

Learning from Twins

Can marketers learn about consumers across the political divide by interacting with AI?

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Motivation

The Problem

Political polarization is high

- ▶ Marketers face ideologically diverse consumers
- ▶ Partisan misperceptions are widespread
- ▶ Can lead to ineffective marketing

Our approach

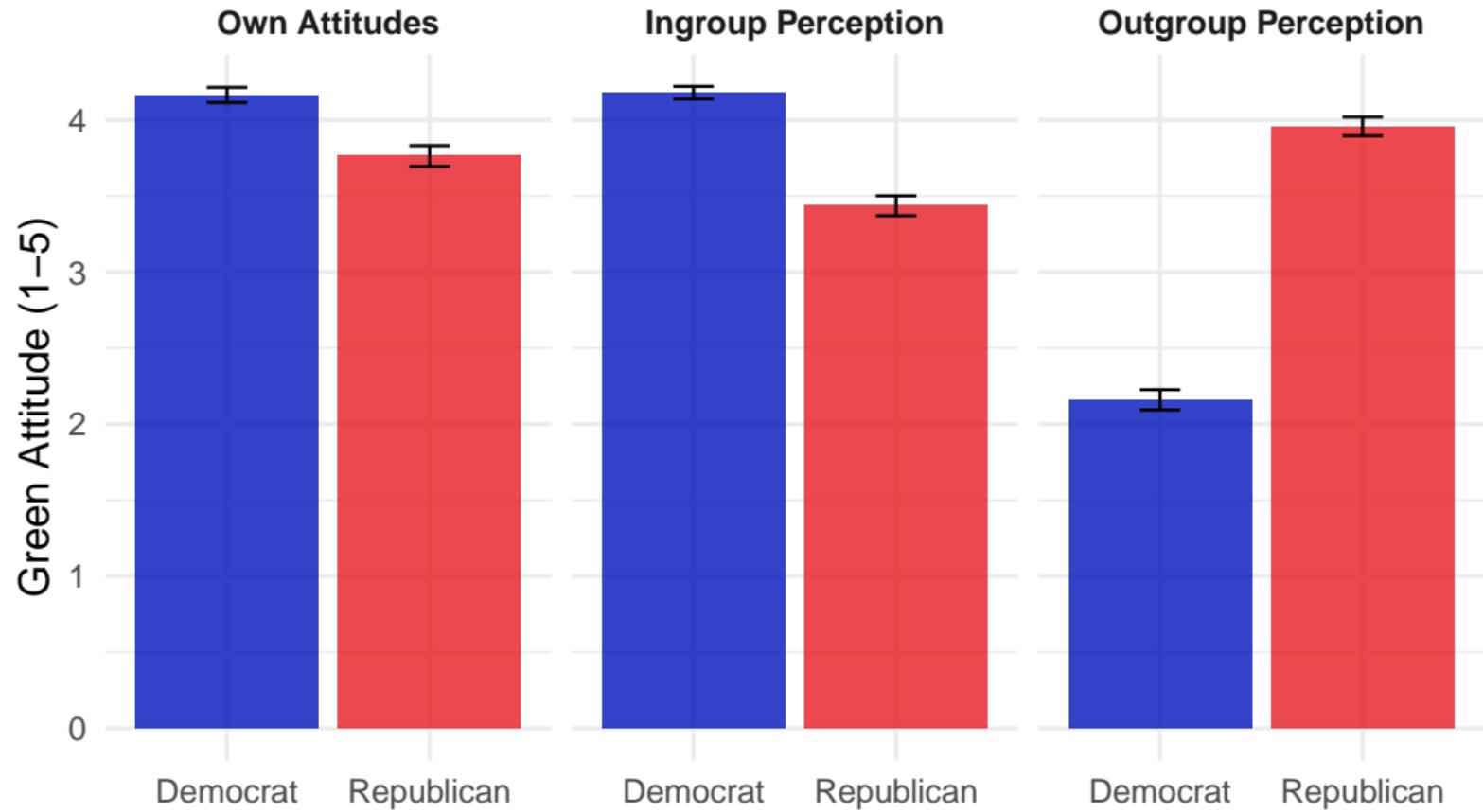
- ▶ Can AI chatbots reduce bias?
- ▶ Focus: environmental attitudes
- ▶ N = 474 participants

Research Questions

1. Do people update beliefs after chatbot interaction?
2. Is updating symmetric across political groups?
3. Do extreme partisans benefit equally?

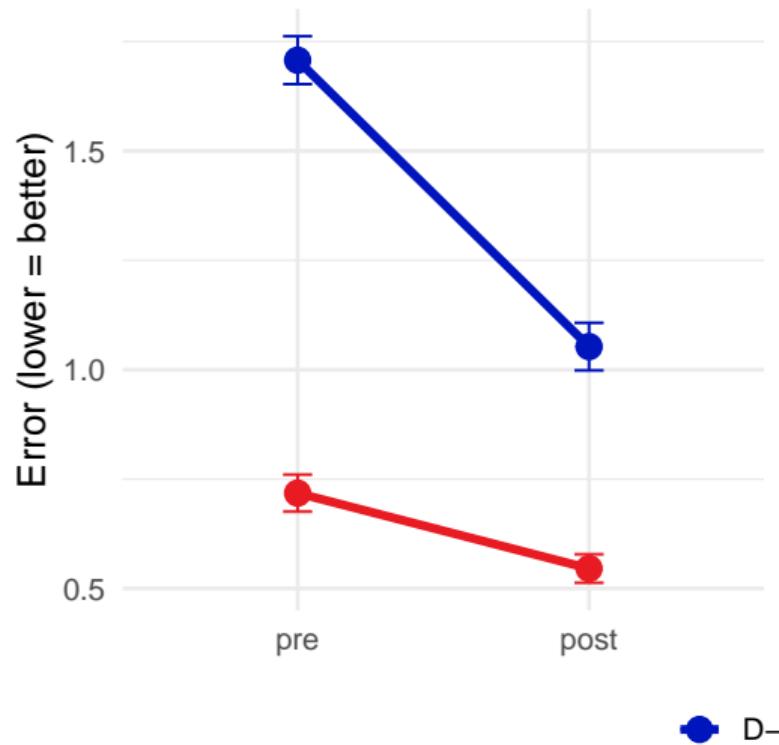
Results

Baseline Bias

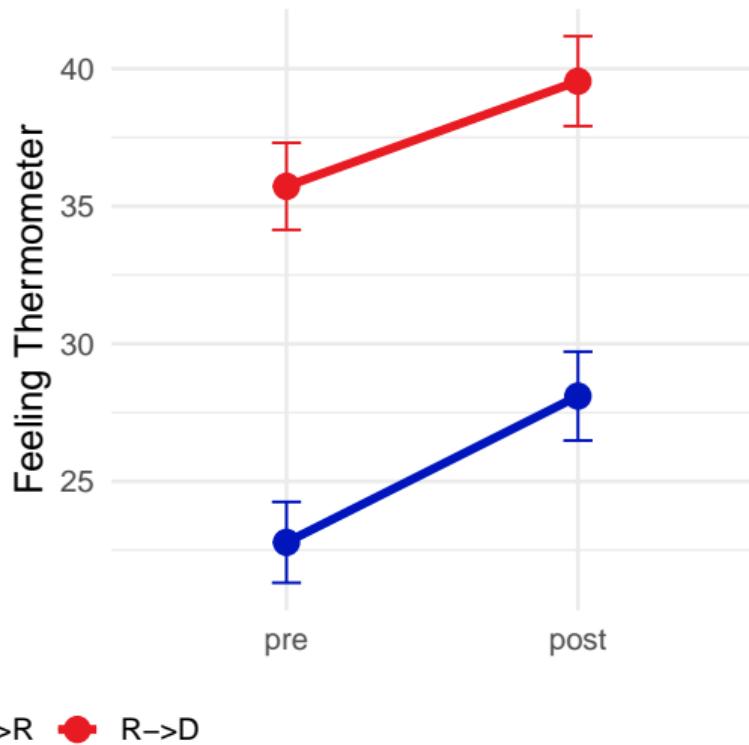


Main Effects: Pre to Post

Belief Accuracy



Outgroup Warmth



Asymmetric Effects

Accuracy

- ▶ Democrats learning about Republicans:
0.65 point improvement***
- ▶ Republicans learning about Democrats:
0.17 point improvement***
- ▶ Democrats show larger gains

Warmth

- ▶ Both groups increase warmth
- ▶ No significant difference
- ▶ 4-5 degree increase on 100-point scale

Political Extremism

Baseline: More extreme partisans have greater bias ($b = 0.013, p < .001$)

Intervention: Political extremism does NOT moderate the effect ($p = 0.152$)

Even highly polarized individuals can learn from AI

What Predicts Learning?

Bot Informativeness predicts better accuracy ($b = -0.101, p = 0.001$)

Bot Empathy does NOT predict outcomes

Engagement (word count, turns) shows mixed results

Discussion

Key Findings

1. **AI chatbots can reduce partisan bias** — participants became more accurate and warmer
2. **Effects are asymmetric** — Democrats showed larger accuracy gains
3. **Works across extremism levels** — even highly polarized individuals benefit
4. **Informativeness matters** — perceived quality of information predicts learning

Limitations & Future Directions

Domain specificity

- ▶ Environmental attitudes may differ from other domains

Temporal dynamics

- ▶ Only immediate effects; durability unknown

AI vs. reality

- ▶ Does learning from AI transfer to real people?

Need comparisons

- ▶ How does this compare to perspective-taking, information search, etc.?

Implications for Practice

Scalable tool for marketers

- ▶ Cheaper than focus groups
- ▶ Accessible way to understand diverse consumers

Particularly valuable for

- ▶ Cross-partisan messaging
- ▶ Environmental/sustainable products
- ▶ Understanding outgroup preferences

Thank You

Questions?

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Appendix

Summary: All Regressions |

	Accuracy			Warmth		
	Accuracy (ME)	Accuracy (Int)	Accuracy (Ext)	Warmth (ME)	Warmth (Int)	Warmth (Ext)
Time (Post)	-0.417*** (0.040)	-0.655*** (0.054)	-0.748*** (0.176)	4.578*** (0.551)	5.317*** (0.773)	5.435*** (1.450)
Learner Party (R→D)		-0.989*** (0.067)	-0.894*** (0.195)		12.939*** (2.228)	14.118* (6.217)
Political Extremism			0.009* (0.004)			-0.418*** (0.114)
Time × Learner Party	0.482*** (0.077)	0.777*** (0.226)			-1.496 (1.101)	-4.621* (1.865)
Time × Extremism		0.002 (0.004)				-0.003 (0.034)
Party × Extremism		-0.002 (0.005)				-0.083 (0.152)
Time × Party × Extremism		-0.008 (0.006)				0.087+ (0.046)
Num.Obs.	948	948	948	948	948	1896

+ p < 0.1, * p < 0.05, ** p < 0.01, *** p < 0.001

Summary: All Regressions II

	Accuracy			Warmth		
	Accuracy (ME)	Accuracy (Int)	Accuracy (Ext)	Warmth (ME)	Warmth (Int)	Warmth (Ext)
Time (Post)	-0.42***	-0.65***	-0.75***	4.58***	5.32***	5.43***
Learner Party (R→D)		-0.99***	-0.89***		12.94***	14.12*
Political Extremism			0.01*			-0.42***
Time × Learner Party		0.48***	0.78***		-1.50	-4.62*
Time × Extremism			0.00			0.00
Party × Extremism			0.00			-0.08
Time × Party × Extremism			-0.01			0.09+
Num.Obs.	948	948	948	948	948	1896