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## Thinking Clearly About Misinformation

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26 **There is concern that many social problems in Western societies have been caused by**  
27 **misinformation. However, some researchers argue that misinformation is merely a**  
28 **symptom of such problems. We argue that (1) this is a false dichotomy, (2)**  
29 **misinformation has had clear impacts, and (3) researchers should consider the different**  
30 **dimensions of misinformation when making such evaluations.**

31 In Western societies, misinformation concern is at an all-time high. Recently,  
32 however, debate has ensued regarding the level of concern that is warranted. Some  
33 researchers note the potential for misinformation to incur significant costs on individuals and  
34 societies, and call for interventions to reduce misinformation susceptibility and impacts<sup>1,2</sup>.  
35 Others warn against “alarmist” narratives, arguing that misinformation exerts only limited  
36 influence over beliefs and behaviours. This view proposes that problematic behaviours, such  
37 as vaccine hesitancy, are caused by systemic socio-economic and psycho-social issues, and  
38 thus calls for interventions to target those societal issues rather than misinformation creation  
39 and consumption, which represent only “symptoms” of these deeper issues<sup>3</sup>. Similarly,  
40 assuming low prevalence of misinformation, researchers have argued that interventions  
41 should focus on increasing trust in factual information<sup>4</sup>.

42 A principled way to resolve these contradicting analyses is needed, to better inform  
43 policies and minimise the risk of enshrining a problematic status quo or investing resources  
44 to address a perhaps negligible problem. Here, we engage with two key questions: (1) Is  
45 misinformation a “symptom” rather than a “cause”? (2) Is misinformation consumption low  
46 and therefore not a reason for concern? In answering these questions, we argue that  
47 misinformation has had clear impacts; that depending on individual and contextual factors, it  
48 can be both a symptom and a cause; and that its multidimensionality (i.e., topic, type, and  
49 depth of dissemination) ought to be more fully considered when making such evaluations.

## 50 **A Call for Causal Clarity**

51           Societal issues can shape individuals' beliefs and produce problematic behaviours.  
52 Behaviours such as vaccine hesitancy and climate-change denial have been facilitated by  
53 factors such as populism, inequality, disenfranchisement, political polarisation, and the  
54 concentration of media ownership<sup>5</sup>. These factors are amplified by low institutional trust,  
55 which is a wicked problem because even if many institutions are generally trustworthy, some  
56 politicians, scientists, media outlets, and corporations have engaged in unethical behaviours  
57 that do warrant scepticism.

58           Yet, even if the misinformation problem is symptomatic of such deeper issues, this  
59 does not negate the fact that “symptoms” can cause outcomes of their own. To illustrate: A  
60 factor such as inequality might increase the “symptom” of misinformation susceptibility,  
61 while misinformation itself might cause belief changes or behaviours (unrelated to inequality)  
62 in a causal chain; alternatively, a factor such as polarization or institutional distrust might  
63 causally affect misinformation susceptibility, which in turn might further entrench  
64 polarization or distrust in a vicious cycle.

65           A counterfactual perspective can provide further clarification: causation is essentially  
66 the difference between a world in which a putative cause is present and a counterfactual  
67 world in which all is equal except for the absence of the cause. Thus, if misinformation were  
68 merely a “symptom,” then nothing in the world would change if all misinformation were to  
69 disappear. This is clearly implausible. Observational and experimental studies have  
70 demonstrated that misinformation can causally alter beliefs and behaviours<sup>1,6,7,8,9</sup>, even  
71 though measurement of misinformation impacts is often impeded by ethical considerations  
72 (e.g., exposing individuals to potentially harmful misinformation) or lack of access to  
73 relevant data (e.g., historical or transnational data; data from social-media platforms or closed  
74 channels such as offline communications and encrypted chat applications). Indeed, in a

75 counterfactual world without any misinformation, false beliefs could only emerge via  
76 spontaneous generation. Such spontaneous generation is not uncommon (e.g., stereotypes and  
77 superstitions can result from social processes or illusory correlations). However, it would be  
78 inadequate as an all-encompassing explanation for the spread of false beliefs that go beyond  
79 individuals' immediate experiences or observations. For example, the widespread false belief  
80 that the mumps-measles-rubella (MMR) vaccine causes autism would be unlikely to gain  
81 traction had fraudulent MMR-vaccine research not received high-profile media coverage.

82         Critically, the counterfactual perspective can account for multicausality. Consider a  
83 situation in which an individual is influenced by a claim that a vaccine is harmful. Both the  
84 misinformation and the existing susceptibility of the individual (e.g., low trust in science) are  
85 causal factors, if, without either, the individual would not have been misinformed to the same  
86 extent (e.g., formed a weaker misconception). Whether the misinformation or the existing  
87 susceptibility is a better explanation then depends on their relative prevalence and the  
88 probability of sufficiency. For example, in case of a fire breaking out after an individual  
89 lights a match, match-lighting may be a better explanation for the fire than the presence of  
90 oxygen, because oxygen is more prevalent than match-lighting and the individual lighting the  
91 match should have anticipated the presence of oxygen (such analyses are used in legal  
92 reasoning to determine damages)<sup>10</sup>. Thus, even if institutional mistrust can partially explain  
93 some individuals' tendency to be affected by vaccine misinformation (alongside other  
94 individual-specific factors such as perceived plausibility, worldview congruence, utility for  
95 behaviour justification, etc.), it does not absolve the causal responsibility of misinformants,  
96 nor negate the potential effects of vaccine misinformation on public health.

97         One way to capture the complexity of such causal networks is through directed  
98 acyclic graphs, as shown in Figure 1. This approach can also illustrate how existing research  
99 has focused on specific direct effects within limited timeframes, often neglecting more

100 indirect causal factors and potentially important context variables. For example, the existing  
 101 misinformation literature is biased towards a liberal-democratic, Western framework and has  
 102 largely overlooked the potential influence of environmental context factors such as state  
 103 capacity and the presence of ethnic conflicts or historical grievances, which may co-  
 104 determine misinformation impacts.

105 **Figure 1**

106 *Directed Acyclic Graphs Illustrating Causal Networks of Misinformation Effects*

107 [Figure 1]

108 *Note.* (a) Directed acyclic graphs are graphical causal models characterized by nodes  
 109 representing variables and edges representing direct causal effects. In the example, both low  
 110 institutional trust and misinformation can cause outcomes such as vaccine hesitancy.  
 111 Additionally, low trust and misinformation can have cross-lagged effects (e.g., low trust at  
 112 Time 1 causes more misinformation at Time 2), and there are likely other relevant factors ( $U_1$   
 113 and  $U_2$ ; e.g., technological and economic conditions, state capacity, or specific events); (b)  
 114 Research leveraging randomization, on average, controls for spurious factors and allows  
 115 causal identification for a subset of misinformation ( $Misinformation_R$ ). However, many  
 116 studies tend to focus on a limited timescale, estimating only specific direct effects and not  
 117 “total” effects (e.g., nodes and arrows within the red box, where the effects of prior  
 118 misinformation and other context factors ( $U_2$ ) might not be captured).

119 In sum, it is important to avoid a false dichotomy. The key question is not whether  
 120 misinformation is better framed as a symptom or a cause of social issues, but rather under  
 121 what conditions one framing is more appropriate than the other. We urge researchers to  
 122 precisely define the boundary conditions of their claims to avoid hasty generalizations, to  
 123 explicate any causal assumptions, and to systematically examine both individual and  
 124 contextual factors as conditions for causal influence. Although an extended discussion of  
 125 such factors is beyond scope, we discuss in the next section a selection of misinformation  
 126 dimensions that should be considered to appropriately recognize misinformation  
 127 heterogeneity.

## 128 **Recognizing Heterogeneity**

129            Objectively and easily identifiable misinformation, typically referred to as “fake  
130 news,” represents only a small portion of the average person’s information diet in Western  
131 societies<sup>11</sup>. However, in our view, (1) the misinformation problem should not be considered  
132 negligible because a subset of obvious misinformation has low prevalence, and (2) it is  
133 unreasonable to expect all types of misinformation to always have strong effects on all  
134 outcomes. Some studies will find misinformation has minimal effects, others may suggest the  
135 opposite<sup>8,9</sup>—as a generalization, both characterizations will be inaccurate unless qualified  
136 with explicit recognition of heterogeneity.

137            To this end, we direct attention to three key dimensions of misinformation—topic,  
138 type, and depth—that will influence its real-world reach and impact. The first dimension,  
139 topic, refers simply to the subject matter of the information. For instance, individuals in  
140 Country A will be impacted more by misinformation about a specific situation (e.g., an  
141 election) in Country A than similar misinformation regarding Country B.

142            Second, with regards to type, we follow McCright and Dunlap<sup>12</sup> in distinguishing  
143 between truthiness (misleading information that simply “feels” true), systemic lies (carefully  
144 crafted misinformation advancing ideological interests), bullshit (persuasive misinformation  
145 used opportunistically with total disregard for evidence), and shock-and-chaos (large volumes  
146 of content that aim to confuse or fatigue). Note that not all information captured in this  
147 framework will need to be literally false; for example, some information that is “truthy” or  
148 part of a shock-and-chaos approach might not be objectively false or even falsifiable (e.g., in  
149 a conflict situation, the narrative that the adversary is scared). Similarly, the selective,  
150 slanted, or miscontextualized presentation of true information can be used to mislead, an  
151 approach sometimes referred to as “paltering.” Table S1 applies this categorization to a  
152 selection of real-world misinformation. Considering this diversity, it becomes clear that much

153 misinformation is advanced—intentionally or unintentionally—by sources that would  
154 typically not be categorized as dubious by researchers estimating misinformation prevalence.  
155 For example, Grinberg and colleagues<sup>11</sup> focussed exclusively on websites known to publish  
156 fabricated stories. This leaves subtler types of misinformation outside of researchers’ tallies;  
157 if these neglected types are considered, misinformation will be found to occupy a greater  
158 portion of the information landscape.

159         The third key dimension, depth, relates to both distribution and repetition. The  
160 distributional aspect refers to whether the misinformation is dispersed haphazardly (e.g.,  
161 individual social-media posts or headlines) or if content is systematically bundled and/or  
162 targeted (e.g., an organized disinformation campaign; a revisionist history curriculum). The  
163 repetitional aspect relates to the well-known finding that repeated and thus familiar  
164 information is more likely judged to be true regardless of veracity<sup>1</sup>. Misinformation depth is  
165 important to consider because pieces of misinformation can have compound impacts<sup>13</sup>. Much  
166 like a river can be fed from multiple *tributaries*, multiple information sources can contribute  
167 to the same false narrative. This “narrative gist” can then be shared by downstream  
168 *distributaries*, which can include individuals never exposed to any initial misinformation, or  
169 news organizations that would never refer to the original low-quality sources. In this manner,  
170 misleading narratives can infiltrate mainstream news coverage and influence public discourse  
171 (e.g., conspiratorial claims influencing public debate during “Pizzagate”). Thus, assessing  
172 prevalence without accounting for narrative gist will systematically underestimate the scale  
173 of the misinformation problem.

174         Critically, potential outcomes can differ across misinformation types and depths, and  
175 can be undesirable even if the misinformation is identified. For example, a Republican  
176 correctly identifying *bullshit* from a Democrat might have lowered opinions of Democrats (or  
177 vice versa), which can fuel polarization even without any direct impact on beliefs. Even the

178 discourse surrounding misinformation itself can have negative effects (e.g., erode satisfaction  
179 with electoral democracy<sup>14</sup>). Figure 2 presents an idealized illustration of some potential  
180 misinformation impacts across types, depths, and outcomes.

181 **Figure 2**

182 *Potential Misinformation Effects Across Types, Depths, and Outcomes*

183 [Figure 2]

184 *Note.* Graphical illustration of some potential misinformation effects. Plotted data are  
185 hypothetical and on an arbitrary scale, to illustrate that different misinformation types at  
186 various depths can have different impacts across outcomes. For example, bullshit, even at  
187 high depth, may have minimal effects on beliefs, but may drive polarization and mistrust  
188 (even if identified as misleading). By contrast, paltering may affect beliefs without affecting  
189 trust (because fewer individuals will identify the misinformation).

190 A final point is that active forces can drive misinformation consumption. For instance,  
191 a vaccine-hesitant individual seeking vaccine information will encounter more vaccine  
192 misinformation than someone who is incidentally exposed. Moreover, vulnerable individuals  
193 may be targeted with misinformation tailored to their psychological vulnerabilities. If this has  
194 the potential to cause harm (to the individual or the public good), then it should be of  
195 concern, even if overall consumption is low. Caution is therefore needed when making  
196 general claims of prevalence and (lack of) impacts based on limited data.

197



**Box 1. Recommendations for Future Research**

199 First, as we have argued in more detail elsewhere<sup>15</sup>, a focus shift in misinformation-  
200 intervention evaluation is recommended. To illustrate: One of the most popular paradigms  
201 presents participants with large sets of true and false claims, with the difference in truth or  
202 belief ratings between the two taken as a measure of discernment. This paradigm limits  
203 studies to short-format misinformation (e.g., headlines, tweets), as tasking participants to  
204 engage with lengthier misinformation (e.g., articles, videos) in large sets can be impractical.  
205 This favours light-touch interventions that might not address persuasive misinformation at  
206 higher depth, even though such misinformation could be more impactful.

207 Second, future research should make more use of observational causal-inference  
208 strategies. Regardless of how realistic or incentivized laboratory-based measures can be, it  
209 remains true that many factors are not manipulable due to ethical or feasibility  
210 considerations. For example, researchers have used the positioning of cable-TV channels  
211 (which varies randomly across localities in the US) in instrumental-variable analyses showing  
212 that exposure to unreliable news sources reduced social-distancing behaviours during the  
213 early stages of the COVID-19 pandemic<sup>9</sup>. However, further studies in other domains are  
214 needed.

215 Finally, as an integrative account of false beliefs is lacking, another promising  
216 direction is to borrow from the broader cognitive-science literature. For instance, cognitive  
217 research has shown that individuals preferentially rely on “gist” representations of  
218 quintessential meanings<sup>13</sup>. Future research attempting to delineate the evolution of narrative  
219 gist at a societal level might therefore benefit from first examining gist processing at the  
220 individual level. Cognitive models of decision making could also be used to explore  
221 misinformation impact beyond observable outcomes. For example, evidence-accumulation  
222 models could be used to decompose choice and response-time data into cognitively  
223 interpretable parameters (e.g., response boundaries represent the varying levels of evidence  
224 individuals require to make decisions and could be interpreted as caution).

**226 Conclusions and Recommendations**

227 Taken together, a clear implication of our discussion is that the standard paradigms  
228 used for evaluating the impacts of misinformation and misinformation interventions are likely  
229 insufficient. Some recommendations for changes to current research practice in the field are  
230 provided in Box 1. We have argued that the evaluation of misinformation impacts is an  
231 important, but complex, research question, particularly in the current era of rising geopolitical  
232 tensions and rapid technological change. We hope that the current Comment will contribute

233 to increasingly nuanced debates about the impact of misinformation and potential  
234 interventions.

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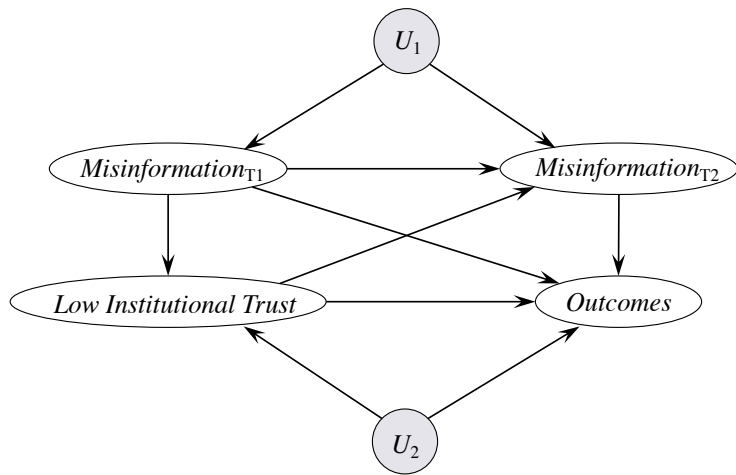
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266 **Competing Interests**

267 The authors declare no competing interests.

(a)



(b)

