research was approved by the institutional review boards at Rutgers University (#Pro2020000977), Harvard University (#IRB20-0593), and Northeastern University (#20-04-12)
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