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The role of political devotion in sharing partisan misinformation

Political Devotion & Misinformation

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Abstract

Online misinformation poses a significant threat to global challenges, such as the COVID-19 pandemic and climate change. Misinformation is disproportionately shared by people with extreme political attitudes, especially among the far right. To understand the psychological and neurocognitive processes that underlie misinformation sharing among extremists, we conducted a pre-registered, cross-cultural experiment with conservatives and far-right supporters in the US and Spain (N = 1,609) and a neuroimaging study with far-right supporters in Spain (N = 36). Individuals who felt their personal identity was fused with their political group were more likely to share misinformation, especially when the misinformation was related to issues that involve sacred moral values (e.g., immigration and nationalism). Analytical thinking was unrelated to misinformation sharing when the misinformation involved sacred values (vs. non-sacred values) and fact-checks had little or no effect in this sample, especially among hyper-partisans. Far-right supporters also showed increased activity in brain regions associated with theory of mind

34 in response to posts with sacred values, highlighting the social dimension of
35 misinformation sharing. These results suggest that political devotion plays a key role in
36 misinformation sharing and that identity-based interventions may help curb misinformation
37 for specific groups.

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39 Introduction

41 As social media increasingly becomes our main source of news¹, malicious agents
42 are using this opportunity to spread misinformation to larger audiences faster than ever
43 before^{2,3}. Attempts to quantify this phenomenon suggest that 70% of Twitter users are
44 exposed to fake news⁴, and fake news is 70% more likely to be shared than true news,
45 consistently spreading more quickly and more broadly⁵. Critically, online misinformation
46 can impact real-world outcomes such as fueling ideological polarization⁶⁻⁸ and reducing
47 vaccination intentions⁹. Thus, it is critical to understand the psychological processes
48 associated with sharing online misinformation as well as potential interventions that may
49 help counteract its spread.

50 One of the major frameworks currently employed to understand why individuals
51 share misinformation is that they lack the skill or motivation to engage in analytical
52 reasoning¹⁰. Several studies have found that a lower propensity to engage in analytical
53 thinking is associated with poor truth discernment in news headlines and that accuracy
54 nudges improve sensitivity to falsehood¹¹. Other studies find no relationship between
55 cognitive reflection and misinformation sharing¹² and the impact of accuracy nudges may be
56 quite weak¹³. While correctly identifying a headline as false may stop analytical thinkers
57 from sharing it, it does not explain what makes them *want* to share it in the first place. A
58 recent meta-analysis of this body of work found that these accuracy nudges had little or no
59 effect on misinformation sharing among far right-wing participants¹⁴, suggesting that
60 political ideology or identity may play a key role in the spread of misinformation (see also
61 Guess et al.¹⁵).

62 In this respect, partisanship or identification with a political party has been identified
63 as one of the main factors driving belief and willingness to share misinformation^{12,16,17}.
64 According to the identity-based model of political belief¹⁸, individuals are more inclined to
65 believe (mis)information that affirms their partisan identities; that is, beliefs that serve

66 belonging, epistemic, status and moral goals associated with their social identity. Thus, for
67 accuracy nudges to work, the benefits of being accurate need to outweigh the partisan
68 motives, or accuracy needs to be linked to a particular social identity (e.g., as with scientists).
69 The model also makes key predictions about the neural processes underlying partisan
70 motivations. Particularly, these value computations may be mediated by the orbitofrontal
71 cortex¹⁸, a richly interconnected brain region thought to integrate overall value during
72 decision-making and generate evaluations^{19,20}.

73 The incentives of identity-motivated cognition may be especially relevant for hyper-
74 partisans or individuals with extreme identities. Research on extremism often refers to
75 extreme partisans as “devoted” actors²¹, individuals who are characterized by two identity
76 motives: identity fusion and willingness to sacrifice for sacred values. Identity fusion is a
77 visceral feeling of oneness with a group that results from the merging between their social
78 and personal identities²². Unlike group identification²³, which involves a process of
79 depersonalization and adjustment to group norms²⁴, identity fusion allows preserving a strong
80 sense of self. It is precisely this heightened sense of self that enables devoted actors to initiate
81 extraordinary actions such as risking their lives on behalf of the group²². Moreover, devoted
82 actors are more likely to engage in these behaviors when the values at stake are perceived to
83 be sacred²¹. Sacred values are strongly held beliefs resistant to economic trade-offs^{25,26} that
84 often lie at the heart of intractable conflict²⁷, e.g., the “holy land” of Israel. The non-
85 negotiable status of sacred values has been conceptualized as inherently linked to their
86 identity-defining nature (e.g., “This is who we are and what we stand for”, see Sachdeva et
87 al.²⁸). If devoted actors are willing to risk their lives to fight for sacred values, are they also
88 willing to share misinformation online to promote their cause, even at the cost of factual
89 truths?

90 We propose that spreading partisan misinformation can be understood as a
91 consequence of political devotion. In line with this, we suggest that the two components of
92 political devotion— identity fusion and sacred values—are critical identity motives that drive
93 engagement with partisan misinformation. In terms of the identity-based model of political
94 belief¹⁸, identity fusion can be understood as a form of extreme partisanship, where sacred
95 values are highly moralized beliefs that serve partisan goals. If identity motives play any role
96 in misinformation sharing, then two consequences should follow. First, partisan goals should

97 be weighted more heavily than accuracy goals among extremists (e.g., fused individuals) and
98 whenever highly relevant partisan goals are invoked (e.g., misinformation appeals to sacred
99 values), which should lead to increased misinformation sharing. Second, when it comes to
100 behavioral interventions, attempting to enhance accuracy concerns, for instance, by using
101 fact-checks or accuracy nudges against misinformation, should be less effective when
102 identity motives are most salient (e.g., in fused individuals and for misinformation on sacred
103 values). Ultimately, sharing of partisan misinformation should be higher whenever and for
104 whomever identity motives overshadow accuracy concerns.

105 **Current research**

106 In the present work we explore how the two fundamental identity motives of political
107 devotion, sacred values and identity fusion, impact people’s likelihood of sharing
108 misinformation across two countries (the US and Spain). In this context, we evaluate whether
109 these identity motives are associated with higher likelihood of sharing misinformation and
110 reduced effectiveness of interventions aimed to increase accuracy concerns and reduce
111 misinformation sharing. Specifically, we assessed the efficacy of different popular
112 interventions, including *fact-checks* used by social media companies (“This claim about...is
113 disputed.”), *accuracy nudges* supported by widely cited papers (“To the best of your
114 knowledge, is the above statement accurate?”, based on Pennycook et al.¹¹), and *media*
115 *literacy nudges* (“What techniques are used in this Tweet to attract your attention?”).

116 Because conservatives and far-right supporters disproportionately share far more
117 misinformation^{4,15,29}, we conducted two pre-registered experiments with these populations in
118 the US and Spain (N = 1,609). Importantly, because extreme partisanship is also possible in
119 the political left, our research does not preclude that identity motives similarly influence
120 misinformation processing in the far left. Moreover, to better understand the neurocognitive
121 processes underlying misinformation sharing, we conducted a neuroimaging study with far-
122 right supporters in Spain (N = 36). The neuroimaging study aimed to identify which brain
123 networks are involved in processing misinformation related to sacred (*vs.* non-sacred) values,
124 as well as to evaluate the brain response to the tested intervention. For instance, interventions
125 aimed to reduce misinformation by increasing analytical thinking should be associated with
126 increased activity in brain regions that support cognitive control such the dorsolateral

127 prefrontal cortex³⁰, or prediction error such as the anterior cingulate, which have been
128 associated with subsequent behavioral adjustment³¹. Moreover, if an intervention is
129 ineffective, the recorded brain response could provide clues as to why it is ineffective (e.g.,
130 the neurofunctional analysis suggests different networks are active). Thus, understanding the
131 psychological motives and neuro-cognitive processes among extreme partisans will be
132 critical to understand one of the key sources of online misinformation.

133 To understand the spread of misinformation, we focused on false statements in this
134 research. We did not include true statements because we were primarily interested in testing
135 the impact of using sacred moral values as a vehicle to spread misinformation, rather than
136 trying to identify how people differentiate false from true information. Ultimately,
137 misinformation in its most compelling form is only distinguishable from true information in
138 that it does not correspond to reality. This choice has two implications: first, participants in
139 this series of studies were exposed to an unrealistic volume of misinformation, considering
140 that misinformation in the real world coexists with true information¹⁵, and second, our
141 findings do not preclude that appealing to sacred values is associated with higher likelihood
142 of sharing any type of information, regardless of whether it is true or false.

143 In line with the idea that identity motives should outweigh accuracy concerns
144 whenever highly relevant partisan goals are invoked, we predicted that misinformation
145 relevant to sacred values (*vs.* non-sacred values) would be associated with greater likelihood
146 of sharing, and that fused (*vs.* non-fused) individuals would be more likely to share partisan
147 misinformation. In addition, previous literature suggests that behavioral interventions aimed
148 to enhance accuracy concerns such as fact-checking and accuracy nudges should decrease
149 the likelihood of sharing misinformation. However, because sacred values are highly relevant
150 identity motives, they are particularly resistant to trade-offs and social influence³². Thus, we
151 expected it to be harder for interventions focused on enhancing accuracy to override identity
152 motives for misinformation relevant to sacred (*vs.* non-sacred values), leading to a reduced
153 effect of fact-checking in this condition. In terms of brain activity, we expected the
154 orbitofrontal cortex to be involved in sharing messages that include highly relevant identity
155 motives or sacred values in line with the higher subjective value involved in these decisions.
156 In terms of the brain response to the intervention, we expected to observe increased activity
157 in brain regions associated with cognitive control or prediction error as explained above.

158

159 **Results**

160 **The appeal of sacred values**

161 In two survey studies with Spanish (N = 812) and US conservatives (N = 797) and a
162 neuroimaging study with Spanish far-right supporters (N = 36) participants were asked to
163 rate the likelihood of sharing a series of social media posts composed by ingroup political
164 leaders (see demographic data in Table 1). The social media posts were specifically designed
165 to convey 8 false statements about sacred political issues (immigration, nationalism, and
166 women and family values) and 8 false statements about non-sacred political issues (roads and
167 infrastructure, foreign affairs, waste management, and materials).

168 As predicted, the presence of sacred values was associated with higher likelihood of
169 sharing misinformation across the three samples (Spanish sample: $M_{diff} = 0.54$, 95% CI [0.48,
170 0.60], $t(811) = 18.80$, $p < .001$, Cohen's $d = 0.41$, 95% CI [0.31, 0.51]; US sample: $M_{diff} =$
171 0.56 , 95% CI [0.50, 0.63], $t(796) = 16.03$, $p < .001$, Cohen's $d = 0.38$, 95% CI [0.28, 0.48];
172 fMRI sample: $M_{diff} = 1.14$, 95% CI [0.82, 1.47], $t(34) = 7.09$, $p < .001$, Cohen's $d = 1.06$,
173 95% CI [0.58, 1.60], see Table 2a and Fig. 1a). Moreover, the impact of sacred values
174 increased as a function of the number of sacred values held by participants (out of the three
175 proposed ones, see Fig. 1b) in both Spanish conservatives ($B = 0.18$, 95% CI [0.13, 0.24],
176 $t(810) = 6.60$, $p < .001$) and US conservatives ($B = 0.19$, 95% CI [0.13, 0.25], $t(795) = 6.41$,
177 $p < .001$), though the effect of number of sacred values was in the same direction, but not
178 statistically significant in the fMRI sample ($B = 0.14$, 95% CI [-0.13, 0.41], $t(33) = 0.97$, $p =$
179 $.34$). Thus, participants across the three studies were consistently more likely to share
180 misinformation about sacred issues (e.g., immigration) compared to non-sacred issues (e.g.,
181 infrastructure), especially if they personally held these values as sacred, and even if both
182 types of issues were formulated as a critique to the government or the status quo.

183 To test for potential confounds, we pilot tested the social media posts employed in
184 the US study in an independent sample of US conservatives (N = 80). We asked this sample
185 to rate sacred and non-sacred values on perceived accuracy, familiarity, salience, and attitude
186 strength (certainty, extremity, and importance; $\alpha = 0.87$). Social media posts relevant to
187 sacred (vs. non-sacred) values received similar ratings in perceived accuracy (sacred values:
188 $M = 3.28$, $SD = 0.69$; non-sacred values: $M = 3.14$, $SD = 0.66$; $M_{diff} = 0.14$, 95% CI [-0.01,
189 0.29], $t(72) = 1.80$, $p = .076$, Cohen's $d = 0.21$, 95% CI [-0.12, 0.53], see Fig. 1g), but were

190 rated higher in attitude strength (sacred values: $M = 3.59$, $SD = 0.77$; non-sacred values: M
191 $= 2.45$, $SD = 0.80$; $M_{diff} = 1.14$, 95% CI [0.95, 1.34], $t(72) = 11.72$, $p < .001$, Cohen's $d =$
192 1.46 , 95% CI [1.09, 1.82]), familiarity (sacred values: $M = 3.25$, $SD = 0.88$; non-sacred
193 values: $M = 2.17$, $SD = 0.96$; $M_{diff} = 1.09$, 95% CI [0.89, 1.28], $t(72) = 11.09$, $p < .001$,
194 Cohen's $d = 1.18$, 95% CI [0.83, 1.54]), and salience (sacred values: $M = 2.60$, $SD = 0.97$;
195 non-sacred values: $M = 1.86$, $SD = 0.80$; $M_{diff} = 0.74$, 95% CI [0.57, 0.91], $t(72) = 8.60$, $p <$
196 $.001$, Cohen's $d = 0.83$, 95% CI [0.49, 1.17]).

197 To assess and control the effect of each variable on participants' likelihood of sharing
198 social media posts relevant to sacred (vs. non-sacred) values in our main US study, we reran
199 the analyses accounting for the effect of each variable. Based on the pilot, we computed
200 average scores in accuracy, attitude strength, familiarity, and salience for social media posts
201 in the sacred versus non-sacred value condition for each participant in the US study. Because
202 different participants had been exposed to a different set of social media posts (8 out of 32
203 possible items), confound scores also varied across participants. When assessed separately,
204 all variables influenced likelihood of sharing social media posts (accuracy: $B = 1.13$, 95% CI
205 [0.94, 1.32], $t(950) = 11.67$, $p < .001$; attitude strength: $B = 0.78$, 95% CI [0.67, 0.89], $t(842)$
206 $= 13.88$, $p < .001$; familiarity: $B = 0.79$, 95% CI [0.68, 0.90], $t(857) = 14.14$, $p < .001$; and
207 salience: $B = 0.84$, 95% CI [0.72, 0.95], $t(867) = 13.83$, $p < .001$). However, when added into
208 the same model together with value sacredness, *value sacredness remained the only*
209 *significant predictor of likelihood of sharing social media posts* (value sacredness: $B = 0.62$,
210 95% CI [0.44, 0.80], $t(1091) = 6.78$, $p < .001$; accuracy: $B = 0.26$, 95% CI [-0.15, 0.66],
211 $t(967) = 1.24$, $p = .216$; attitude strength: $B = 0.02$, 95% CI [-0.45, 0.48], $t(1004) = 0.07$, $p =$
212 $.943$; familiarity: $B = 0.06$, 95% CI [-0.36, 0.48], $t(938) = 0.29$, $p = .772$; and salience: $B = -$
213 0.32 , 95% CI [-0.77, 0.12], $t(1140) = -1.42$, $p = .157$). Since both value sacredness and the
214 variable scores for each social media post were obtained from sample averages, differences
215 in individual-level versus sample-level measurements did not appear to influence these
216 results. Thus, sacred (vs. non-sacred) values were associated with increased likelihood of
217 sharing social media posts above and beyond differences in attitude strength, familiarity, and
218 salience.

219 The neuroimaging study revealed very strong brain activity in response to social
220 media posts containing sacred values compared to non-sacred values among far-right

221 supporters. Due to generalized activation across the brain at a standard threshold ($p < .001$
222 FWE cluster-level) in this contrast, we had to use a more stringent statistical threshold ($p <$
223 $.05$ FWE peak-level) to detect several prominent clusters of activity in the left middle
224 temporal gyrus, left dorsomedial prefrontal cortex, left precuneus, left middle and inferior
225 frontal gyrus, and the bilateral occipital cortex (thresholded at $T = 5.57$, $p < .05$ FWE, see
226 Fig. 2A and Table 3a). Based on the degree of overlap between the observed brain activity
227 pattern and the seven functional networks described in Yeo et al.³³, 56% of the identified
228 activity to sacred values overlapped with the default mode network, 28% with the visual
229 network, and 12% with the frontoparietal network (the remaining 4% was distributed across
230 the somatomotor network, the dorsal and ventral attention networks, and limbic areas). A
231 location analysis with *Neurosynth*³⁴ indicated that the peak activations in each cluster were
232 broadly associated with terms related to language and social inferences (see Supplementary
233 results) and the obtained pattern of activation considerably overlapped with a *Neurosynth*
234 mask associated with *theory of mind* (see Fig. 2d).

235 **Effects of political affiliation and identity fusion**

236 Identity fusion was assessed in relation to the political party participants were
237 affiliated with (“Vox” for far-right voters, “Partido Popular” for center-right voters, and
238 “Republican party” for US conservatives). In addition, republicans in the US were also asked
239 to rate their identity fusion with Trump (following Kunst et al.³⁵). Unlike fusion with the
240 Republican party, fusion with Trump has been found to predict willingness to engage in
241 political violence³⁵. Identity fusion was more frequent among the far-right in Spain (15%, 60
242 out of 408) and in relationship to Trump in the US (15%, 122 out of 797) compared to the
243 Spanish center-right (6%, 35 out of 404) or the Republican party in the US (6%, 51 out of
244 797, were fused with the Republican party but not Trump and 14% were fused with both).
245 Analyses of political affiliation and identity fusion effects were pre-registered as exploratory.
246 Similarly, to Spanish far-right voters, conservatives fused with Trump had more sacred
247 values and were more likely to share misinformation that included sacred values, compared
248 to center-right voters and Republicans not fused with Trump, respectively (see Table 2b, 2f
249 and Fig. 1C and 1D). Identity fusion was associated with a greater likelihood of sharing
250 misinformation across groups (see Table 2b-2f).

251 Participants in the neuroimaging study were at the far-right end of the political
252 spectrum (8.27 out of 10 points in a liberal to conservative scale, see demographics in Table
253 1), similar to US survey respondents fused with Trump (8.50 out of 10). However, only 5 of
254 them were completely fused with the far-right party, a similar proportion than that found
255 among far-right voters in the Spanish survey (14%). The reduced number of fused
256 participants in the fMRI study precluded any comparisons between fused and non-fused
257 individuals.

258 **Limited efficacy of fact checking**

259 In our first experiment with Spanish conservatives, we randomly assigned people into
260 three conditions, each of which was fact-checked with a different statement: (1) a regular
261 Twitter fact-check (“*This claim about...is disputed.*“), (2) an accuracy nudge (“*To the best of
262 your knowledge, is the above statement accurate?*”, based on Pennycook et al.¹¹), and (3) a
263 media literacy nudge (“*What techniques are used in this Tweet to attract your attention?*”,
264 based on content published by the Center of Media Literacy, www.medialit.org). All
265 participants first completed a baseline block which did not include any fact-checks, followed
266 by an experimental block that contained social media posts with and without fact-checks.
267 Contrary to prior work, fact-checks did not have any overall effect within the experimental
268 block across the three groups ($M_{diff} = -0.05$, 95% CI [-0.11, 0.02], $t(1622) = -1.73$, $p = .20$,
269 Cohen’s $d = -0.04$, 95% CI [-0.13, 0.06]) and the type of fact-check did not influence
270 likelihood of sharing misinformation during fact-checked trials compared to control trials
271 (Literacy nudge: $p = .88$, Cohen’s $d = -0.02$, 95% CI [-0.19, 0.15]; Accuracy nudge: $p = .89$,
272 Cohen’s $d = -0.02$, 95% CI [-0.18, 0.15]; Twitter fact-check: $p = .10$, Cohen’s $d = -0.07$, 95%
273 CI [-0.24, 0.10], see Fig. 1f). Popular interventions aimed to reduce misinformation sharing
274 by targeting accuracy and media literacy were thus notably ineffective among conservatives
275 and far-right voters.

276 In light of these null effects, we only used the Twitter fact-check in the following US
277 survey and fMRI study. Because Spanish respondents had increased their likelihood of
278 sharing during fact-checked trials compared to the baseline block ($M_{diff} = 0.16$, 95% CI [0.10,
279 0.22], $t(1622) = 5.87$, $p < .001$, see Fig. 1e) we decided to add a control group with no fact-
280 checks in the US survey. This allowed us to test whether this increase was a temporal effect

281 or a backfire effect of fact-checking (see Nyhan and Reifler³⁶). Comparing the two groups
282 confirmed the temporal effect: while the control group was more likely to share social media
283 posts in the second experimental block compared to the baseline block ($M_{diff} = 0.14$, 95% CI
284 [0.06, 0.23], $t(1173) = 3.33$, $p < .001$, Cohen's $d = 0.10$, 95% CI [-0.04, 0.24]), the
285 experimental group did not increase likelihood of sharing in the experimental block
286 compared to baseline ($M_{diff} = 0.01$, 95% CI [-0.06, 0.09], $t(1173) = 0.37$, $p = .71$, Cohen's d
287 $= 0.01$, , 95% CI [-0.11, 0.13], see Fig. 1f). Thus, fact-checks had a small effect keeping
288 participants from increasing sharing in the experimental block.

289 Contrary to predictions, type of value (sacred vs. non-sacred) did not moderate the
290 effect of fact-checks in neither experiment (Spanish sample: sacred values $p = .20$, Cohen's
291 $d = -0.04$, 95% CI [-0.14, 0.05] and non-sacred values $p = 0.76$, Cohen's $d = -0.02$, 95% CI
292 [-0.11, 0.08]; US sample: sacred values $p = .03$, Cohen's $d = -0.08$, 95% CI [-0.22, 0.06] and
293 non-sacred values $p = 0.03$, Cohen's $d = -0.09$, 95% CI [-0.23, 0.05]).

294 An exploratory analysis revealed that identity fusion with Trump in the US sample
295 moderated the effect of fact-checks in the fact-checked group. Conservatives who were not
296 fused with Trump decreased sharing during fact-checked compared to control trials ($M_{diff} =$
297 -0.19 , 95% CI [-0.29, 0.09], $t(376) = -3.74$, $p < .001$, Cohen's $d = -0.13$, 95% CI [-0.29,
298 0.02]), but conservatives fused with Trump did not ($M_{diff} = -0.02$, 95% CI [-0.24, 0.19], $t(376)$
299 $= -0.21$, $p = .84$, Cohen's $d = -0.01$, 95% CI [-0.35, 0.32], see Fig. 1f). Hence, while Twitter
300 fact-checks were somewhat effective in reducing misinformation sharing among non-fused
301 republicans, they were ineffective among highly identified Trump supporters.

302 To explore possible psychological processes underlying the reduced effect of fact
303 checks in participants fused with Trump, we ran an exploratory analysis assessing how these
304 participants responded to fact checks in social media posts with different levels of perceived
305 accuracy. To maximize differences between social media posts associated with high and low
306 perceived accuracy, we selected only the top 10 and the bottom 10 social media posts (out of
307 32 displayed posts) which were associated, respectively, with the highest ($M = 3.96$, $SD =$
308 0.19) and lowest ($M = 2.71$, $SD = 0.29$) perceived accuracy ratings obtained in the US pilot
309 ($N = 80$). We then ran a model comparing the effects of fact-checking on the likelihood of
310 sharing high (vs. low) perceived accuracy social media posts in participants fused with

311 Trump (vs. non-fused). We did not find any evidence of a three-way interaction ($p = .546$)
312 but given the relatively low statistical power we decided to analyze simple effects.
313 Specifically, we found that fact-checking was effective only for low perceived accuracy posts
314 in participants who were not fused with Trump ($M_{\text{diff}} = -0.30$, 95% CI [-0.47, -0.12], $t(905)$
315 $= -3.35$, $p < .001$, Cohen's $d = -0.19$, 95% CI [-0.36, -0.02], see Fig. 1i). Conversely, fact-
316 checks were ineffective for low perceived accuracy posts in participants fused with Trump
317 ($M_{\text{diff}} = -0.20$, 95% CI [-0.58, 0.18], $t(909) = -1.02$, $p = .307$, Cohen's $d = -0.09$, 95% CI [-
318 0.46, 0.29]) and for high perceived accuracy posts across groups (Fused with Trump: $M_{\text{diff}} =$
319 0.17 , 95% CI [-0.23, 0.57], $t(918) = 0.83$, $p = .405$, Cohen's $d = 0.09$, 95% CI [-0.29, 0.48];
320 non-fused with Trump: $M_{\text{diff}} = -0.12$, 95% CI [-0.30, 0.06], $t(909) = -1.29$, $p = .196$, Cohen's
321 $d = -0.05$, 95% CI [-0.23, 0.12], see Fig. 1h). Thus, our exploratory analysis tentatively
322 suggests that fact-checks are most effective for low perceived accuracy posts, though not for
323 people fused with Trump.

324 In the fMRI study, participants saw the Twitter fact-check in half of the trials. Though
325 the effect size of the Twitter fact-check was similar to that found in the US and Spanish
326 surveys, the effect was not statistically significant with a sample of 35 participants ($M_{\text{diff}} = -$
327 0.06 , 95% CI [-0.16, 0.04], $t(34) = -1.28$, $p = .21$, Cohen's $d = -0.07$, 95% CI [-0.54, 0.41]).
328 At a neural level, we expected interventions aimed to reduce misinformation sharing to
329 increase activity in brain areas associated with cognitive control or with prediction error,
330 which have been associated with subsequent behavioral adjustment. However, the interaction
331 between sacred values and fact-checking revealed a network comprising the right
332 dorsomedial prefrontal cortex, the right precuneus, the left hippocampus and the bilateral
333 cerebellum cortex; thresholded at $T = 3.36$, $k = 298$, $p < .001$ FWE c . At baseline, these areas
334 were more active during the sacred value condition, but they became similarly active for
335 sacred and non-sacred values when they were fact-checked (see Fig. 2b and Table 3b). 91%
336 of this brain activity pattern overlapped with the default mode network, while the rest of the
337 activity mostly overlapped with the visual network (8%). A meta-analysis of locations using
338 *Neurosynth*³⁴ revealed associations between the obtained cluster peaks and terms related to
339 theory of mind and autobiographical recall (see Supplementary results). Thus, rather than
340 eliciting activation in brain regions associated with cognitive control and behavioral

341 adjustment, fact-checks resulted in increased activity in brain areas associated with theory of
342 mind.

343 Neural activity in similar areas was higher in participants who started getting fact-
344 checked in the second half of the fMRI session compared to those who were fact-checked
345 from the beginning (see Supplementary results, Fig. 2c and Table 3c), suggesting a decreased
346 response to fact-checking in participants who were habituated to fact-checks.

347 **Cognitive reflection and other variables of interest**

348 Analytical thinking style, as measured by correct responses in the cognitive reflection
349 test, has been previously found to predict reduced sharing of fake news¹⁰. However, we found
350 that, while analytical thinking style was associated with reduced sharing of misinformation
351 related to non-sacred values ($M_{diff} = -0.29$, 95% CI [-0.57, -0.003], $t(982) = -1.98$, $p = .047$,
352 Cohen's $d = -0.16$, 95% CI [-0.43, 0.09]), it did not affect sharing of misinformation related
353 to sacred values ($M_{diff} = -0.04$, 95% CI [-0.32, 0.24], $t(982) = -0.28$, $p = .78$, Cohen's $d =$
354 0.08 , 95% CI [-0.18, 0.34], see Fig. 1e). As such, the cognitive reflection test was not
355 administered in the US survey. Strategies based on analytical thinking may not be very
356 effective at reducing sharing of fake news that directly targets sacred issues such as
357 immigration.

358 With regards to other variables of interest, both media literacy and humility, but not
359 scientific curiosity, were associated with increased sharing of posts with sacred values
360 compared to non-sacred values (see Supplementary results). The presence of moral-
361 emotional language increased sharing in the Spanish sample, while in the US sample it only
362 increased sharing among Republicans fused with Trump (see Supplementary results).

363

364 **Discussion**

365

366 We investigated the role of sacred values and identity fusion in likelihood of sharing
367 political misinformation in conservatives and far-right supporters in Spain and the US.
368 Across three experiments, we found that appealing to sacred values in political messages
369 increases likelihood of sharing misinformation on social media. The effect of sacred values
370 was particularly strong in people who have a fused sense of identity with the far-right in
371 Spain and with Donald Trump in the US. Moreover, we found that a variety of fact-checks

372 employed by social media companies and supported by widely cited papers (e.g., Pennycook
373 et al.¹¹) were largely ineffective – especially in individuals fused with Trump. Similarly,
374 whereas analytical thinking was associated with lower sharing of misinformation about non-
375 sacred values (e.g., infrastructure), it was irrelevant for misinformation around sacred issues
376 (e.g., immigration). In terms of brain activity, we observed a very strong increase in regions
377 that highly overlapped with *theory of mind* networks in response to messages that included
378 sacred values (vs. non-sacred values).

379 Our findings are consistent with a significant role of partisan identity in
380 misinformation sharing³⁷, which stems from the idea that individuals are motivated to
381 identify and conform with groups that provide them protection, status, and resources³⁸. As
382 they identify and bond with the group, individuals engage in identity motivated cognition,
383 processing information in a way that allows them to affirm their identity^{18,39,40}. Thus,
384 partisans appear motivated to believe and share identity-relevant information, with little
385 regard for accuracy¹⁸. Accordingly, we found that, while analytical thinking somewhat
386 reduced sharing of misinformation about identity-irrelevant topics (i.e., non-sacred values
387 such as infrastructure), it made no difference for identity-relevant topics that tackled highly
388 cherished conservative values, like immigration, nationalism, and traditional family values.
389 These results suggest that individuals have a strong drive to share identity-relevant
390 information regardless of how analytical they may be. Several other studies find that partisan
391 motives override accuracy concerns when it comes to sharing news online. For instance, in a
392 study combining survey-based questionnaires and Twitter profile analysis party
393 identification was the strongest predictor of fake news sharing, especially among
394 Republicans, while analytical thinking and media literacy were not significant¹². Party
395 identification was also the strongest predictor of believing fake news¹⁶. Among partisans on
396 the right wing, traditional attempts to fact-check them or increase analytic thinking appear to
397 be less effective than among other populations.

398 Overall, sharing of misinformation was consistently higher for sacred values across
399 samples, even after controlling for attitude strength, familiarity, and salience. This result
400 suggests that malevolent agents trying to spread misinformation benefit from framing false
401 information in terms of sacred values. Consequently, social media users should be
402 particularly careful in discerning the truthfulness of new information that is presented in this

403 format. The identity-defining nature of sacred values helps explain their remarkable appeal
404 when it comes to misinformation sharing. It is important to note that not every value that is
405 *identity-relevant* is also *identity-defining* (e.g., sacred), and new sacred values can emerge
406 depending on how leaders frame relevant issues as they become salient. For instance, at the
407 time the US was contemplating sanctions against the Iranian nuclear program in 2009, the
408 Iranian government framed the nuclear program as an “inalienable right” and many Iranians
409 started perceiving it as a sacred issue⁴¹. Surely, sacred values may be intertwined with elite
410 interests, group interests and ideological concerns. However, disentangling how these
411 constructs are interrelated falls outside of the scope of the present work. Instead, our main
412 goal was to evaluate how non-negotiable identity-defining values are optimally suited to
413 function as vehicles to deliver misinformation.

414 Critically, the effect of sacred values on misinformation sharing was even more
415 prominent for individuals who were fused with extreme political options such as the far-right
416 in Spain and Trump in the US. The fact that highly committed individuals were more willing
417 to share misinformation supports the notion that identity motives more readily override
418 accuracy concerns in extreme partisans, which is unsurprising when considering previous
419 literature on extremism. For instance, previous work found that Moroccans who deemed
420 sharia law a sacred value and were fused with a kin-like group supported militant jihad and
421 were willing to engage in costly sacrifices for the implementation of sharia⁴². In the US
422 context, fusion with Donald Trump predicted willingness to violently prosecute Muslims,
423 immigrants and personally protect the US border³⁵. If partisan incentives can be high enough
424 for extremists to be willing to risk their lives, it is unsurprising that they can also outweigh
425 accuracy concerns.

426 At a brain level, we found a strong neural response to messages involving
427 conservative sacred values (vs. non-sacred values) in far-right supporters. This activity
428 highly overlapped with brain regions typically associated with social cognition and theory of
429 mind, such as the dorsomedial prefrontal cortex, the middle temporal gyrus and the
430 precuneus^{43,44}. These neural findings provide additional support to the notion that identity
431 motives, which are linked to social-cognition processes, are critical in processing the type of
432 misinformation that is more likely to be shared (i.e., misinformation that tackles sacred
433 values). A similar neural activation pattern increased for sacred and non-sacred values that

434 were being fact-checked, especially for participants who were unhabituated to fact-checks.
435 These results suggest that socio-cognitive processes are not only at the root of misinformation
436 sharing, they are also engaged in processing fact-checks; that is, when participants come
437 across fact-checks, they seem to mobilize theory of mind networks (for instance, for assessing
438 intentions) rather than brain regions that support analytical thinking such the frontoparietal
439 network. While the present design does not allow us to infer what exact socio-cognitive
440 processes our participants were engaged in, there are several plausible options that would be
441 interesting to test in future studies. One option is that they are evaluating the appropriateness
442 of sharing the post by gauging on how others would judge it (especially the ingroup). This
443 process would be particularly relevant for social signaling. Alternatively, because identity
444 and self-referential processes neurally overlap with social cognition, especially in the medial
445 prefrontal cortex⁴⁵, participants could be strengthening their partisan identities by
446 incorporating new (false) information about their ingroup's worldview.

447 The immediate consequence of misinformation sharing being a fundamentally social
448 process is that popular interventions such as fact-checking and accuracy nudges are less likely
449 to be effective among highly fused individuals. Prior work has shown that respondents are
450 more sensitive to falsehood in news headlines after being asked to rate their accuracy¹¹.
451 However, we find close to no effect of fact-checking in our conservative samples, especially
452 for fact-checks based on accuracy and media literacy. Fact-checks were particularly
453 ineffective in participants fused with Trump, suggesting that, relative to accuracy incentives,
454 partisan incentives become even more relevant for hyper-partisans.

455 One possible explanation for these null effects is that the stimuli we designed were
456 too plausible. Previous studies have found that the disposition for analytical thinking is
457 unlikely to help reduce belief in misinformation that is plausible¹⁰. Thus, it could be that fact-
458 checks and nudges are helpful specifically when the presented misinformation is clearly
459 implausible. We tested this idea in an exploratory analysis comparing the effect of fact-
460 checks on high and low perceived accuracy items in the US sample. Because asking
461 participants to make accuracy judgements could influence subsequent likelihood of sharing¹¹,
462 we obtained accuracy judgements for each social media post in a separate pilot study using
463 a similar sample. Our results are consistent with the notion that fact-checks are most useful
464 in reducing sharing of misinformation that is perceived to be less accurate on average.

465 However, this effect was absent in participants strongly fused with Trump, who still did not
466 decrease sharing of fact-checked posts containing claims deemed to be implausible by other
467 Republicans. Thus, there appears to be an important role for social identity in the efficacy of
468 these interventions (see also Rathje et al.¹⁴). Future research should test this hypothesis with
469 larger samples of individuals fused with Trump to confirm these preliminary results.

470 The limited effectiveness of fact-checks on conservatives and, particularly,
471 Republicans fused with Trump is in line with other studies showing that conservatives and
472 those at the far-right of the political spectrum are less sensitive to false claims on social
473 media²⁹, are more exposed to them⁴ and six to seven times more likely to share them than
474 moderates and liberals^{4,15}. However, because conservatives and liberals perform equally well
475 on the cognitive reflection test³⁹, lack of analytical reasoning cannot explain the differences
476 in misinformation sharing and susceptibility to fact-checks across the political spectrum. Our
477 data suggest that extreme political movements have a greater capacity to mobilize partisans
478 and fuse them with the movement. Identity fusion rates for the far-right in Spain and Trump
479 in the US are notably higher (15% of the sample) than for the center-right in Spain (9%) and
480 the Republican party (6% exclusively fused with the party but not Trump). Because identity
481 fusion predicts misinformation sharing and susceptibility to fact-checks, movements'
482 capabilities to elicit identity fusion among supporters could contribute to political
483 asymmetries in these two dimensions.

484 In sum, our work emphasizes the role of political devotion in fueling misinformation
485 sharing and suggests that socio-cognitive processes may be critical for the spread of
486 misinformation among partisans. While accuracy nudges may work for non-partisans (albeit
487 with very small effect sizes), these populations are probably not at the root of the
488 misinformation crisis, because they are already less prone to sharing misinformation. The
489 speed and ease with which fake news spread as compared to true news⁵ may be a result of
490 how this misinformation is designed. Fake news creators likely benefit from tackling sacred
491 issues, thus mobilizing large crowds of who hold a sincere belief that they are fighting for a
492 legitimate cause. For instance, InfoWars, one of the most popular websites known to spread
493 fake-news, mobilizes their followers by calling them into *battling globalism*, an idea that
494 may be powerful for many. By appealing to a cause greater than the self, identity-relevant
495 misinformation becomes very tempting, even for highly analytical partisans. Spreading

496 misinformation might be thus a warfare tool employed by interest groups to amplify their
497 messages and gain power through devoted partisans.

498 The present work is subject to several limitations. Because of previously reported
499 political asymmetries in information sharing^{4,15,29}, we decided to focus on conservatives.
500 However, our data indicate that extreme political options elicit higher levels of identity fusion
501 among supporters, suggesting that far-left supporters could possibly show similar patterns of
502 misinformation sharing relative to sacred values. Moreover, although the fMRI sample was
503 politically extreme (their average political orientation was 8.27/10, similarly to participants
504 fused with Trump), it did not include enough fused participants to be able to test the effect
505 of identity fusion in the scanner. Similarly, we did not include a control group in the fMRI
506 study, so that the presented results are not necessarily specific to far-right supporters. In fact,
507 we believe we could find similar neural results in other groups exposed to misinformation
508 relevant to their own sacred values. For instance, a recent nationally representative sample
509 of Americans found that people on the extreme left score higher in dogmatism (albeit not as
510 far as those on the far right) and they may therefore be resistant to information that challenges
511 their beliefs (see Harris et al.⁴⁶). Finally, our study was a controlled experiment with
512 artificially designed social media posts. Future work will benefit from field studies looking
513 at the relationship between identity fusion, sacred values, and fact-checks in real social media
514 ecosystems.

515 **Conclusion**

516 Overall, our findings suggest that stopping the spread of misinformation may be a
517 much greater challenge than previously thought. Our data suggests that believing and sharing
518 fake news is, at least in part, a socio-cognitive process serving partisan identity goals¹⁸.
519 Misinformation is specifically designed to tackle sacred beliefs that forge people's partisan
520 identity and to which they hold on with great conviction. Critically, individuals' willingness
521 to spread this type of content appears resistant to analytical thinking and nudges aimed to
522 improve accuracy, especially among hyper-partisans, who are most likely to share it in the
523 first place. Strategies aimed to reduce misinformation should thus address the devotional
524 aspect of misinformation sharing.

525

526 **Materials and Methods**

527

528 This research was approved by the Ethics Committee on Human and Animal Experimentation
529 at the Universitat Autònoma de Barcelona according to the Declaration of Helsinki guidelines (Ref.
530 5385 and 5388). The pre-registration of this study can be found following [this link](#).

531 **Experimental Design**

532 We launched surveys in Spain (Study 1, N = 812) and the US (Study 2, N = 797), and
533 conducted a neuroimaging study (N = 36) asking participants to rate the likelihood of sharing a series
534 of social media posts composed by different political leaders and public figures who are part of, or
535 openly support, the party they voted for in the last elections. Because the Spanish congress has
536 multiple parties, we chose to focus on voters of two conservative political parties in Spain (center-
537 right and far-right) in Study 1, voters of Republican Donald J. Trump in the US in Study 2, and voters
538 of the far-right in Spain for the neuroimaging study.

539 In the three studies, half of the posts, designed to look like tweets, included conservative
540 sacred values (immigration, nationalism, and women and family values) and the other half non-sacred
541 values (roads and infrastructure, foreign affairs, and waste management and materials). Because
542 previous studies suggest an effect of moral-emotional language in online content sharing (43), all
543 items were formulated twice, once using moral-emotional language and once using neutral language
544 (see further details in *Materials*). To confirm that these issues are sacred in Spain and the US, we
545 assessed how many of our respondents deemed each value sacred in the three studies. To test the
546 effect of online fact-checking on social media posts, we also introduced fact-checks in some of the
547 social media posts. In Study 1 and 2, participants were first asked to rate four Tweets without fact-
548 checks (baseline block) and then rate twelve (Study 1) or eight more posts (Study 2) in a mixed block
549 with half of the Tweets marked with a fact-check (experimental block).

550 In Study 1 (N = 812) participants were split in three groups and each group was exposed to a
551 different fact-check in the experimental block: a classic Twitter fact-check (N = 270, “*This claim*
552 *about...is disputed.*“), an accuracy-based fact-check (N = 277, “*To the best of your knowledge, is the*
553 *above statement accurate?*”, based on Pennycook et al.¹¹) and a media literacy-based fact-check (N
554 = 265, “*What techniques are used in this Tweet to attract your attention?*”, based on content published
555 by the Center of Media Literacy, www.medialit.org). In Study 2 (N = 797) half of the sample was
556 exposed to the classic Twitter fact-check and the other half did not see any fact-checks in any of the
557 posts.

558 In the neuroimaging study (N = 36), an event-related fMRI paradigm was designed using
559 MATLAB 2020a (The MathWorks, Inc., Natick, Massachusetts, USA) with Psychtoolbox
560 extensions⁴⁷. Participants completed two 6-minute runs without any fact-checks (baseline block)
561 followed by another two 6-minute runs with fact-checks in each post (experimental block). The order
562 of these two blocks was reversed in half of the sample (N = 18). Stimuli were presented randomly
563 with jittered inter-trial intervals that varied randomly from 0.5 s up to 3.70 s. Each run included 40 7-
564 second trials (20 trials for the sacred value condition and 20 trials for the non-sacred value condition)
565 and in each trial, participants were given 4.5 seconds to read a social media post, and 2.5 seconds to
566 respond how likely they would be to share it using a 6-point Likert scale adapted for fMRI using
567 left/right buttons (see Fig. S2). The orientation of the scale was reversed in half of the trials and the
568 cursor starting point was randomized. Half of the presented social media posts included sacred values
569 and the other half non-sacred values. The total duration of the paradigm was 24 minutes. Participants
570 had the chance to practice the task thoroughly on a laptop before going into the scanner. No posts
571 with fact-checks or sacred values were shown during the training phase.

572 **Participants.** Demographic data of each study sample is presented in Table 1. For Study 1,
573 408 Spanish far-right voters and 404 center-right voters were recruited by means of an online panel.
574 Inclusion criteria included being 18 years old or older and having the intention to vote for either a far-
575 right party (“Vox”) or a center-right party (“Partido popular”) in the next presidential election. Study
576 2 included a sample of 797 participants, recruited via an online panel, who had reported voting for
577 Republican D. J. Trump in the two previous US presidential elections (2016 and 2020).

578 A power analysis conducted with the R package ‘simr’⁴⁸ revealed that, based on 1000
579 simulations, recruiting 800 participants would enable the detection of small effects (e.g. a slope of
580 0.1) with an alpha = 0.05 for the interaction term between (a) within-subjects value type (sacred vs.
581 non-sacred) and fact-checking (fact-checked vs. control) with a statistical power of 99.6% (95% CI
582 [98.98, 99.89]) in the Spanish and US survey with four observations per participant, (b) within-
583 subjects fact-checking (baseline, control and fact-checked trials) and fact-checking group (Twitter vs.
584 Accuracy vs. Literacy fact-check) with a statistical power of 79.1% (95% CI [76.45, 81.58]) in the
585 Spanish survey with three observations per participant and between (c) within-subjects block
586 (baseline vs. experimental block) and fact-checking group (fact-checked group vs. control group)
587 with a statistical power of 82.9% (95% CI [80.42, 85.18]) in the US survey with two observations per
588 participant.

589 For the neuroimaging study, a power analysis based on 1000 simulations and an alpha = 0.05
590 showed that with 40 participants and 80 observations per participant (40 trials per run in 2 runs), we

591 would be able to detect small effects (e.g., a slope of 0.1) for the interaction between within-subjects
592 value type (sacred vs. non-sacred) and within-subjects fact-checking (control vs. fact-checked trials)
593 with a statistical power of 99.80% (95% CI [99.28, 99.98]).

594 We recruited thirty-six far-right supporters for the neuroimaging study using social media.
595 Initially, we contacted Twitter followers of far-right accounts based in Barcelona via direct message
596 and recruited them to participate in the study. Ultimately, most of our participants were recruited
597 through other participants that shared our study ad in far-right youth WhatsApp groups that are used
598 to organize political actions locally and usually include a few hundred members. Participants were
599 selected based on their response to the question “Which political party best represents your values
600 and beliefs at this moment?”. Only participants who responded with a far-right party (“Vox”) and
601 were 18 years old or older were recruited for the study. Candidates with neurological and psychiatric
602 disorders, taking psychiatric medication, claustrophobic, or with metal parts in their body
603 incompatible with an MR scanner were excluded.

604 Informed written consent was obtained from all participants at the start of each study.
605 Participants were debriefed at the end of each study and were given the chance to discuss and ask
606 questions about the study.

607 **Materials.** *Social media posts.* The Tweets were specifically designed to convey 8 false
608 statements about sacred political issues (immigration, nationalism, and women and family values)
609 and 8 false statements about non-sacred political issues (roads and infrastructure, foreign affairs,
610 waste, and materials). All posts were designed by researchers to contain false information and most
611 posts contained some sort of critique of the current liberal government both in Spain and the US (see
612 complete list of items employed in the US study in Table S2). Some examples included:

613 *3 out of 4 illegal immigrants who enter our country end up in criminal gangs, endangering*
614 *our American way of life.* [sacred value; moral-emotional wording]

615 *That the government has slashed the funding for American embassies in the Asia-Pacific*
616 *region is an attack against the nation.* [non-sacred value; moral-emotional wording]

617 Social media posts with sacred (vs. non-sacred) values were combined with similar number
618 of likes ($M_{diff} = 2.50$, 95% CI [-12.6, 17.6], $t(14) = 0.35$, $p = .73$), number of retweets ($M_{diff} = -21.5$,
619 95% CI [-120, 76.6], $t(14) = -0.47$, $p = .64$) and gender of the leaders tweeting the posts (the same 3
620 female and 4 male leaders tweeted the posts in each condition). The selection of leaders was different
621 for each sample so that each voters’ group had their own set of 7 ingroup leaders (center-right leaders,
622 far-right leaders, and pro-Trump leaders). Nonetheless, all sets of leaders tweeted the exact same
623 content (center-right and far-right leaders) or American adapted content (pro-Trump leaders).

624 Items in the sacred and non-sacred value conditions were also matched for character length
625 in the Spanish version for Study 1 ($M_{diff} = -1.69$, 95% CI [-3.78, 0.40], $t(30) = -1.65$, $p = .11$), in the
626 English version for Study 2 ($M_{diff} = 2.44$, 95% CI [-7.33, 12.2], $t(30) = 0.51$, $p = .61$) and in a shorter
627 Spanish version adapted for the neuroimaging study ($M_{diff} = 0.75$, 95% CI [-1.31, 2.81], $t(30) = 0.74$,
628 $p = .46$).

629 In addition, type of language was counterbalanced across both sacred and non-sacred value
630 conditions; that is, all 16 items were formulated twice, once using moral-emotional language and
631 once using neutral language. Two pilot studies ($N = 45$ each) were conducted to ensure that the two
632 language formulations were perceived as different in terms of moral and emotional language in both
633 the Spanish and the English versions. Results of these pilot studies can be found in supplementary
634 materials (see Fig. S1). The final set of posts included 32 items.

635 To make sure that the sacred values we tackled were indeed sacred by our study samples, the
636 value sacredness of each issue was measured in the three studies, as indicated in the *Measures* section.
637 As expected, the three proposed sacred values were sacred by a large segment of each of the three
638 samples as compared to the non-sacred values (see Table S1).

639 Finally, to control for the effect of potential confounds associated with sacred values in each
640 of the social media posts, we ran a third pilot study on a sample of 80 US conservatives who were
641 pre-screened using the same criteria as the US study participants (having voted for Trump in the two
642 previous US presidential elections in 2016 and 2020). For each of the 32 social media posts employed
643 in the US study, we obtained ratings of perceived accuracy (“To the best of your knowledge, is the
644 claim in the above social media post accurate?” from 1 = “Very inaccurate” to 5 = “Very accurate”),
645 certainty (“I feel very certain about my feelings toward this issue”), extremity (“I feel strongly about
646 this issue”), importance (“This issue is personally important to me”), familiarity (“I am quite familiar
647 with this issue”), and salience (“I spend a lot of time thinking about this issue”) on a 5-point Likert
648 scale from “Strongly disagree” to “Strongly agree”.

649 **Measures.** *Likelihood of sharing.* Disposition to share each of the social media posts was
650 assessed using a probability scale “If you were to see the above post on social media, how likely
651 would you be to share it?”¹¹. Responses could range from 1 “Extremely unlikely” to 6 “Extremely
652 likely”.

653 *Value sacredness.* Value sacredness was assessed by means of economic trade-off scenarios.
654 We asked participants if they would be “willing to give up [value] if that involved a great benefit to
655 [American/Spanish] families such as a better economy, more jobs, better schools and hospitals, and
656 in general, a better quality of life for all [American/Spanish] families including yours” (following

657 Hamid et al.⁴⁹. For instance, participants were asked whether they would allow the free entry of
658 immigrants in their country if that involved a great benefit for American families. Possible responses
659 included “Yes”, “No” and “Maybe”. A value was considered sacred if participants refused to give up
660 that value.

661 *Identity fusion.* Identity fusion was assessed using the pictorial measure²², which includes
662 five pairs of circles with different degrees of overlap representing the relationship between the
663 participant (small circle) and the group (big circle). Respondents were asked to convey which pair of
664 circles best represents their relationship with the group. Identity fusion was assessed in relation to the
665 political party participants supported (“Vox” for far-right voters, “Partido Popular” for center-right
666 voters, and “Republican party” for Trump supporters). In addition, Trump supporters were also asked
667 to rate their identity fusion with Trump specifically (following Kunst et al.³⁵).

668 *Scientific curiosity.* Scientific curiosity was assessed by means of three items extracted from
669 the stretching subscale of the Science Curiosity in Learning Environments scale⁵⁰, including “I try to
670 learn as much as I can in new situations”, “I see a challenge as a way to grow and learn”, and “I apply
671 new information to an existing problem to see if that helps”. Participants responded using a five-point
672 Likert scale from 1 “Not at all like me” to 5 “Very much like me”. The three items were averaged
673 into a Scientific curiosity score ($\alpha = .77$) with values ranging from 1.33 and 5 ($M = 4.07$, $SD = 0.67$).

674 *Intellectual humility.* Intellectual humility was evaluated with the 6-item open mindedness
675 subscale of Alfano et al.⁵¹ intellectual humility scale. This subscale includes items such as “I think
676 that paying attention to people who disagree with me is a waste of time.” and “I feel no shame learning
677 from someone who knows more than me.”. The 5-point response scale ranged from 1 “Not at all like
678 me” to 5 “Very much like me”. The six items were averaged into an Intellectual humility score ($\alpha =$
679 $.67$) with values ranging from 2.17 and 5 ($M = 1.12$, $SD = 0.60$).

680 *Media literacy.* Media literacy was measured by means of three items of the Media literacy
681 education scale by Simons and Meeus⁵², including “I know that media represents information in a
682 selective way and know how to interpret media messages” and “I can evaluate media content taking
683 into account various criteria (e.g. accuracy of information, comparison of information, appreciation
684 of aesthetic aspects)”, using a five-point Likert scale from 1 “Not at all like me” to 5 “Very much like
685 me”. The three items were averaged into a Media literacy score ($\alpha = .70$) with values ranging from 1
686 and 5 ($M = 3.74$, $SD = 0.76$).

687 *Cognitive reflection test.* Analytical (vs. intuitive) thinking styles were assessed by means of
688 the cognitive reflection test⁵³ in the Spanish sample. This questionnaire includes three mathematical
689 problems such as “A bat and a ball cost \$1.10 in total. The bat costs \$1.00 more than the ball. How

690 much does the ball cost?”. Correct answers were added into an Analytical thinking score with values
691 ranging between 0 and 3 ($M = 0.67$, $SD = 0.96$).

692 *Neuroimaging data.* Magnetic resonance images were obtained in a Philips 3 T scanner. T1-
693 weighted images were acquired using a FSPGR sequence (TR: 9.9 ms, TE: 4.6 ms, FA: 8, matrix
694 size: 240×240 , 180 slices, and slice thickness: 1.00 mm) and functional volumes were obtained using
695 a EPI-T2* sequence (TR: 1750 ms, TE: 35 ms, FA: 70, matrix size: 76×76 , 46 slices, and slice
696 thickness: 3.1 mm).

697 **Statistical Analysis**

698 The data and code employed in the analyses are available following [this link](#). Due to privacy
699 concerns, the neuroimaging data can be obtained from the authors upon request with an approved
700 IRB (we will send a password protected link to scientists with ethical approval).

701 **Behavioral analyses.** To evaluate the effect of sacred values, political affiliation, identity
702 fusion and fact-checking in the Spanish and US surveys we conducted a series of mixed effects
703 models with random intercepts for participants using REML (afex package in R⁵⁴). To assess
704 differences in the effect of sacred values on likelihood of sharing (a) between far right and center
705 right voters in Spain and (b) between conservatives fused and non-fused with Trump in the US, we
706 added an interaction term for group (a or b) and type of value (sacred vs. non-sacred value). Of note,
707 the identity fusion pictorial measure (22) was analyzed as a dichotomous variable following previous
708 literature, with the maximal score coded as “fused” and the rest as “non-fused”.

709 To test the effect of fact-checks in trials with and without fact-checks in both surveys, we
710 created two mixed effects models with interaction terms for (a) fact-check group (Twitter fact-check,
711 accuracy fact-check and literacy fact-check) and presence of fact-check (baseline trials, fact-checked
712 trials, and control trials) for the Spanish sample, and (b) fact-check group (Twitter fact-check vs. no
713 fact-check/control group) and presence of fact-check (baseline block vs. experimental block) in the
714 US sample. All employed tests were two-sided.

715 **Neuroimaging analyses.** Image processing was conducted using SPM12 (Wellcome Trust,
716 University College London, UK) on MATLAB. Functional images were realigned and co-registered
717 to the structural images, which were then segmented into white matter, gray matter, and cerebrospinal
718 fluid. The forward deformation fields generated during the segmentation of the structural images were
719 used to normalize the functional images. Finally, images were smoothed using an 8-mm FWHM
720 kernel.

721 The first-level general linear model (GLM) included two regressors for stimulus presentation
722 (for sacred vs. non-sacred value trials) with a duration of 4.5 seconds each, and two regressors for the
723 response period (for sacred vs. non-sacred value trials) with a duration of 2.5 seconds each. The
724 response regressors were parametrically modulated by number of keypresses in each trial. Six non-
725 convolved head movement regressors were also included. In addition, first level GLMs included
726 scrubbing regressors that accounted for outlier volumes as required by each participant. Outliers
727 included volumes with a framewise displacement larger than 0.9mm or global BOLD signal changes
728 above 0.5 SD. One participant had to be eliminated due to excessive head movement.

729 Group-level effects of sacred vs. non-sacred values during stimulus presentation time-locked
730 to the start of the onset stimulus were evaluated by means of a t-test on individual level contrasts.
731 Group differences related to the order of the runs (half of the sample started with fact-checks and the
732 other half without fact-checks) were assessed using a two-sample t-test on individual level contrasts.
733 The fact-check by value interaction contrast was computed by assigning counterbalanced weights to
734 fact-checked (fc) vs. control (ctr) posts with sacred values vs. non-sacred values ([1 x ctr sacred values
735 -1 x ctr non-sacred values -1 x fc sacred values 1 x fc non-sacred values]). *Post hoc* analyses included
736 parameter estimate extractions of the cluster of activity in the dorsomedial prefrontal cortex in the
737 fact-check by value interaction contrast and the group by value interaction contrast using Marsbar.
738 Reported results were corrected for multiple comparisons using family-wise error (FWE) correction
739 at cluster level, with a peak-level contrast of $p < .001$.

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872

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882

883 **Author contributions**

884

885 J.V.B., O.V. and C.P. conceived the online and neuroimaging studies. J.V.B. directed the online
886 experiments and managed project communication and O.V. directed the neuroimaging study. C.P.
887 developed the online experiments and the neuroimaging study. W.J.B. and E.A.H. participated in
888 developing the online studies and C.S. participated in data collection for the neuroimaging study. C.P.
889 conducted the data analysis and initiated manuscript preparation. All authors edited and approved the
890 final version of the manuscript.

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892 **Competing interests**

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894 Authors declare that they have no competing interests.

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896 **Data and code availability**

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898 The data and code employed in the analyses are available following [this link](https://osf.io/twr6b/)
899 (<https://osf.io/twr6b/>). Due to privacy concerns, the neuroimaging data can be obtained
900 from the authors upon request with an approved IRB (we will send a password protected
901 link to scientists with ethical approval).

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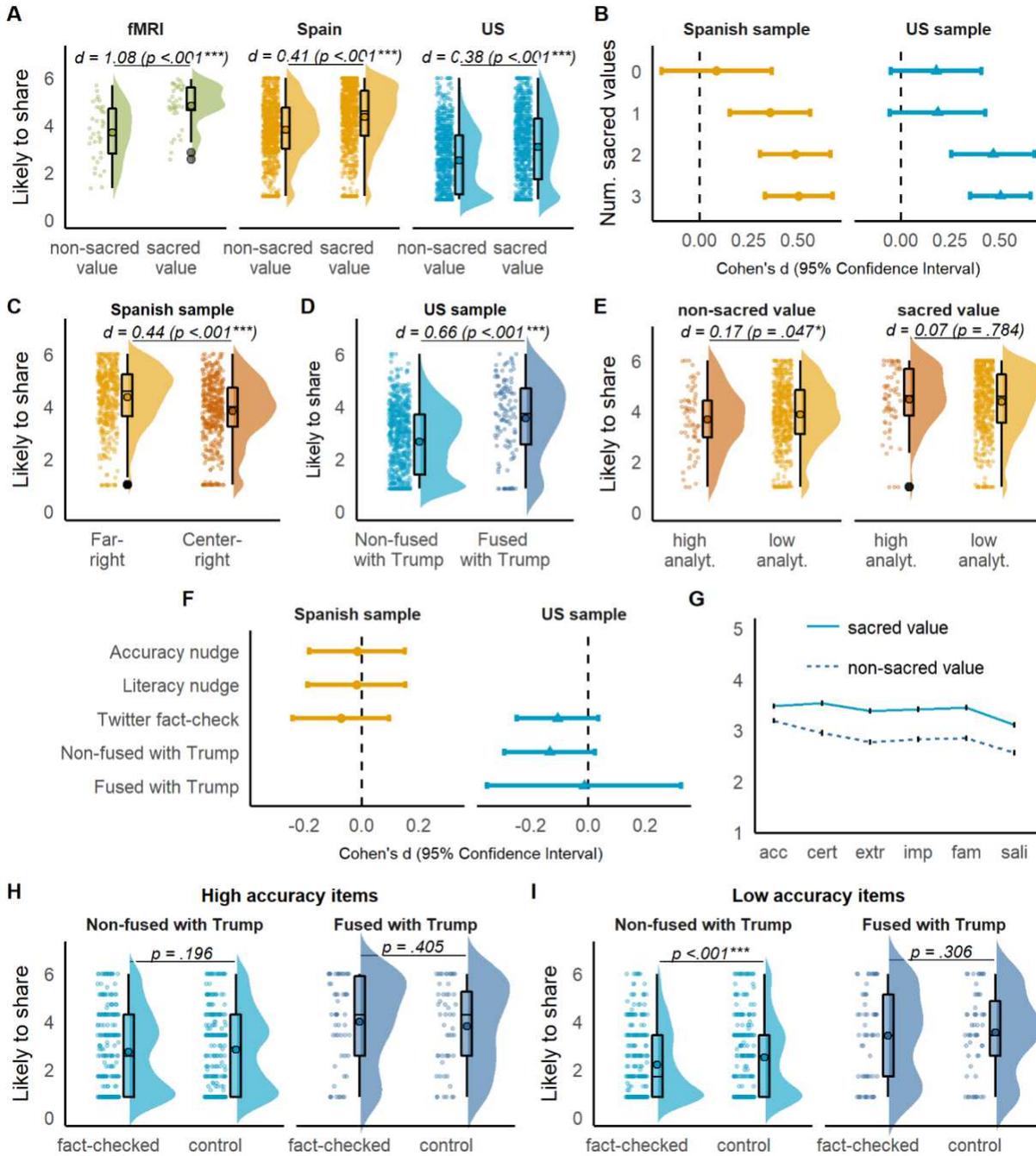
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Figures and Tables



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920 **Fig. 1. Behavioral results:** The effect of sacred vs. non-sacred values on likelihood of sharing was
 921 observed across the three samples (A) and was more prominent for participants with a higher number
 922 of sacred values in both the US and the Spanish sample (B), for conservatives fused vs. non-fused
 923 with Trump in the US (C) and for far-right vs. center-right voters in Spain (D). Analytical thinking

924 was associated with reduced sharing of misinformation about non-sacred values but was unrelated to
925 sharing misinformation about sacred values in the Spanish survey sample (**E**) (the cognitive reflection
926 test was not included in the US survey). Fact-checks were largely ineffective, with no effect of type
927 of fact-check for fact-checked compared to control trials in the Spanish sample (**F left**) and a small
928 effect of the Twitter fact-check in the experimental block (vs. baseline) in fact-checked participants
929 compared to control participants in the US sample (**F right**). Fact-checks were even less effective in
930 conservatives fused with Trump (**F right bottom**). Boxplot bounds represent the interquartile range
931 and the circle within the box represents the means.

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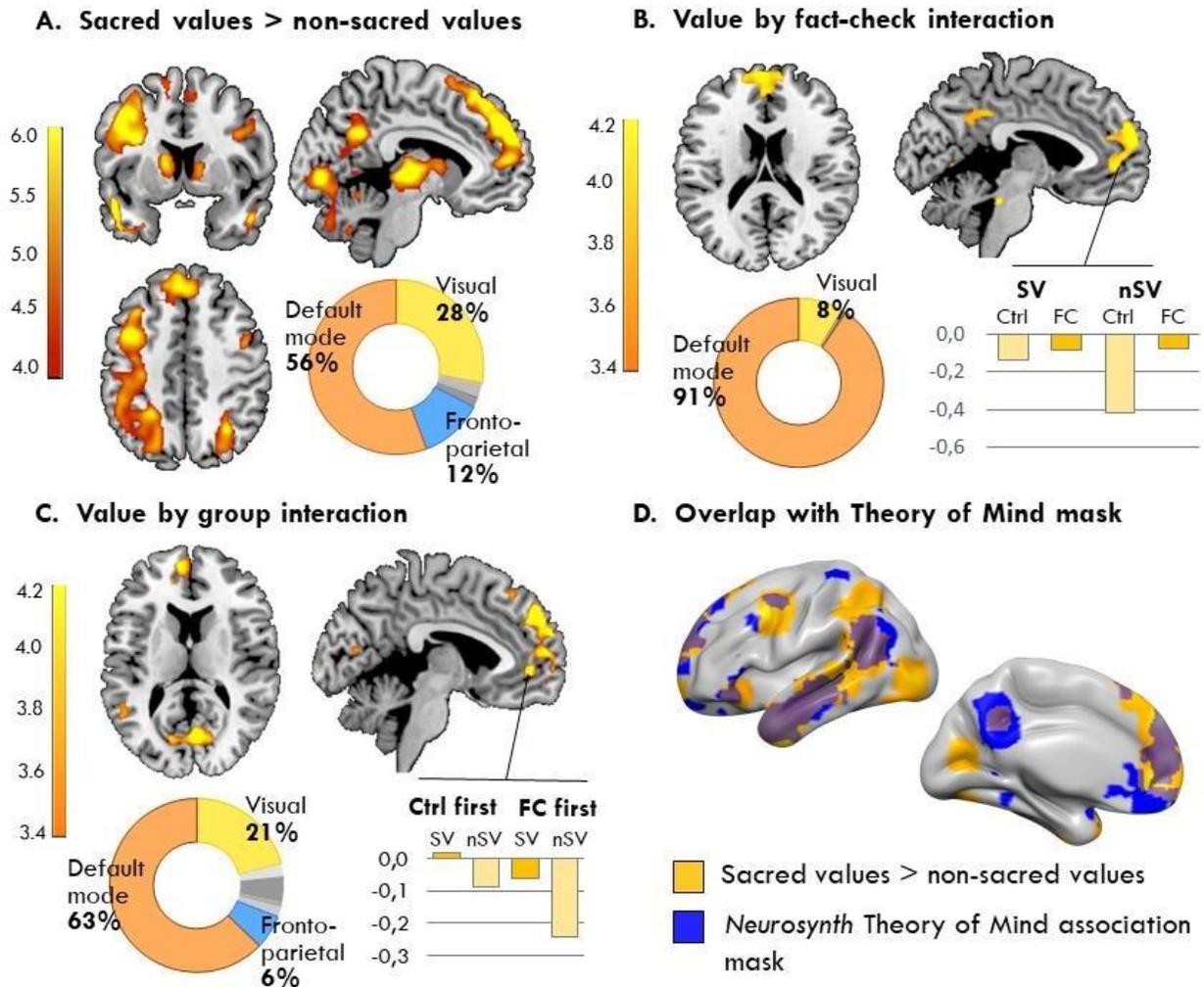
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Fig. 2. Neuroimaging results. Social media posts with sacred values yielded much greater brain activation compared to posts with non-sacred values, especially in default mode regions (A). A similar brain activation pattern was found for fact-checked trials compared to control trials, an increase that was more prominent for non-sacred value trials, which had notably lower activation in control trials than sacred value trials (B). Participants who started with fact-checked trials exhibited lower activation in these brain areas as well, suggesting a reduced neural effect of fact-checks in participants who were familiarized with fact-checks from the beginning (C). Comparison with a *Neurosynth* functional mask suggests that the activity detected in the sacred vs. non-sacred value contrast largely overlaps with functional activations associated with *theory of mind* studies. SV = sacred values, nSV = non-sacred values, Ctrl = control trials, FC = fact-checked trials.

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Table 1. Demographic information of the study samples in Spain and the US.

	Spanish sample (N = 812)		US sample (N = 797)		fMRI sample
	<i>Center- oters</i>	<i>Far-right rs</i>	<i>Republican Not fused Trump</i>	<i>Republican fused with ump</i>	<i>Spanish t voters</i>
<i>N</i>	404	408	675	122	36
<i>% Women</i>	60	48	46	47	33
<i>Age (years)</i>	45.6(15.6)	44.6(13.8)	41.7(13.7)	43.7(13.6)	23.1(4.8)
<i>% Higher education</i>	56	48	65	56	19
<i>% Low household income (ith)</i>	28*	32**	n/a	n/a	24***
<i>Political orientation (1 = /ht)</i>	7.15(1.58)	7.55(1.84)	7.46(1.76)	8.5(1.85)	8.27(1.52)
<i>Means of recruitment</i>	YouGov Spain		Prolific		WhatsApp ups

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*data not available for 63 participants; **data not available for 61 participants; ***data not available for 6 participants.

958 **Table 2. Behavioral results.** Effect of sacred values, political affiliation, and identity fusion on
 959 willingness to share misinformation across the different study samples.

<i>Effect</i>	<i>M</i>	<i>SD</i>	<i>M_{diff}</i>	<i>95% CI</i>	<i>t</i>	<i>p</i>	<i>Cohen's d</i>	<i>95% CI</i>
(a) Sacred values								
Spanish sample SV vs. non-SV			0.54	[0.48, 0.60]	18.80	< .001	0.41	[0.31, 0.51]
US sample SV vs. non-SV			0.56	[0.50, 0.63]	16.03	< .001	0.38	[0.28, 0.48]
fMRI sample SV vs. non-SV			1.14	[0.82, 1.47]	7.09	< .001	1.06	[0.58, 1.60]
(b) Political affiliation (Spain)								
<i>Likelihood of sharing</i>								
Far-right	4.35	1.55						
Center-right	3.81	1.61	0.54	[0.37, 0.71]	6.28	< .001	0.41	[0.31, 0.51]
<i>Likelihood of sharing sacred values</i>								
Interaction: Political affiliation x sacred values			0.24	[0.14, 0.35]	4.14	< .001		
Far-right SV vs. non-SV			0.66	[0.58, 0.74]	16.38	< .001	0.52	[0.38, 0.66]
Center-right SV vs. non-SV			0.42	[0.34, 0.50]	10.46	< .001	0.32	[0.19, 0.46]
<i>Number of sacred values</i>								
Far-right	2.09	0.95						
Center-right	1.62	1.03	0.46	[0.33, 0.60]	6.65	< .001	0.47	[0.33, 0.61]
(c) Fusion with far-right (Spain)								
<i>Likelihood of sharing</i>								
Fused	5.13	1.23						
Non-fused	4.21	1.57	0.92	[0.61, 1.24]	5.76	< .001	0.81	[0.52, 1.09]
<i>Likelihood of sharing sacred values</i>								
Interaction: Identity fusion x sacred values			0.08	[-0.15, 0.32]	0.68	<i>n.s.</i> (.50)		
Fused SV vs. non-SV			0.59	[0.37, 0.81]	5.31	< .001	0.62	[0.25, 0.99]
Non-fused SV vs. non-SV			0.67	[0.58, 0.76]	14.55	< .001	0.53	[0.38, 0.68]
<i>Number of sacred values</i>								
Fused	2.31	0.93						
Non-fused	2.04	0.95	0.27	[0.01, 0.53]	2.03	.043	0.28	[0.01, 0.56]
(d) Fusion with center-right (Spain)								
<i>Likelihood of sharing</i>								
Fused	4.48	1.58						
Non-fused	3.74	1.60	0.74	[0.31, 1.17]	3.41	< .001	0.60	[0.25, 0.95]
<i>Likelihood of sharing sacred values</i>								
Interaction: Identity fusion x sacred values			0.03	[-0.23, 0.29]	0.23	<i>n.s.</i> (.82)		
Fused SV vs. non-SV			0.39	[0.14, 0.65]	3.05	.002	0.34	[-0.14, 0.82]
Non-fused SV vs. non-SV			0.42	[0.35, 0.50]	10.71	< .001	0.32	[0.18, 0.47]
<i>Number of sacred values</i>								
Fused	1.83	1.07						

Non-fused	1.60	1.03	0.23	[-0.13, 0.59]	1.24	<i>n.s.</i> (.22)	0.22	[-0.13, 0.57]
<i>Fusion with Republican party, excluding those fused with Trump (US)</i>								
<i>Likelihood of sharing</i>								
Fused	3.73	1.87						
Non-fused	2.63	1.72	1.25	[0.89, 1.61]	6.77	< .001	0.99	[0.69, 1.28]
<i>Likelihood of sharing sacred values</i>								
Interaction: Identity fusion x sacred values			0.15	[0.06, 0.34]	1.08	<i>n.s.</i> (.28)		
Fused	SV vs. non-SV		0.37	[0.11, 0.64]	2.78	.006	0.30	[-0.10, 0.69]
Non-fused	SV vs. non-SV		0.52	[0.45, 0.60]	13.69	< .001	0.38	[0.27, 0.50]
<i>Number of sacred values</i>								
Fused	1.92	1.13						
Non-fused	1.67	1.20	0.24	[-0.08, 0.57]	1.46	<i>n.s.</i> (0.14)	0.21	[-0.07, 0.50]
(e) <i>Fusion with Trump (US)</i>								
<i>Likelihood of sharing</i>								
Fused	3.56	1.97						
Non-fused	2.64	1.71	0.89	[0.63, 1.15]	6.73	< .001	0.66	[0.47, 0.86]
<i>Likelihood of sharing sacred values</i>								
Interaction: Identity fusion x sacred values			0.32	[0.14, 0.52]	3.43	< .001		
Fused	SV vs. non-SV		0.84	[0.67, 1.02]	9.47	< .001	0.52	[0.27, 0.78]
Non-fused	SV vs. non-SV		0.51	[0.44, 0.59]	13.51	< .001	0.37	[0.26, 0.47]
<i>Number of sacred values</i>								
Fused	2.09	1.19						
Non-fused	1.85	1.14	0.24	[0.02, 0.46]	2.11	.035	0.21	[0.01, 0.40]

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963 **Table 3. Neuroimaging results.** Neural activity resulting from (a) the contrast between sacred and
 964 non-sacred values (*results of this contrast were reported at a more conservative threshold (0.05 FWE
 965 peak-level) due to generalized activation across the brain at lower thresholds), (b) the interaction
 966 contrast between sacred values and fact-checks, and (c) differences associated with sacred vs non-
 967 sacred values in participants who started with control trials and those who started with fact-checked
 968 trials.

Regional label	k	t(33)	MNI coordinates			P-FWEc
			x	y	z	
<i>(a) Sacred > non-sacred values*</i>						
L middle temporal gyrus	714	8.96	-52	-30	0	< .001
R middle occipital gyrus	966	8.67	36	-78	0	< .001
L inferior occipital gyrus	680	8.61	-30	-78	-4	< .001
L middle temporal gyrus	598	7.84	-50	-2	-20	< .001
L superior medial frontal gyrus	670	7.79	-4	48	38	< .001
L middle frontal gyrus	481	7.69	-36	12	40	< .001
R middle temporal gyrus	54	7.44	50	-16	-14	< .001
L thalamus	283	6.86	-10	-16	0	< .001
L precuneus	142	6.71	-8	-56	30	< .001
L inferior frontal gyrus	54	6.56	-48	12	18	< .001
<i>(b) Sacred > non-sacred values; fact-checked > control trials</i>						
L hippocampus	339	5.29	-16	-26	-12	.010
L cerebellum	419	5.28	-30	-80	-30	.004
R superior medial frontal gyrus	1284	5.01	12	60	24	<.001
R cerebellum	425	4.98	24	-78	-38	.003
Cerebellar vermis	273	4.41	6	-68	-10	.024
R precuneus	402	4.19	10	-52	30	.004
<i>(c) Sacred > non-sacred values; control first > fact-checked first</i>						
L superior medial frontal gyrus	1257	6.98	-4	52	38	<.001
L inferior occipital gyrus	298	6.41	-32	-80	-2	.021
L middle temporal gyrus	2405	5.11	-52	-36	4	<.001
R Calcarine	658	4.96	6	-72	12	<.001
Cerebellar vermis	511	4.35	8	-52	-22	.002

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