

# Survey Experiments

Business Experimentation and Causal Methods

# Revealed vs Stated Preference

- Reveal preference:
  - What people actually do.
    - Buying products, donating, voting, etc...
- Stated preference:
  - What people say they would do.

# Stated Preference Is Often Biased

- Social desirability bias
- Hypothetical bias
- Strategic Bias
- Information Bias
- Preference Uncertainty
- Lots of others

# Why do survey experiments?

- Cost
- Speed
- More control over design
- Ethical considerations
- Access to subjective perceptions.
- What you're interested may not have happened before.
  - 2024 election, self-driving cars, general AI, etc...

# Two main experiment types

- Across subjects
  - Randomization unit is the subject.
- Within subjects.
  - Randomization unit is subject by question.
  - Advantage is more power, disadvantage is spillovers.
    - Often called 'carry-over effects'.

# News from Generative Artificial Intelligence is Believed Less

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# Research Question

- Paper written prior to ChatGPT.
- What is the effect of labeling news articles as being written by AI on accuracy perceptions?
- Important question, since already some news articles were being generated.
- Various proposals to label such articles.

# Setup

- Pick true and false headlines / photos from Snopes (site that tracks disputed articles). A lot of these were about COVID.
- Main Outcome:
  - Perception of accuracy (1 - Not at all accurate, 4 - very accurate).

*News headline by experiment, experimental wave, and date of fact-checking*

Code name	Headline	Date it appeared on Snopes.com	Experiment
TRUE NEWS			
T1	Ivanka Trump Holds Variety of Trademarks in China, Including One For Coffins	14 April 2020	Experiment 1 (wave 1))
T2	Obama Urged US Pandemic Preparedness in 2014	13 April 2020	Experiment 1 (wave 1)
T3	Trump Praises China for Its 'Transparency' on COVID-19	16 April 2020	Experiment 1 (wave 1)



# Across subject design.

- A 2-cell, between-subject design.
- 3,029 participants for Experiment 1.

*Results of Experiment 1: Negative effect of AI disclosure on perceptions of news accuracy by regression specification*

	Perceptions of News Accuracy			
	(1)	(2)	(3)	(4)
<i>AI reporter condition</i>	-0.076*** (0.015)	-0.076*** (0.015)	-0.068*** (0.018)	-0.085*** (0.017)
<i>M</i>	2.56	2.56	2.72	2.41
<i>SD</i>	1.04	1.04	1.03	1.02
<i>Sample</i>	All	All	True News	False News
<i>Item FE</i>	No	Yes	Yes	Yes
<i>Observations</i>	109,068	109,068	54,534	54,534
<i>Adjusted R<sup>2</sup></i>	0.001	0.093	0.059	0.085

*Note: \*\*\*  $p < .001$*

*Table 2 displays the effect of AI disclosure (vs. human/control) on perceptions of news accuracy in Experiment 1 using linear regressions. Each observation is one participant by news item. All standard errors, reported in parentheses, are clustered by participant. Column 1 presents the baseline regression. Column 2 includes fixed effects (FE) for individual news items. Columns 3 and 4 present the treatment effects for news items that are either true (3) or false (4). These results are based on the entire dataset: we did not remove responses by those who (i) reported searching on Google (15% of the sample), (ii) reported responding randomly 22% of the sample), or (iii) failed the manipulation check (i.e., if they incorrectly recalled whether the reporter was AI or human; 18% of the sample). Statistical conclusions do not change if we restrict analysis to those who did not search on Google, did not respond randomly, or passed the manipulation check.*

# Within subject design

- In a 2-cell, within-subject design, participants saw both news items tagged as written by an AI and by a human reporter.
- 1,005 participants for Experiment 2.

Table 3

*Results of Experiment 2: Negative Effect of AI disclosure on perceptions of news accuracy by regression specification*

	Perceptions of News Accuracy			
	(1)	(2)	(3)	(4)
<i>AI reporter condition</i>	-0.145 *** (0.015)	-0.142*** (0.015)	-0.140*** (0.020)	-0.143*** (0.019)
<i>M</i>	2.62	2.62	2.71	2.52
<i>SD</i>	1.01	1.01	1.01	1.00
<i>Sample</i>	All	All	True News	False News
<i>Item FE</i>	No	Yes	Yes	Yes
<i>Observations</i>	20,120	20,120	10,060	10,060
<i>Adjusted R<sup>2</sup></i>	0.005	0.093	0.059	0.085

Note: \*\*\*  $p < .001$

Table 3 displays the treatment effect of AI disclosure (vs. human/control) on perceptions of news accuracy in Experiment 2 using linear regressions. Each observation is one participant by news item. All standard errors, reported in parentheses, are clustered by participant. Column 1 presents the baseline regression. Column 2 includes fixed effects (FE) for individual news items. Columns 3 and 4 present the treatment effects for items that are either true (3) or false (4). The results of the analysis are based on the entire dataset: we did not remove responses by those who (i) searched on Google (17% of sample) or (ii) responded randomly (18% of sample). Statistical conclusions do not change if we restrict the analyses to those who did not search on Google or responded randomly.

# Summary

- Survey experiments are useful, even if imperfect.
- Two main types:
  - Across subject.
  - Within subject.
- Example experiment:
  - People view AI generated news headlines as less accurate.