

Prompting better sharing quality of COVID-related headlines

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Abstract

Asking people to evaluate the accuracy of a non-COVID-related headline decreases their likelihood to share COVID-related headlines, especially if they are false.

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1 Target article

2 A. A. Arechar, J. Allen, A. J. Berinsky, R. Cole, Z. Epstein, K. Garimella, A. Gully, J. G. Lu,
3 R. M. Ross, M. N. Stagnaro, Y. Zhang, G. Pennycook *et al.*, *Understanding and combatting*
4 *misinformation across 16 countries on six continents*, *Nature Human Behaviour* 7(9), 1502
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6 1 Goal

7 This robustness report investigates whether an alternative analysis supports the increase in
8 sharing quality after encountering an accuracy prompt reported by [1].

9 2 Methods

10 Sharing behaviour is assessed using an online questionnaire with people rating how likely they
11 would share a COVID-related article based on the headline (outcome *sharing likelihood* on a
12 scale from (1) = ‘moderately unlikely’ to (6) = ‘extremely likely’). Half of the headlines report
13 false while the other half report true statements (predictor *Truth*). I define sharing quality
14 as the difference between true and false headlines. Data was collected in 16 countries, us-
15 ing translated headlines. Some participants encounter an accuracy prompt before rating their
16 sharing likelihood (predictor *Condition*). This prompt asks them to evaluate the accuracy of a
17 single, non-COVID-related headline. The online questionnaire also contains attention checks
18 and screeners. I excluded the data of 8,592 participants who failed at least one of these tests,
19 whose total duration was longer than the 99% percentile (94.55 min) or who chose the exact
20 same rating for all 20 headlines they encountered. For the included participants ($n = 8,587$), I

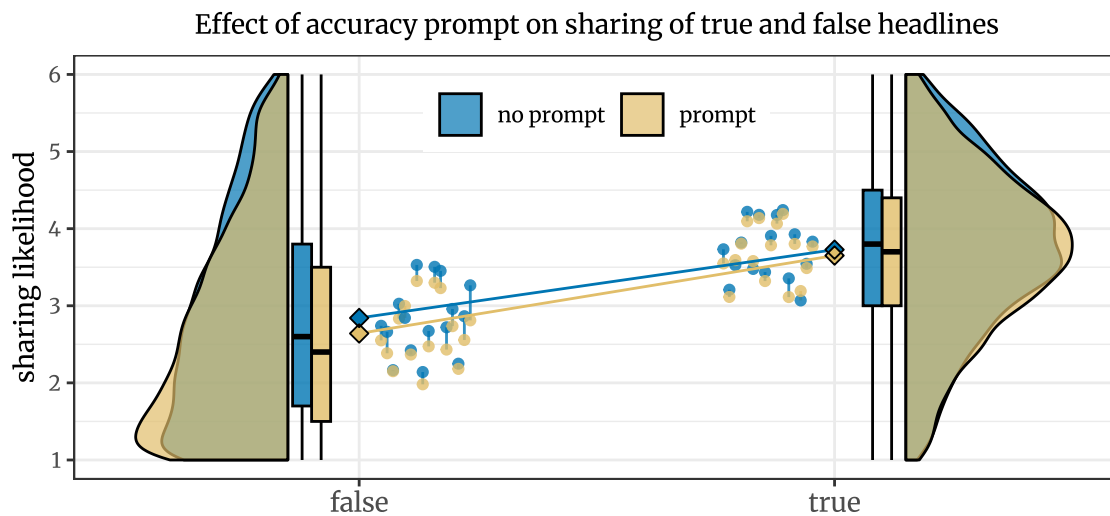


Figure 1: Density distributions and boxplots of participant averages. Dots show country averages with lines in the colour of the condition leading to a higher mean sharing likelihood. Diamonds show averages over all participants, regardless of country. Sharing likelihood was lower in the prompt condition for both true and false headlines, a pattern that was observed in almost all countries. However, this effect was slightly more pronounced for false headlines.

21 computed separate averages for their sharing likelihood of true and false headlines. These av-
 22 erages were modelled with a Gaussian Bayesian linear mixed model implemented in brms [2].
 23 The model included two sum-coded population-level predictors, *Truth* (true or false headline)
 24 and *Condition* (prompt or no prompt), and their interaction as well as group-level intercepts
 25 for participant and country. Additionally, I included slopes for both population-level predictors
 26 and their interaction for each country. Posterior predictive checks revealed slight deviations
 27 in the overall shape but good capture of predictor means in the real data by predicted means
 28 based on the model. Standardised effect sizes δ were computed by dividing effects by the
 29 square root of the sum of all squared variances following [3].

30 3 Results

31 Overall, participants were more likely to share true than false headlines (*estimate* = 0.96 [0.85,
 32 1.07], *posterior probability* = 100%; δ = 0.783 [0.665, -0.897]). Participants who received an
 33 accuracy prompt before rating sharing likelihoods exhibited a credibly better sharing quality
 34 than participants who did not receive such a prompt, meaning that there was a greater dif-
 35 ference between sharing likelihoods of true and false headlines (*estimate* = 0.12 [0.06, 0.18],
 36 *posterior probability* = 99.76%; δ = 0.098 [0.036, 0.160]). Participants in the prompt condi-
 37 tion rated their likelihood to share lower (*estimate* = -0.13 [-0.17, -0.08], *posterior probability*
 38 = 99.99%; δ = -0.103 [-0.147, -0.059]), with this reduction applying to both true (*estimate*
 39 = -0.07 [-0.12, -0.01], *posterior probability* = 97.98%; δ = -0.054 [-0.104, -0.002]) and false
 40 headlines (*estimate* = -0.19 [-0.24, -0.13], *posterior probability* = 100%; δ = -0.152 [-0.208,
 41 -0.096]). However, the reduction was more pronounced for false headlines (see 1).

42 4 Conclusion

43 My results indicate that sharing quality was improved by a preceding accuracy prompt: while
44 people rated their overall likelihood to share lower for both true and false headlines, this effect
45 was more pronounced for false headlines. Thus, this re-analysis supports the original claim by
46 [1] using a Bayesian approach.

47 Acknowledgments and Disclosures

48 **Reproducibility** We were able to computationally reproduce the original analysis and re-
49 sults.

50 **Code and Data Availability** Data and R code are available on OSF: <https://osf.io/7wgv2/>.

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