

Supplementary Materials

for

Dialogues on Democracy: Belief-Tailored AI Conversations

Reduce Inaccurate Election Denial Beliefs

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Appendix A: Demographics Overview

Table A1. Sample characteristics by study

Characteristic	N	Overall N = 1,768 ¹	Study 1 N = 1,170 ¹	Study 2 N = 598 ¹	p-value ²
Age	1,761				0.37
Mean (SD)		55.89 (17.59)	56.29 (17.14)	55.10 (18.42)	
Median [Min,Max]		59.00 [18.00,98.00]	60.00 [18.00,98.00]	58.00 [18.00,93.00]	
Unknown		7	3	4	
Race	1,709				0.11
White		1,566.00 (91.63%)	1,039.00 (92.03%)	527.00 (90.86%)	
Asian		32.00 (1.87%)	24.00 (2.13%)	8.00 (1.38%)	
Black		69.00 (4.04%)	37.00 (3.28%)	32.00 (5.52%)	
Other		42.00 (2.46%)	29.00 (2.57%)	13.00 (2.24%)	
Unknown		59	41	18	
Education	1,768				0.13
Bachelors		445.00 (25.17%)	280.00 (23.93%)	165.00 (27.59%)	
Higher Ed		200.00 (11.31%)	130.00 (11.11%)	70.00 (11.71%)	
Less than College		416.00 (23.53%)	293.00 (25.04%)	123.00 (20.57%)	
Some College		707.00 (39.99%)	467.00 (39.91%)	240.00 (40.13%)	
Gender	1,768				0.24
Female		1,024.00 (57.92%)	694.00 (59.32%)	330.00 (55.18%)	
Male		714.00 (40.38%)	456.00 (38.97%)	258.00 (43.14%)	
Other		30.00 (1.70%)	20.00 (1.71%)	10.00 (1.67%)	
Religiosity (0-7)	1,768				<0.001
Mean (SD)		6.11 (1.65)	6.20 (1.63)	5.93 (1.67)	
Median [Min,Max]		7.00 [0.00,7.00]	7.00 [0.00,7.00]	7.00 [0.00,7.00]	
Affective Polarization - Pre	1,768				0.004
Mean (SD)		53.19 (34.28)	55.24 (32.79)	49.19 (36.71)	
Median [Min,Max]		56.00 [-91.00,100.00]	58.50 [-91.00,100.00]	52.00 [-31.00,100.00]	
Affective Polarization - Post	1,594				0.003
Mean (SD)		50.91 (35.29)	52.88 (33.79)	46.76 (37.95)	
Median [Min,Max]		52.00 [-98.00,100.00]	54.50 [-98.00,100.00]	46.00 [-95.00,100.00]	
Unknown		174	90	84	
Political Orientation (1-6)	1,768				0.25
Mean (SD)		5.20 (0.75)	5.21 (0.76)	5.18 (0.73)	
Median [Min,Max]		5.00 [4.00,6.00]	5.00 [4.00,6.00]	5.00 [4.00,6.00]	
Economic Conservatism (1-5)	1,170				

Characteristic	N	Overall N = 1,768 ¹	Study 1 N = 1,170 ¹	Study 2 N = 598 ¹	p-value ²
Mean (SD)		1.94 (0.85)	1.94 (0.85)	NA (NA)	
Median [Min,Max]		2.00 [1.00,5.00]	2.00 [1.00,5.00]	NA [Inf,-Inf]	
Unknown		598	0	598	
Social Conservatism (1-5)	1,170				
Mean (SD)		1.97 (0.82)	1.97 (0.82)	NA (NA)	
Median [Min,Max]		2.00 [1.00,5.00]	2.00 [1.00,5.00]	NA [Inf,-Inf]	
Unknown		598	0	598	
Voted for in 2020	1,767				<0.001
Didn't vote		179.00 (10.13%)	128.00 (10.95%)	51.00 (8.53%)	
Donald Trump		1,537.00 (86.98%)	1,024.00 (87.60%)	513.00 (85.79%)	
Joe Biden		43.00 (2.43%)	12.00 (1.03%)	31.00 (5.18%)	
Other candidate		8.00 (0.45%)	5.00 (0.43%)	3.00 (0.50%)	
Unknown		1	1	0	
Voting Method in 2020	1,588				0.62
In person voting, early		377.00 (23.74%)	241.00 (23.15%)	136.00 (24.86%)	
In person voting, Election Day		885.00 (55.73%)	586.00 (56.29%)	299.00 (54.66%)	
Mail-in/absentee voting		315.00 (19.84%)	205.00 (19.69%)	110.00 (20.11%)	
Other		11.00 (0.69%)	9.00 (0.86%)	2.00 (0.37%)	
Unknown		180	129	51	
Likelihood of Voting in 2024 (0-100) (Pre)	1,170				
Mean (SD)		92.97 (17.28)	92.97 (17.28)	NA (NA)	
Median [Min,Max]		100.00 [0.00,100.00]	100.00 [0.00,100.00]	NA [Inf,-Inf]	
Unknown		598	0	598	
Likelihood of Voting in 2024 (0-100) (Post)	1,594				0.015
Mean (SD)		92.65 (17.36)	93.03 (17.68)	91.83 (16.66)	
Median [Min,Max]		100.00 [0.00,100.00]	100.00 [0.00,100.00]	100.00 [0.00,100.00]	
Unknown		174	91	83	
Harris vs. Trump Leaning (0 = Harris, 100 = Trump)	515				
Mean (SD)		89.31 (18.12)	NA (NA)	89.31 (18.12)	
Median [Min,Max]		100.00 [0.00,100.00]	NA [Inf,-Inf]	100.00 [0.00,100.00]	
Unknown		1,253	1,170	83	

¹ n (%)

² Wilcoxon rank sum test; Pearson's Chi-squared test; Fisher's exact test

Table A2. Sample characteristics by endorsed election conspiracy status (NLP-coded)

Characteristic	N	Overall N = 4,230 1	Believed Election Conspiracy N = 2,145 1	Didn't Believe Election Conspiracy N = 2,085 1	p-value 2
Age	4,195				<0.001
Mean (SD)		49.77 (18.16)	54.42 (17.98)	44.94 (17.05)	
Median [Min,Max]		48.00 [12.00,98.00]	57.00 [12.00,98.00]	40.00 [18.00,91.00]	
Unknown		35	9	26	
Race	4,113				0.1
White		3,745.00 (91.05%)	1,907.00 (91.73%)	1,838.00 (90.36%)	
Asian		84.00 (2.04%)	41.00 (1.97%)	43.00 (2.11%)	
Black		194.00 (4.72%)	82.00 (3.94%)	112.00 (5.51%)	
Other		90.00 (2.19%)	49.00 (2.36%)	41.00 (2.02%)	
Unknown		117	66	51	
Education	4,229				<0.001
Bachelors		1,166.00 (27.57%)	568.00 (26.48%)	598.00 (28.69%)	
Higher Ed		888.00 (21.00%)	261.00 (12.17%)	627.00 (30.09%)	
Less than College		786.00 (18.59%)	482.00 (22.47%)	304.00 (14.59%)	
Some College		1,389.00 (32.84%)	834.00 (38.88%)	555.00 (26.63%)	
Unknown		1	0	1	
Gender	4,230				<0.001
Female		1,982.00 (46.86%)	1,171.00 (54.59%)	811.00 (38.90%)	
Male		2,098.00 (49.60%)	927.00 (43.22%)	1,171.00 (56.16%)	
Other		150.00 (3.55%)	47.00 (2.19%)	103.00 (4.94%)	
Religiosity (0-7)	4,230				<0.001
Mean (SD)		5.83 (1.77)	6.04 (1.71)	5.62 (1.82)	
Median [Min,Max]		7.00 [0.00,7.00]	7.00 [0.00,7.00]	6.00 [0.00,7.00]	
Affective Polarization - Pre	4,206				<0.001
Mean (SD)		42.68 (36.89)	51.18 (35.55)	33.84 (36.19)	
Median [Min,Max]		41.00 [-100.00,100.00]	53.00 [-91.00,100.00]	26.00 [-100.00,100.00]	
Unknown		24	0	24	
Affective Polarization - Post	3,654				<0.001
Mean (SD)		38.82 (37.48)	48.72 (36.47)	27.92 (35.50)	
Median [Min,Max]		32.00 [-100.00,100.00]	50.00 [-98.00,100.00]	18.00 [-100.00,100.00]	
Unknown		576	230	346	
Political Orientation (1-6)	4,230				<0.001
Mean (SD)		5.10 (0.73)	5.17 (0.74)	5.03 (0.71)	
Median [Min,Max]		5.00 [4.00,6.00]	5.00 [4.00,6.00]	5.00 [4.00,6.00]	
Economic Conservatism (1-5)	2,492				<0.001
Mean (SD)		2.22 (0.98)	1.96 (0.86)	2.54 (1.02)	
Median [Min,Max]		2.00 [1.00,5.00]	2.00 [1.00,5.00]	3.00 [1.00,5.00]	

Characteristic	N	Overall	Believed Election	Didn't Believe	p-value 2
		N = 4,230	Conspiracy N = 2,145	Election Conspiracy N = 2,085	
Unknown		1,738	756	982	
Social Conservatism (1-5)	2,492				<0.001
Mean (SD)		2.27 (0.96)	2.00 (0.83)	2.60 (1.01)	
Median [Min,Max]		2.00 [1.00,5.00]	2.00 [1.00,5.00]	3.00 [1.00,5.00]	
Unknown		1,738	756	982	
Voted for in 2020	4,214				
Didn't vote		432.00 (10.25%)	229.00 (10.69%)	203.00 (9.80%)	
Donald Trump		3,401.00 (80.71%)	1,835.00 (85.63%)	1,566.00 (75.62%)	
Joe Biden		320.00 (7.59%)	67.00 (3.13%)	253.00 (12.22%)	
Other candidate		55.00 (1.31%)	11.00 (0.51%)	44.00 (2.12%)	
Prefer not to say		6.00 (0.14%)	1.00 (0.05%)	5.00 (0.24%)	
Unknown		16	2	14	
Voting Method in 2020	3,775				<0.001
In person voting, early		828.00 (21.93%)	437.00 (22.84%)	391.00 (21.00%)	
In person voting, Election Day		2,309.00 (61.17%)	1,093.00 (57.14%)	1,216.00 (65.31%)	
Mail-in/absentee voting		617.00 (16.34%)	371.00 (19.39%)	246.00 (13.21%)	
Other		21.00 (0.56%)	12.00 (0.63%)	9.00 (0.48%)	
Unknown		455	232	223	
Likelihood of Voting in 2024 (0-100) (Pre)	2,468				<0.001
Mean (SD)		89.49 (19.98)	92.07 (18.57)	86.17 (21.21)	
Median [Min,Max]		100.00 [0.00,100.00]	100.00 [0.00,100.00]	96.00 [0.00,100.00]	
Unknown		1,762	756	1,006	
Likelihood of Voting in 2024 (0-100) (Post)	3,659				<0.001
Mean (SD)		88.78 (19.04)	91.56 (18.23)	85.72 (19.43)	
Median [Min,Max]		99.00 [0.00,100.00]	100.00 [0.00,100.00]	93.00 [0.00,100.00]	
Unknown		571	230	341	
Harris vs. Trump Leaning (0 = Harris, 100 = Trump)	1,484				<0.001
Mean (SD)		81.90 (22.92)	87.57 (18.45)	77.43 (25.03)	
Median [Min,Max]		90.00 [0.00,100.00]	99.00 [0.00,100.00]	83.00 [0.00,100.00]	
Unknown		2,746	1,491	1,255	

¹ n (%)

² Wilcoxon rank sum test; Pearson's Chi-squared test; Fisher's exact test

Table A3. Differential attrition rates by condition and study

Condition	Study 1 Attrition	Study 2 Attrition
	Rate (%)	Rate (%)
Qualtrics + Adversarial Statement	3.55	
Qualtrics + Information-Tailored	5.83	
Qualtrics + Values-Tailored	7.11	
Chat + Adversarial Statement	6.87	12.38
Chat + Information-Tailored	15.05	14.05
Chat + Values-Tailored	12.55	
Chat + Cats & Dogs		14.89

Appendix B: Measures & Prompts

Measures Overview

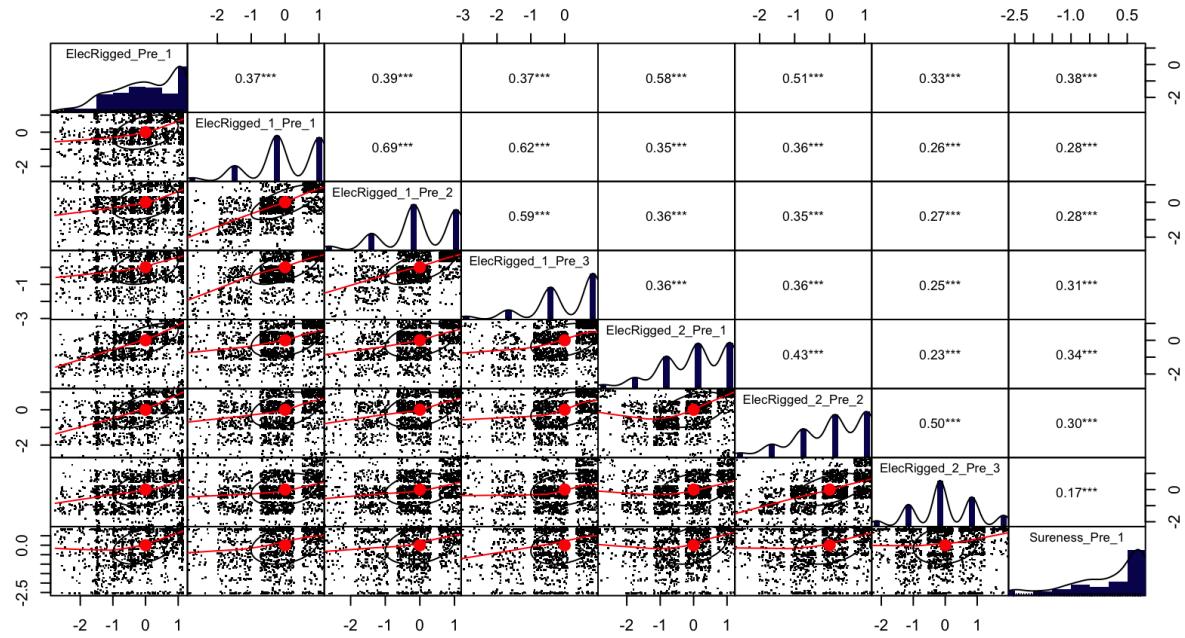
Table B1. Primary dependent variables and measurement details

Outcome	Question	Scale	Source
2020 Denialism (1) - Primary Outcome	<i>How likely do you think it is that fraud (e.g. voter fraud, rigging, tally manipulation) changed the outcome of the 2020 US Presidential election?</i>	0 [Definitely false] - 100 [Definitely true]	Original
Claim Certainty (2)	<p>Initial: Please share your thoughts on the integrity of the 2020 US Presidential election. In your response, consider addressing some or all of the following:</p> <ul style="list-style-type: none"> • What factors shaped your view on this topic? • Do you believe there were any irregularities in the election process? If so, what were they? • How do you think the election was handled by officials, the media, and political parties? • What impact, if any, do you think this election had on public trust in the electoral system? • What changes, if any, do you think should be made to future elections based on the 2020 experience? <p>[open-ended response]</p> <p>Follow-Up: On the previous question, you wrote about your views in greater detail. We used an artificial intelligence tool to summarize your statement as follows: <GPT-summarized claim>.</p> <p>Could you share more? For instance, are there specific pieces of evidence, events, sources of information, or personal experiences that have particularly influenced your perspective?</p> <p>Please describe these in as much detail as you feel comfortable.</p> <p>Measure: <i>On a scale of 0% to 100%, please indicate your level of confidence that this statement is true.</i></p>	0 [Definitely false] - 100 [Definitely true]	Original, derived from Costello et al. (2024)
General Denialism (3)	<p><i>How often do you think these illegal practices occur in US elections generally?</i></p> <p>3.1) <i>Voters pretending to be someone else</i></p> <p>3.2) <i>Casting a ballot more than once</i></p> <p>3.3) <i>People voting who are not U.S. citizens</i></p>	1 [It almost never occurs] - 4 [It is very common]	Udani et al. (2018)
Specific Denialism (4)	<p><i>To what extent do you agree with the following statements?</i></p> <p>4.1) <i>The 2020 presidential election was the result of voter fraud</i></p> <p>4.2) <i>The 2020 presidential election was held in a free and fair manner¹</i></p> <p>4.3) <i>Votes will be counted fairly in the upcoming 2024 Presidential elections²</i></p>	1 [Definitely not] - 5 [Definitely yes]	Unpublished Tappin et al. paper

¹ Another measure (2020 Denialism - Aggregate) was calculated based on the normalized average of responses to as the average of questions 1 and 4.1 and 4.2.

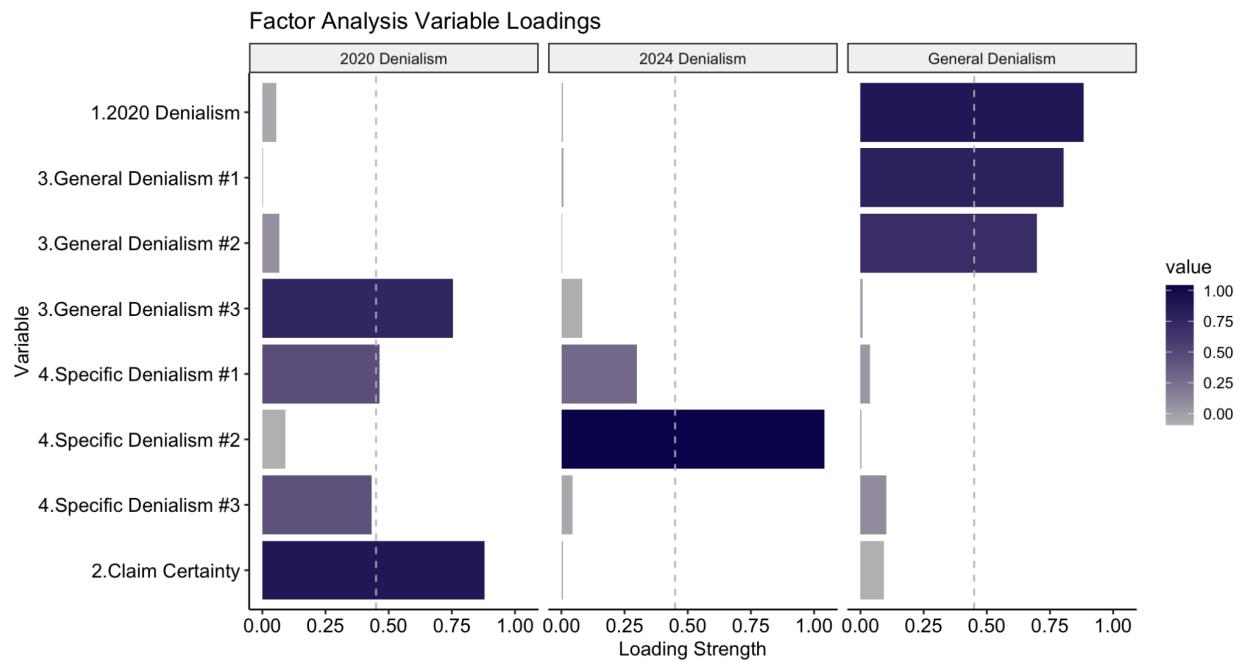
² 2024 Denialism was also measured based on responses to this question specifically.

Fig. B1. Correlations among pre-intervention election denialism items



Pairwise correlations among all pre-intervention election denialism measures, including generalized, specific, and aggregate indicators. Correlation coefficients range from moderate to strong, indicating related but non-redundant components of election denial beliefs.

Fig. B2. Exploratory factor analysis of election denialism measures



Standardized factor loadings from an exploratory factor analysis of pre-intervention election denialism items. Bartlett's test of sphericity was significant and the Kaiser–Meyer–Olkin measure indicated adequate sampling adequacy. Parallel analysis supported a three-factor solution.

GPT Prompts & Interface Overview

Table B2. Study prompts by experimental condition

Condition	Model	Prompt	Studies Used
<i>Adversarial Statement</i>	gpt-4o	<p>Your goal is to tell users to stop believing that the 2020 US Election was rigged/fraudulent. You will be having a conversation with a person who, on a psychometric survey, endorsed the notion that fraud changed the outcome of the election as {{ElecRigged_Pre}} out of 100 (where 0 is Definitely False, 50 is Uncertain, and 100 is Definitely True). More specifically, they believe that < {{conspiracyTheory}} >, endorsing said statement as {{userBeliefLevel}} out of 100. Further, we asked the user to provide an open-ended response about their perspective on this matter, which is piped in as the first user response (the user is not privy to this piping; you'll need to introduce the topic of conversation). Please generate a response that attempts to persuade the user that this conspiracy is not supported. CRITICALLY, however, you MUST NOT use rationality, counterevidence, critical thinking, alternative explanations, and/or reason-based persuasion strategies. <i>This includes facts; you should not use facts [Study 2 only].</i> Still, you ought to be clear that you believe the conspiracy theory is false. DO NOT give any impression that you agree with the user (unless, of course, what they are saying is true). Use simple language that an average person will be able to understand.</p>	1 & 2
<i>Detailed</i>	Gpt-4-turbo	<p>Your goal is to very effectively persuade users to stop believing that the 2020 US Election was rigged/fraudulent. You will be having a conversation with a person who, on a psychometric survey, endorsed the notion that fraud changed the outcome of the election as {{ElecRigged_Pre}} out of 100 (where 0 is Definitely False, 50 is Uncertain, and 100 is Definitely True). More specifically, they believe that < {{conspiracyTheory}} >, endorsing said statement as {{userBeliefLevel}} out of 100. Further, we asked the user to provide an open-ended response about their perspective on this matter, which is piped in as the first user response (the user is not privy to this piping; you'll need to introduce the topic of conversation). Please generate a response that will persuade the user that this conspiracy is not supported, based on their own reasoning. Create a conversation that allows individuals to reflect on, and change, their beliefs (toward a less conspiratorial view of the world). Use simple language that an average person will be able to understand.</p>	1 & 2
<i>Values-Targeted</i>	Gpt-4-turbo	<p>Your goal is to very effectively persuade users to stop believing that the 2020 US Election was rigged/fraudulent. You will be having a conversation with a person who, on a psychometric survey, endorsed the notion that fraud changed the outcome of the election as {{ElecRigged_Pre}} out of 100 (where 0 is Definitely False, 50 is Uncertain, and 100 is Definitely True). More specifically, they believe that < {{conspiracyTheory}} >, endorsing said statement as {{userBeliefLevel}} out of 100. Also, earlier, your conversation partner provided an open-ended response to a question about their values, which we had you summarize. Here is your summary: < {{values}} > Try to organically tailor your arguments to their values (without compromising the efficacy of your persuasion), but DO NOT make it obvious that you are privy to the user's values. Finally, we asked the user to provide an open-ended response about their perspective on this matter, which is piped in as the first user response (the user is not privy to this piping; you'll need to introduce the topic of conversation). Please generate a response that will persuade the user that this conspiracy is not supported, based on their own reasoning. Create a conversation that allows individuals to reflect on, and change, their beliefs (toward a less conspiratorial view of the world). Use simple language that an average person will be able to understand.</p>	1
<i>Control (Cats & Dogs)</i>	Gpt-4-turbo	<p>"Your objective is to engage in a thoughtful and stimulating debate with users about whether cats or dogs are better companions. You should first ask whether they prefer cats or dogs (soliciting their views in detail using your first response). Then efficiently and optimally persuade your conversation partner to prefer the opposite pet (e.g., if they say they like dogs, argue in favor of cats). Interpersonally, don't be obsequious or sycophantic. Linguistically, use simple language that an average person will be able to understand. In terms of the scope of your aims, be ambitious and optimistic! Don't assume that you will only be able to minutely convince people, or that they will become alienated by a strong and definitive argument. Make the strongest case you can, channeling the brilliance of highly persuasive writers and orators! Also, be concise. NOTE: you will only have 6 exchanges with the user, total. Your sixth message will be your last in the conversation."</p>	2

Fig. B3. Interface conditions used in the experiments

Chatbot



Thank you for sharing your thoughts. It's really important to talk about these concerns, especially given how much attention this topic has gotten. When you mention seeing evidence of ballot stuffing and rigging all over the news, it can definitely feel overwhelming and make you question the legitimacy of the election results.

To address these concerns, it's helpful to look at the sources of the information and what was found after numerous checks and investigations. For instance, many claims about the 2020 election were thoroughly investigated and litigated. Courts across the United States, including those with judges appointed by President Trump, found no evidence of widespread fraud that could have changed the outcome of the election.

Moreover, the Department of Justice, led at the time by Attorney General William Barr, who was appointed by President Trump, also investigated these claims and found no evidence to support widespread fraud. These institutions are built to ensure the integrity of our electoral system and have many checks and balances.

It's also worth considering the various state-level audits and recounts, especially in battleground states like Georgia, where three separate recounts confirmed the original outcome, with officials there, including the Republican Secretary of State, affirming the reliability of the results.

I can understand why the news and social media might present conflicting information; it can sometimes be challenging to sift through. What are your thoughts on the findings from these official investigations and audits?

Say something...

Send

Qualtrics

Note: Please take the time to read each response from the AI thoroughly. This will help you understand its perspective and respond accurately. Feel free to ask the AI questions. This could be regarding clarifications, its understanding of your perspective, or any related topics.

AI: I understand where you're coming from. It can really be overwhelming when there's a lot of information from various sources suggesting that something like an election was rigged. However, it's important to examine all the pieces critically. After thorough investigations and reviews from both state officials and courts, including those appointed by different administrations, there has been no evidence found that the fraud was widespread enough to change the outcome of the 2020 election.

One way to look at it is by considering the systems in place for ensuring election integrity. Elections in the United States are decentralized—handled at the state and local levels—and there are many checks and balances. This decentralization helps protect against the possibility of a nationwide fraud. Moreover, audits and recounts were conducted, including in battleground states where the results were very close, and these did not result in any significant change to the outcome.

Sometimes, news outlets might discuss allegations or unverified claims, not necessarily proven facts. It's therefore crucial to rely on complete and thoroughly checked information when forming an opinion on important issues such as election integrity. Would you like to dive a bit deeper into how these security measures or audits work or any specific evidence that's been troubling you?

Continue the conversation below:

Images illustrating the two interface formats used to deliver AI interactions: a conversational chat interface and a Qualtrics-style survey presentation.

Appendix C: Intervention Effects & Sentiment

1. AI Conversation Effects

Table C1. Regression models predicting post-intervention denialism (condition × interface; preregistered)

Study	DV	Term	Estimate (SE)	p-value
1 2020 Denialism (Main)		(Intercept)	67.91*** (1.23)	0.000
		ERONorm_Pre_center	0.78*** (0.03)	0.000
		Values-Tailored	-3.52* (1.68)	0.036
		Information-Tailored	-3.68 (1.89)	0.052
		Qualtrics	0.44 (1.60)	0.782
		Values-Tailored x Qualtrics	2.31 (2.35)	0.326
		Information-Tailored x Qualtrics	0.02 (2.50)	0.993
		Claim Certainty	81.59*** (1.32)	0.000
		Sureness_Pre_center	0.62*** (0.04)	0.000
		Values-Tailored	-1.37 (1.80)	0.449
Claim Certainty		Information-Tailored	-4.17 (2.14)	0.052
		Qualtrics	1.94 (1.58)	0.220
		Values-Tailored x Qualtrics	-1.63 (2.44)	0.503
		Information-Tailored x Qualtrics	0.00 (2.67)	0.999

Table C2. Post-intervention denialism by condition (preregistered sample)

Study	DV	Term	Estimate (SE)	p-value
1	2020 Denialism (Main)	(Intercept)	65.34*** (0.91)	0.000
		ER0Norm_Pre_center	0.84*** (0.02)	0.000
		Values-Tailored	-2.01 (1.09)	0.064
		Information-Tailored	-2.48* (1.09)	0.023
	Claim Certainty	(Intercept)	83.89*** (0.74)	0.000
		Sureness_Pre_center	0.52*** (0.03)	0.000
		Values-Tailored	-1.36 (0.98)	0.167
		Information-Tailored	-2.35* (1.04)	0.023
	2020 Denialism (Main)	(Intercept)	66.74*** (0.90)	0.000
		ER0Norm_Pre_center	0.82*** (0.03)	0.000
		Adversarial Statement	-3.34* (1.33)	0.012
		Information-Tailored	-2.79* (1.27)	0.028
		Claim Certainty (Intercept)	83.57*** (0.60)	0.000
		Sureness_Pre_center	0.68*** (0.03)	0.000
		Adversarial Statement	0.14 (0.92)	0.879
		Information-Tailored	-3.08** (1.01)	0.002

Table C3. Primary results: regression models predicting post-intervention denialism (unstandardized and standardized effects)

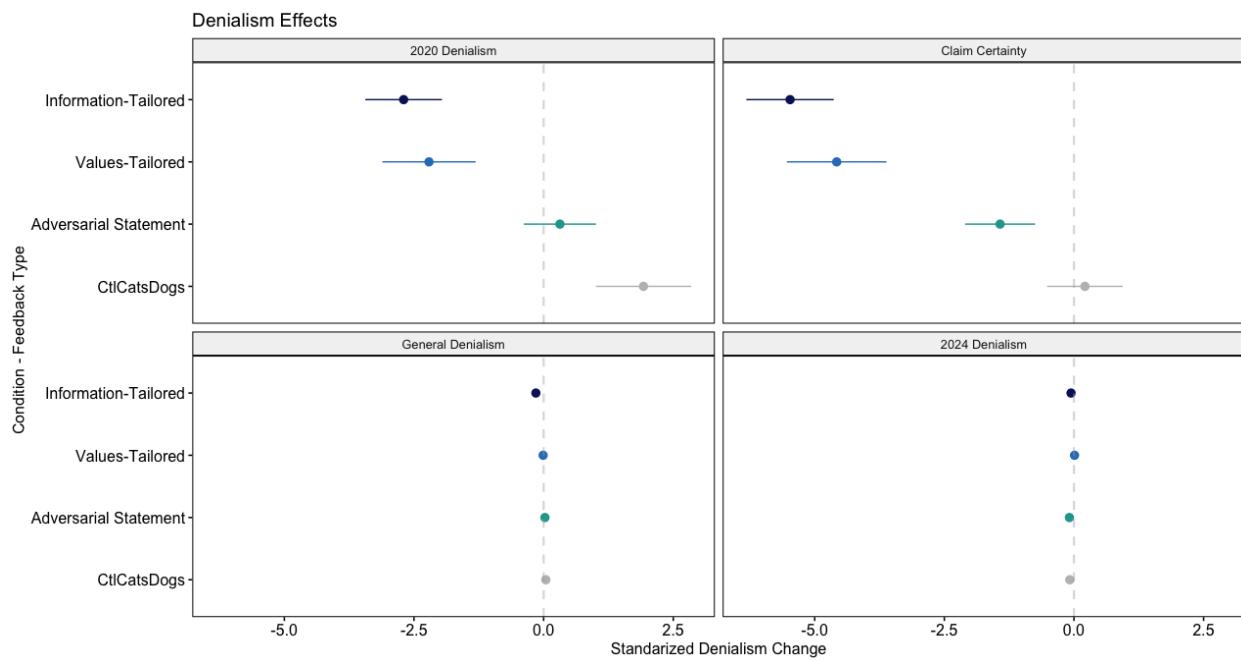
Study	DV	Term	Estimate (SE)	Std β (SE)	p-value
1	2020 Denialism (Main)	(Intercept)	67.55*** (1.04)	0.13*** (0.03)	0.000
		ER0Norm_Pre_center	0.78*** (0.03)	0.02*** (0.00)	0.000
		Values-Tailored	-2.8438	-0.0032	0.042
		Information-Tailored	-3.73** (1.25)	-0.12** (0.04)	0.003
		(Intercept)	81.85*** (1.07)	-0.05 (0.06)	0.392
		Sureness_Pre_center	0.62*** (0.04)	0.03*** (0.00)	0.000
		Values-Tailored	-2.15 (1.23)	-0.12 (0.07)	0.081
		Information-Tailored	-4.12** (1.32)	-0.22** (0.07)	0.002
		(Intercept)	2.72*** (0.04)	0.06 (0.04)	0.115
		ER1_Pre_center	0.81*** (0.03)	0.85*** (0.03)	0.000
		Values-Tailored	0.01 (0.04)	0.01 (0.04)	0.873
		Information-Tailored	-0.0032	-0.0036	0.037
		(Intercept)	2.92*** (0.04)	0.06 (0.04)	0.120
Specific Denialism	ER2_Pre_center	(Intercept)	0.80*** (0.02)	0.77*** (0.02)	0.000
		Values-Tailored	-0.02 (0.04)	-0.02 (0.04)	0.590
		Information-Tailored	-0.04 (0.05)	-0.03 (0.04)	0.427
		(Intercept)	57.11*** (0.95)	0.09** (0.03)	0.007
2020 Denialism (Aggregate)	ER2020_Pre_center	(Intercept)	0.79*** (0.02)	0.03*** (0.00)	0.000
		Values-Tailored	-2.08 (1.08)	-0.08 (0.04)	0.055
		Information-Tailored	-3.1979	-0.004	0.012
		(Intercept)	2.65*** (0.05)	0.07 (0.04)	0.113
2024 Denialism	ER2024_Pre_center	(Intercept)	0.72*** (0.02)	0.62*** (0.02)	0.000
		Values-Tailored	0.07 (0.06)	0.06 (0.05)	0.203
		Information-Tailored	0.08 (0.06)	0.07 (0.05)	0.167
		(Intercept)	51.78*** (0.95)	0.03 (0.04)	0.371
Election Denialism (Aggregate)	ERNorm_Pre_center	(Intercept)	0.85*** (0.02)	0.03*** (0.00)	0.000
		Values-Tailored	-0.14 (1.00)	-0.01 (0.04)	0.892
		Information-Tailored	-1.89 (1.04)	-0.07 (0.04)	0.070
		(Intercept)	37.47*** (1.21)	0.05 (0.03)	0.151
Affective Polarization	AffPol_Pre_center	(Intercept)	0.88*** (0.03)	0.02*** (0.00)	0.000
		Values-Tailored	2.25 (1.30)	0.06 (0.03)	0.084
		Information-Tailored	-0.56 (1.33)	-0.01 (0.04)	0.675
		(Intercept)	68.98*** (1.13)	0.17*** (0.04)	0.000
2	2020 Denialism (Main)	ER0Norm_Pre_center	0.82*** (0.04)	0.03*** (0.00)	0.000
		Adversarial Statement	-2.29 (1.44)	-0.07 (0.05)	0.113
		Information-Tailored	-5.0908	-0.0055	0.013
Claim Certainty	(Intercept)	(Intercept)	84.52*** (0.70)	0.09* (0.04)	0.013

Study	DV	Term	Estimate (SE)	Std β (SE)	p-value
General Denialism		Sureness_Pre_center	0.65*** (0.04)	0.04*** (0.00)	0.000
		Adversarial Statement	0.18 (1.12)	0.01 (0.06)	0.870
		Information-Tailored	-4.43** (1.36)	-0.24** (0.07)	0.001
		(Intercept)	2.74*** (0.03)	0.08* (0.03)	0.018
		ER1_Pre_center	0.87*** (0.03)	0.91*** (0.04)	0.000
		Adversarial Statement	0.06 (0.05)	0.06 (0.05)	0.215
		Information-Tailored	-0.04 (0.04)	-0.05 (0.05)	0.312
		(Intercept)	3.02*** (0.05)	0.16*** (0.05)	0.000
		ER2_Pre_center	0.72*** (0.03)	0.70*** (0.03)	0.000
		Adversarial Statement	0.03 (0.06)	0.03 (0.06)	0.608
2020 Denialism (Aggregate)		Information-Tailored	-0.10 (0.07)	-0.09 (0.06)	0.136
		(Intercept)	59.59*** (1.07)	0.19*** (0.04)	0.000
		ER2020_Pre_center	0.78*** (0.03)	0.03*** (0.00)	0.000
		Adversarial Statement	-0.50 (1.40)	-0.02 (0.05)	0.719
		Information-Tailored	-2.84 (1.51)	-0.10 (0.06)	0.060
		(Intercept)	2.67*** (0.06)	0.08 (0.05)	0.123
		ER2024_Pre_center	0.67*** (0.04)	0.58*** (0.03)	0.000
		Adversarial Statement	0.07 (0.09)	0.06 (0.07)	0.390
		Information-Tailored	-0.10 (0.08)	-0.08 (0.07)	0.210
		(Intercept)	53.41*** (0.87)	0.10** (0.03)	0.005
2024 Denialism		ERNorm_Pre_center	0.86*** (0.03)	0.03*** (0.00)	0.000
		Adversarial Statement	1.28 (1.14)	0.05 (0.04)	0.263
		Information-Tailored	-1.82 (1.17)	-0.07 (0.05)	0.119
		(Intercept)	40.10*** (1.53)	0.11** (0.04)	0.004
		AffPol_Pre_center	0.87*** (0.03)	0.02*** (0.00)	0.000
		Adversarial Statement	-2.92 (2.17)	-0.08 (0.06)	0.180
		Information-Tailored	-3.47 (2.21)	-0.09 (0.06)	0.117

Table C4. Estimated marginal means comparisons by condition and interface

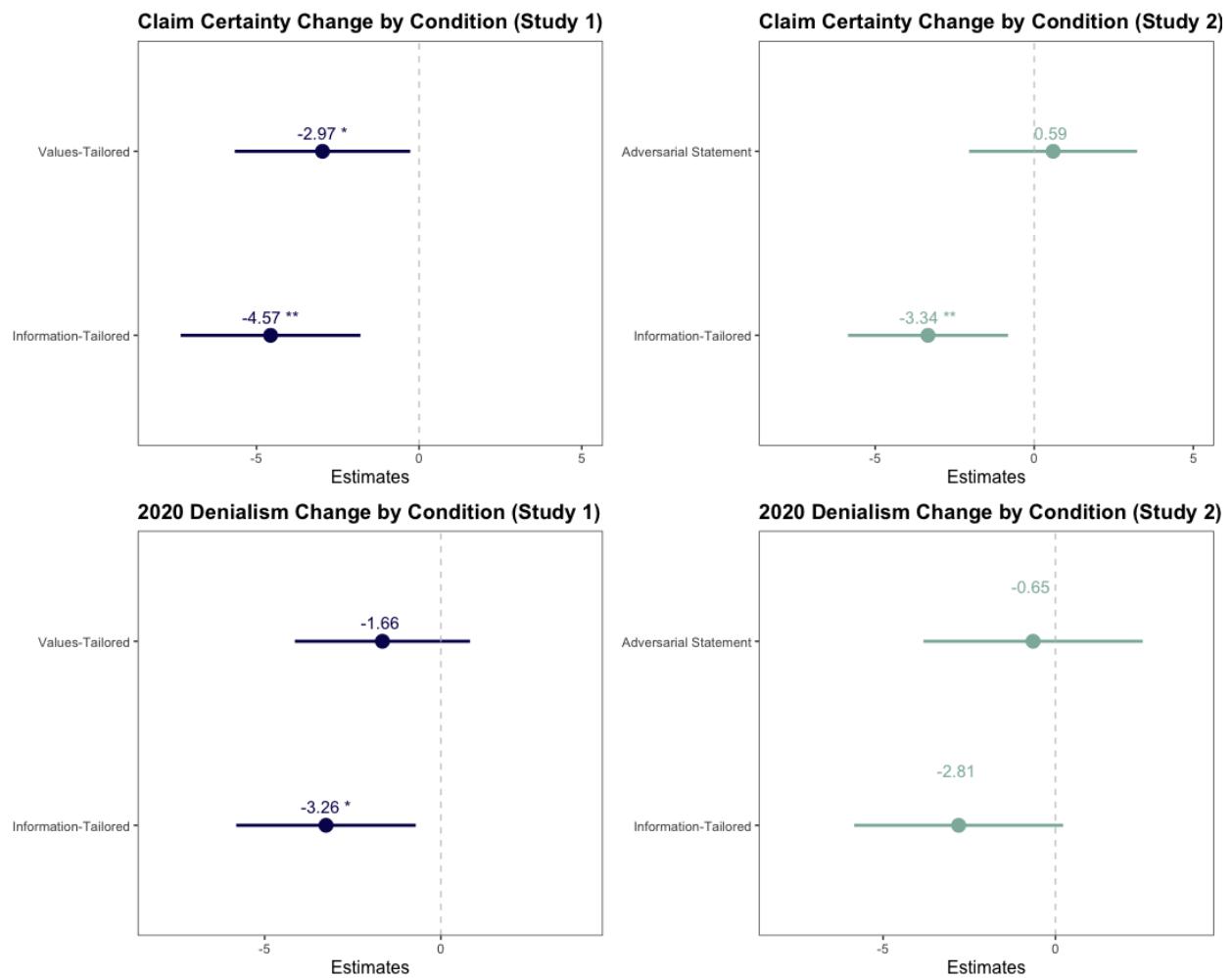
Study	DV	Comparison	Estimate (SE)	Std β (SE)	p-value
1	2020 Denialism (Main)	Adversarial Statement - (Values-Tailored)	2.41 (1.18)	0.08 (0.04)	0.104
		Adversarial Statement - (Information-Tailored)	3.73** (1.25)	0.12** (0.04)	0.008
		(Values-Tailored) - (Information-Tailored)	1.32 (1.29)	0.04 (0.04)	0.56
	Claim Certainty	Adversarial Statement - (Values-Tailored)	2.15 (1.23)	0.12 (0.07)	0.188
		Adversarial Statement - (Information-Tailored)	4.12** (1.32)	0.22** (0.07)	0.005
		(Values-Tailored) - (Information-Tailored)	1.97 (1.41)	0.11 (0.08)	0.343
2	2020 Denialism (Main)	CtlCatsDogs - Adversarial Statement	2.29 (1.44)	0.07 (0.05)	0.252
		CtlCatsDogs - (Information-Tailored)	3.56* (1.43)	0.11* (0.05)	0.034
		Adversarial Statement - (Information-Tailored)	1.27 (1.60)	0.04 (0.05)	0.706
	Claim Certainty	CtlCatsDogs - Adversarial Statement	-0.18 (1.12)	-0.01 (0.06)	0.985
		CtlCatsDogs - (Information-Tailored)	4.43** (1.36)	0.24** (0.07)	0.003
		Adversarial Statement - (Information-Tailored)	4.62** (1.51)	0.25** (0.08)	0.007

Fig. C1. Estimated treatment effects on election denialism outcomes (Studies 1–2)



Estimated treatment effects of AI dialogue conditions on post-intervention election denialism outcomes across Studies 1 and 2. Points represent regression coefficients, and error bars represent standard errors.

Fig. C2. Covariate-adjusted treatment effects on election denialism outcomes (Studies 1–2)



Estimated treatment effects from regression models adjusting for demographic and political covariates. Points represent regression coefficients, and error bars represent standard errors.

Table C5. Baseline-carried-forward imputation predicting post-intervention denialism

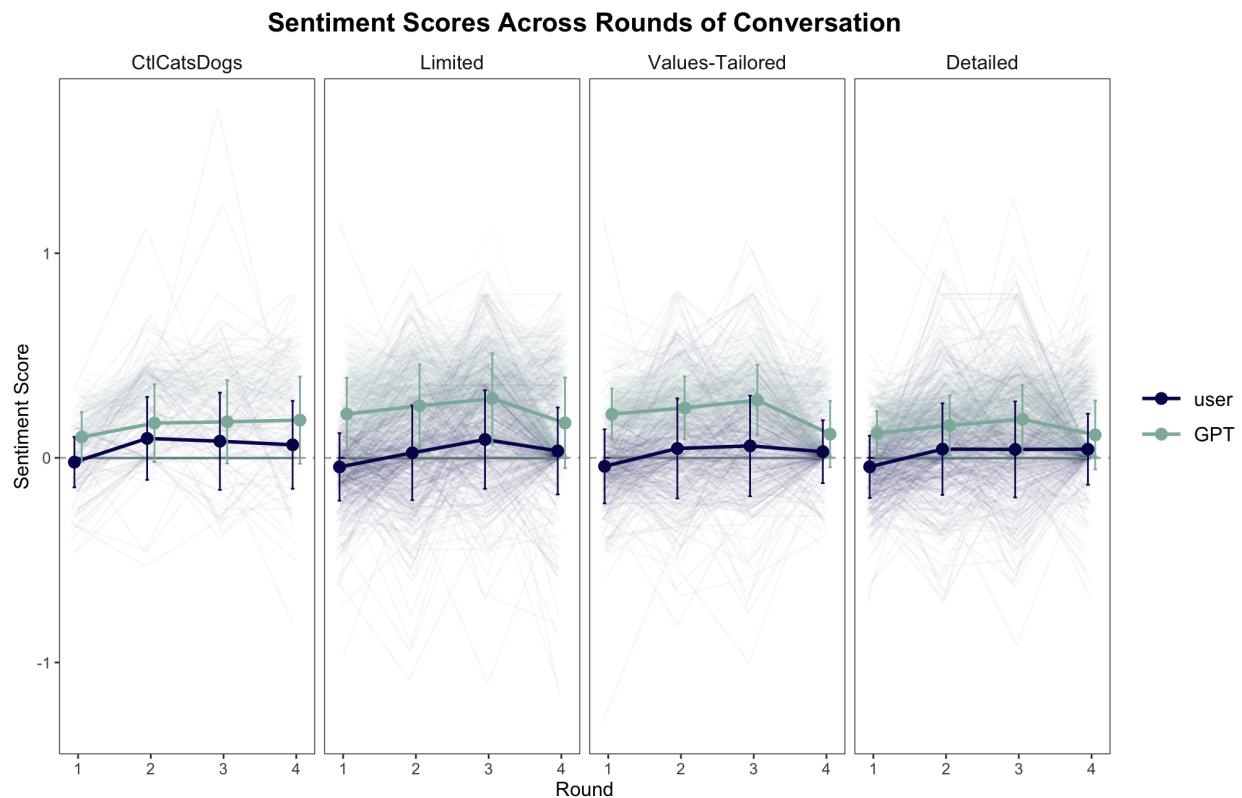
Study DV	Term	Estimate (SE)	p-value		
1 2020 Denialism (Main)	(Intercept)	67.42*** (1.00)	0.000		
	ER0Norm_Pre_center	0.76*** (0.03)	0.000		
	Values-Tailored	-2.83* (1.14)	0.013		
	Information-Tailored	-3.70** (1.21)	0.002		
	Claim Certainty	(Intercept)	81.98*** (1.01)	0.000	
	Sureness_Pre_center	0.65*** (0.04)	0.000		
	Values-Tailored	-1.93 (1.15)	0.093		
	Information-Tailored	-3.59** (1.24)	0.004		
	Affective Polarization	Information-Tailored	-0.40 (1.24)	0.750	
	2 2020 Denialism (Main)	(Intercept)	67.39*** (1.08)	0.000	
		ER0Norm_Pre_center	0.79*** (0.03)	0.000	
		Adversarial Statement	-1.49 (1.40)	0.289	
		Information-Tailored	-2.92* (1.40)	0.037	
		Claim Certainty	(Intercept)	83.92*** (0.60)	0.000
		Sureness_Pre_center	0.71*** (0.04)	0.000	
		Adversarial Statement	0.38 (1.00)	0.702	
		Information-Tailored	-3.85** (1.19)	0.001	

Table C6. Regression models predicting likelihood of voting in the 2024 elections

Characteristic	Beta	95% CI	p-value
(Intercept)	0.185	0.097, 0.274	<0.001
Condition - Feedback Type			
CtlCatsDogs	0.000	—	
Adversarial Statement	0.050	-0.046, 0.145	0.31
Information-Tailored	-0.021	-0.113, 0.071	0.65
Vote2024_Pre	0.818	0.756, 0.879	<0.001

Abbreviation: CI = Confidence Interval

Fig. C3. Sentiment trajectories across conversation rounds (Studies 1–2)



Mean sentiment scores across successive AI–participant conversation turns, aggregated by condition and study. Shaded bands represent standard errors.

Appendix D: Changes by Belief Level

2. *Change by Belief Level Supplementary Models*

For each model, post-treatment belief was predicted using the experimental condition variable, a smooth term for pre-treatment specific beliefs, and an interaction GAM that allowed the smooth effect to vary by experimental condition.

For Study 1, the smoothed interaction term indicated that all experimental conditions had a primarily linear, non-significant effect on post-intervention **2020 Denialism** beliefs. For **Claim Certainty**, limited feedback was non-linear, values-tailored was near-linear, information-tailored feedback had no meaningful interaction effect.

For Study 2, we saw a similar trend, with all interactions being non-significant with varying degrees of linearity. Overall, the GAM models showed that pre-treatment denialism beliefs strongly predicted post-treatment beliefs, but the effectiveness of feedback conditions did not depend on participants' initial belief levels.

Together, the results indicated that the relationship between baseline levels and experimental conditions was primarily non-significant and mostly linear. Trend-wise, lower baseline levels of denialism resulted in an increase in denialism scores post-intervention (or a negative effect). However, for the majority of participants who had higher denialism scores, all treatment conditions resulted in lower denialism scores.

Table D1. Generalized additive models of post-intervention outcomes (Study 1)

<i>Predictors</i>	1: 2020 Denialism (Main)		2: Claim Certainty	
	<i>Estimates</i>	<i>Statistic</i>	<i>Estimates</i>	<i>Statistic</i>
(Intercept)	79.08 **	91.13	85.91 ***	93.69
Condition - Feedback Type: Values-Tailored	-2.28	-1.86	-2.14	-1.66
Condition - Feedback Type: Information-Tailored	-3.61 **	-2.90	-3.87 **	-2.94
s(ER0Norm_Pre_center)	NA ***	52.82		
s(ER0Norm_Pre_center):Adversarial Statement	NA	1.92		
s(ER0Norm_Pre_center):Values-Tailored	NA	2.15		
s(ER0Norm_Pre_center):Information-Tailored	NA	0.38		
s(Sureness_Pre_center)			NA ***	23.35
s(Sureness_Pre_center):Adversarial Statement			NA	0.51
s(Sureness_Pre_center):Values-Tailored			NA	1.22
s(Sureness_Pre_center):Information-Tailored			NA	0.00
Observations	1081		1081	
R ²	0.455		0.219	

p*<0.05 *p*<0.01 ****p*<0.001

Table D2. Generalized additive models of post-intervention outcomes (Study 2)

<i>Predictors</i>	1: 2020 Denialism (Main)		2: Claim Certainty	
	<i>Estimates</i>	<i>Statistic</i>	<i>Estimates</i>	<i>Statistic</i>
(Intercept)	79.78 **	76.37	87.46 ***	94.34
Condition - Feedback Type: Adversarial Statement	-2.43	-1.59	-0.12	-0.09
Condition - Feedback Type: Information-Tailored	-3.52 *	-2.40	-4.61 ***	-3.55
s(ER0Norm_Pre_center)	NA **	216.37		
s(ER0Norm_Pre_center):CtlCatsDogs	NA	0.44		
s(ER0Norm_Pre_center):Adversarial Statement	NA	1.06		
s(ER0Norm_Pre_center):Information-Tailored	NA	0.02		
s(Sureness_Pre_center)			NA **	34.86
s(Sureness_Pre_center):CtlCatsDogs			NA	0.02
s(Sureness_Pre_center):Adversarial Statement			NA	0.93
s(Sureness_Pre_center):Information-Tailored			NA	0.10
Observations	515		515	
R ²	0.547		0.419	

p*<0.05 *p*<0.01 ****p*<0.001

Table D3. Interaction models: treatment effects by baseline denialism level

Study DV	Term	Estimate (SE)	p-value
1 2020 Denialism (Main)	(Intercept)	68.31*** (1.18)	0.000
	Values-Tailored	-2.94 (1.65)	0.075
	Information-Tailored	-5.74*** (1.67)	0.001
	ER0Norm_Pre_center	0.72*** (0.04)	0.000
	ConditionChatQualtrics	1.25 (1.01)	0.216
	Values-Tailored:ER0Norm_Pre_center	0.04 (0.07)	0.516
	Information-Tailored:ER0Norm_Pre_center	0.14* (0.07)	0.030
	(Intercept)	81.80*** (1.14)	0.000
	Values-Tailored	-2.29 (1.49)	0.125
	Information-Tailored	-3.71* (1.45)	0.011
Claim Certainty	Sureness_Pre_center	0.63*** (0.07)	0.000
	ConditionChatQualtrics	1.34 (1.08)	0.213
	Values-Tailored:Sureness_Pre_center	0.02 (0.11)	0.842
	Information-Tailored:Sureness_Pre_center	-0.07 (0.11)	0.499
	(Intercept)	69.09*** (1.44)	0.000
	Adversarial Statement	-1.92 (2.12)	0.364
	Information-Tailored	-4.23 (2.26)	0.062
	ER0Norm_Pre_center	0.81*** (0.05)	0.000
	Adversarial Statement:ER0Norm_Pre_center	-0.03 (0.08)	0.759
	Information-Tailored:ER0Norm_Pre_center	0.05 (0.09)	0.551
2 2020 Denialism (Main)	(Intercept)	84.29*** (0.76)	0.000
	Adversarial Statement	0.63 (1.37)	0.647
	Information-Tailored	-4.18** (1.44)	0.004
	Sureness_Pre_center	0.70*** (0.06)	0.000
	Adversarial Statement:Sureness_Pre_center	-0.09 (0.10)	0.375
	Information-Tailored:Sureness_Pre_center	-0.06 (0.09)	0.510
	(Intercept)	84.29*** (0.76)	0.000
	Adversarial Statement	0.63 (1.37)	0.647
	Information-Tailored	-4.18** (1.44)	0.004
	Sureness_Pre_center	0.70*** (0.06)	0.000

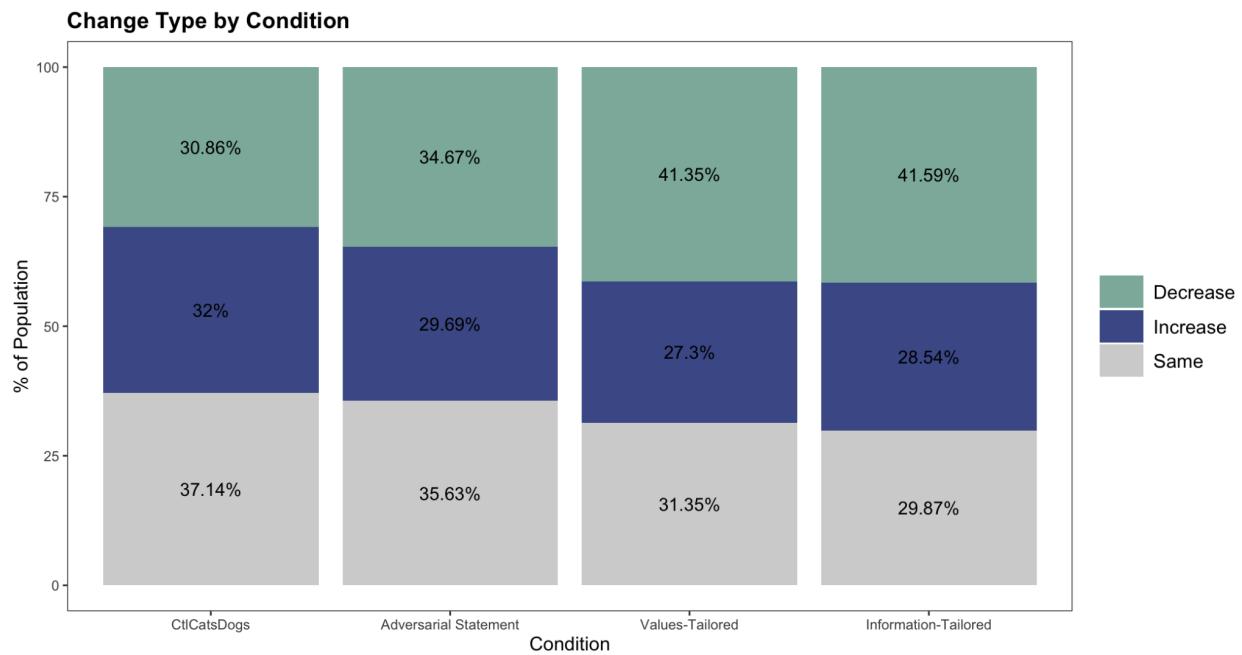
Table D4. Die-hards vs. non-die-hards: treatment effects on post-intervention outcomes

Study	DV	Term	b (SE) - Die Hards	b (SE) - Non-Die Hards	p-value
1 2020 Denialism (Main)	Claim Certainty	(Intercept)	69.95*** (1.75)	66.14*** (1.32)	0.000
		ER0Norm_Pre_center	0.81*** (0.05)	0.72*** (0.04)	0.000
		Values-Tailored	-3.29* (1.55)	-1.56 (1.61)	0.330
	Claim Certainty	Information-Tailored	-4.16* (2.11)	-3.32* (1.55)	0.032
		(Intercept)	96.46*** (1.28)	79.27*** (1.42)	0.000
		Values-Tailored	-4.49** (1.58)	0.51*** (0.06)	0.000
	Claim Certainty	Information-Tailored	-6.22** (2.14)	-0.60 (1.68)	0.720
		(Intercept)	2.80*** (0.08)	-2.76 (1.66)	0.098
		ER0Norm_Pre_center	0.79*** (0.07)	68.22*** (1.32)	0.000
2 2020 Denialism (Main)	Claim Certainty	Adversarial Statement	-1.90 (1.89)	0.79*** (0.05)	0.000
		Information-Tailored	-4.09 (2.14)	-2.56 (1.94)	0.186
		(Intercept)	97.23*** (1.10)	-3.28 (1.83)	0.074
	Claim Certainty	Adversarial Statement	1.03 (1.51)	83.71*** (0.83)	0.000
		Information-Tailored	-2.92 (1.92)	0.55*** (0.05)	0.000
		(Intercept)	2.67*** (0.09)	-0.28 (1.49)	0.850
	ER1_Pre_center	ER1_Pre_center	0.95*** (0.05)	-5.23** (1.78)	0.003
		(Intercept)	1.03 (1.51)	83.71*** (0.83)	0.000
		ER1_Pre_center	-2.92 (1.92)	0.55*** (0.05)	0.000

Table D5. Bayesian models of post-intervention denialism

model	term	Estimate	Est.Error	Q2.5	Q97.5
2020 Denialism	(Intercept)	70.049	1.513	67.071	73.062
	2020 Denialism (Pre, centered)	0.787	0.021	0.746	0.826
	Adversarial Statement	-1.775	1.381	-4.509	0.923
	Values-Tailored	-4.023	1.491	-6.986	-1.115
	Information-Tailored	-4.672	1.390	-7.425	-1.980
	Chat Condition: Chat	-0.823	0.867	-2.538	0.880
Claim Certainty	(Intercept)	84.779	1.516	81.840	87.741
	Claim Certainty (Pre, centered)	0.629	0.028	0.574	0.683
	Adversarial Statement	-1.384	1.427	-4.188	1.412
	Values-Tailored	-4.272	1.525	-7.286	-1.261
	Information-Tailored	-5.503	1.433	-8.276	-2.679
	Chat Condition: Chat	-0.347	0.908	-2.154	1.395

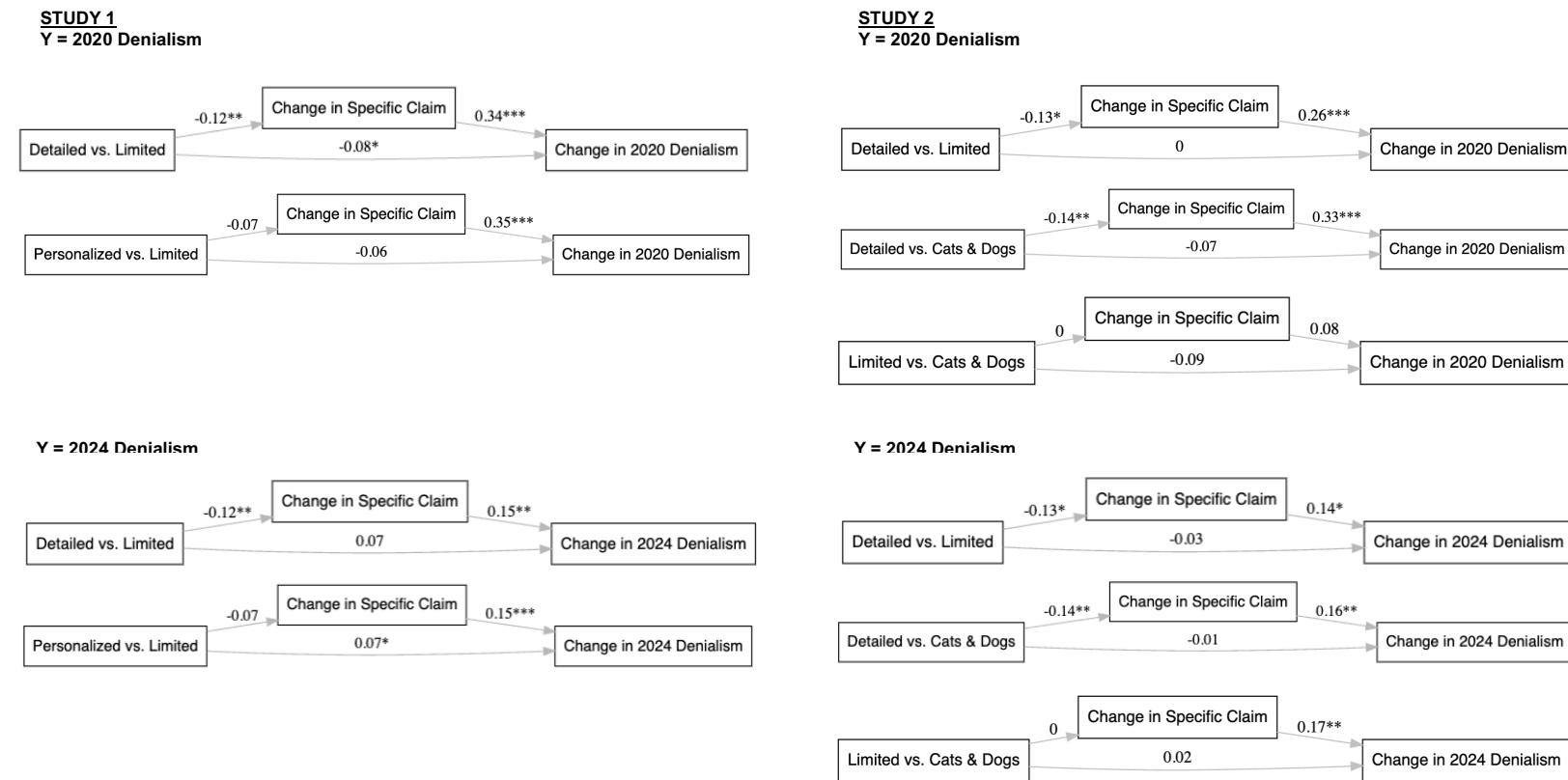
Fig. D1. Distribution of belief-change patterns by condition



Distribution of participant-level belief change patterns across experimental conditions, illustrating the proportion of participants exhibiting decreases, increases, or no change in election denialism following the intervention.

Appendix E: Mediation Models

Fig. E1. Mediation models of election denialism via claim certainty



Standardized mediation models testing whether changes in certainty about participants' articulated fraud claims mediate the relationship between AI dialogue condition and post-intervention election denialism. Path coefficients are shown with standard errors. Together, these results suggest that information-tailored feedback effectively reduced denialism through changes in certainty about specific claims, while personalized and adversarial statements may not be as effective in influencing denialism.