

# Learning from Twins

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

Benjamin Lira

2025-12-17

# Motivation

# The Problem

## **Political polarization is high**

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

## **Our approach**

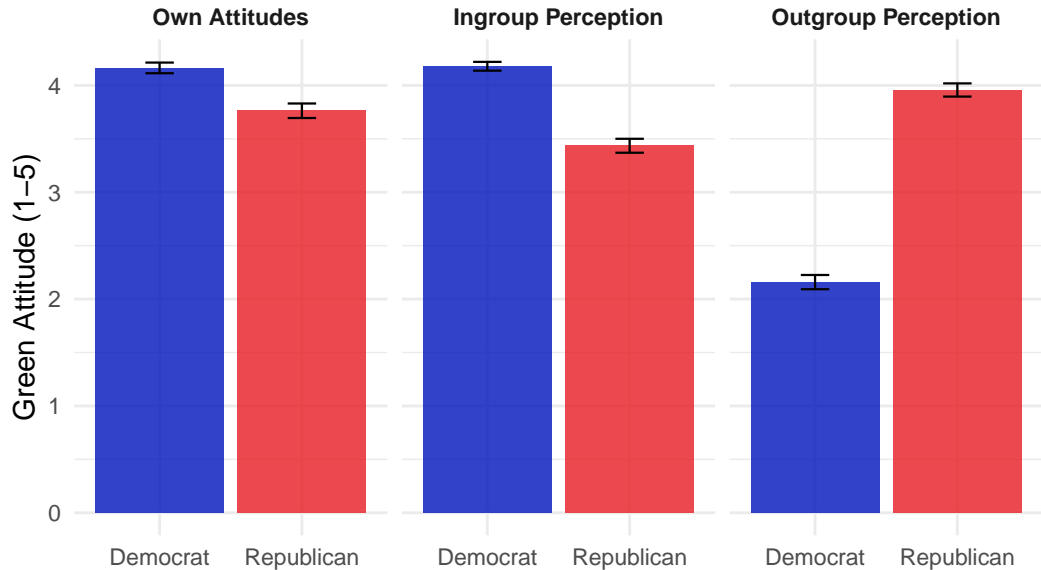
- ▶ Can AI chatbots reduce bias?
- ▶ Focus: environmental attitudes
- ▶  $N = 474$  participants

# Research Questions

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

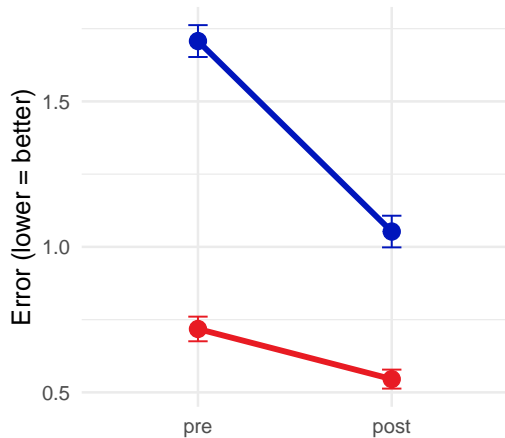
# Results

## Baseline Bias

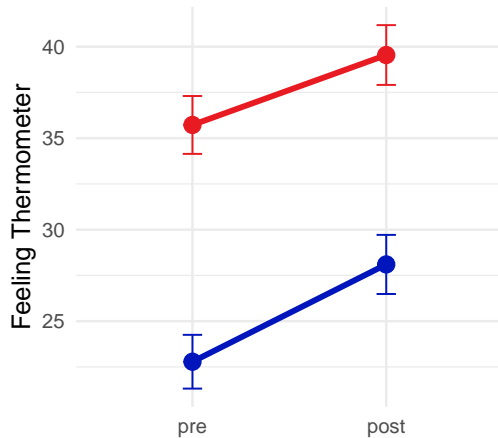


## Main Effects: Pre to Post

### Belief Accuracy



### Outgroup Warmth



● D->R ● R->D

# 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?

Benjamin Lira

`blira@wharton.upenn.edu`



# Appendix

# Summary: All Regressions I

	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