



Targeting Panelists Who Are Chronically Providing Suboptimal Responses

Interventions Leveraging Auxiliary Data

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CELEBRATING

10
YEARS

The issue of inattentive respondents

- **Online panels often face potential data quality issues due to satisficers or inattentive respondents**
 - Leads to measurement errors that are disproportionately larger within key subgroups
 - Less of an issue in probability-based panels than opt-in panels, but still a concern
- **No widely accepted industry standards for removing suboptimal responses from the data, or retiring respondents who chronically provide suboptimal responses**

Methodology

A randomized experimental intervention

- Probability of satisficing is correlated with respondent motivation
 - Is it possible to reduce suboptimal behavior by boosting respondent motivation?
- **Conducted randomized experiments via NORC's AmeriSpeak panel**
- **Study objectives:**
 1. Identify panelists who are **chronically providing suboptimal responses**
 2. Experiment with various interventions to encourage them to **more consistently provide higher-quality responses**
 - Do interventions overall improve data quality?
 - Which interventions are the most impactful?

Identifying suboptimal responses

- **Mostly identified using paradata**
- **Types of suboptimal responses that this study focuses on:**
 - **Straightlining:** Providing the same answer to each question in a grid
 - **Skipping:** Failing to answer a sufficient number of survey questions
 - **Speeding:** Taking an implausibly short time to complete a survey
 - **Other types of suboptimal responses include:** failing consistency checks, failing trap questions, or giving gibberish responses to open-ended questions

Sample for intervention

- **We identified panelists who have chronically exhibited suboptimal behavior since joining the panel**
 - Defined as **straightlining, skipping, or speeding (SSS)** on surveys
 - The rate of each panelist's SSS behavior was calculated based on all surveys they have taken; ~2,000 panelists were identified for intervention
 - Random sample was drawn, balanced on age, education level, and gender

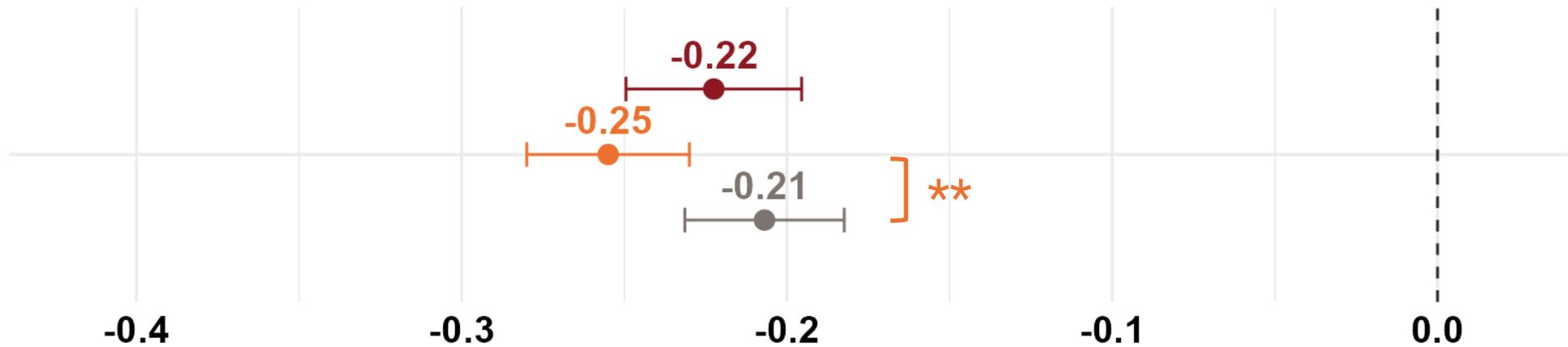
Experimental approach

- We experimented with various interventions to encourage them toward more consistently high-quality behavior:
 - 1 – A no intervention control group**
 - 2 – An email contact** describing the importance of good data, how as a panelist they can help, and how much we value them as panelists
 - 3 – A short questionnaire** subtly asking about their behavior/seeking to learn more about their survey-taking behaviors
- Examine the change in **suboptimal responses (SSS)** before and after interventions
 - Mean SSS DIFF = Mean SSS (6 Months Before Intervention) – Mean SSS (6 Months After Intervention)

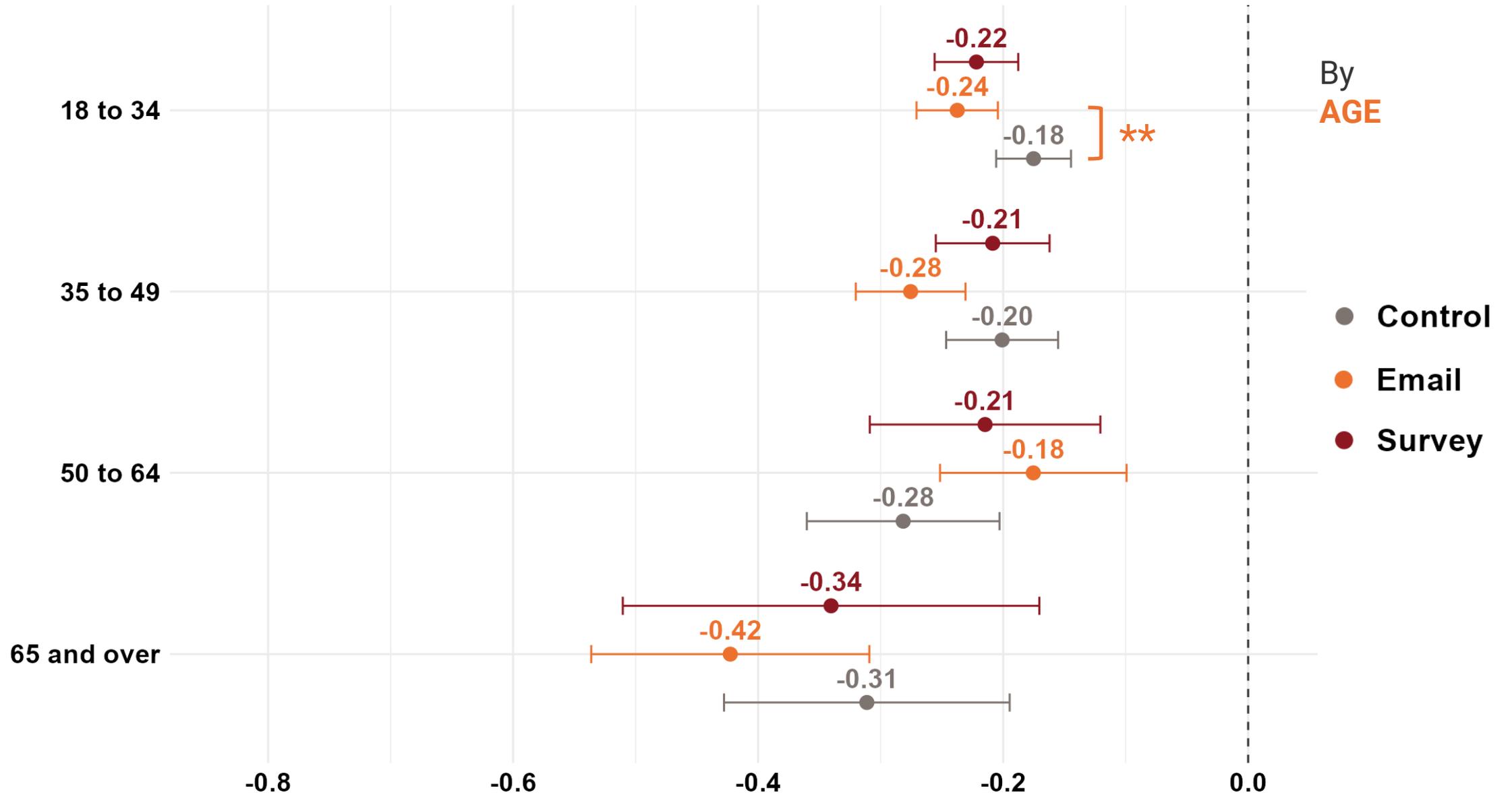
Initial Findings

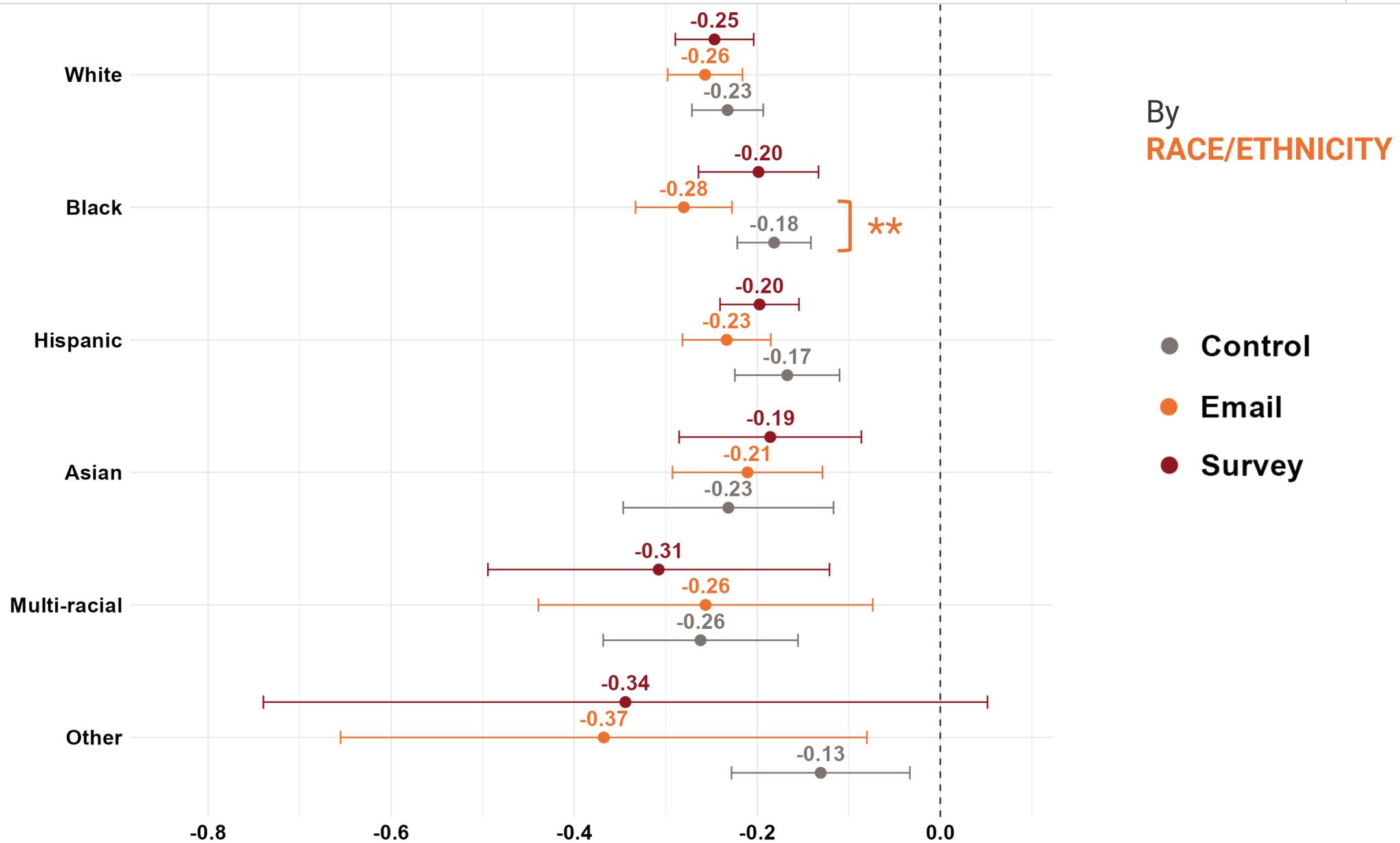
**DIFFERENCE IN MEANS:
Mean SSS (6 months Before Intervention) – Mean SSS (6 months After Intervention)**

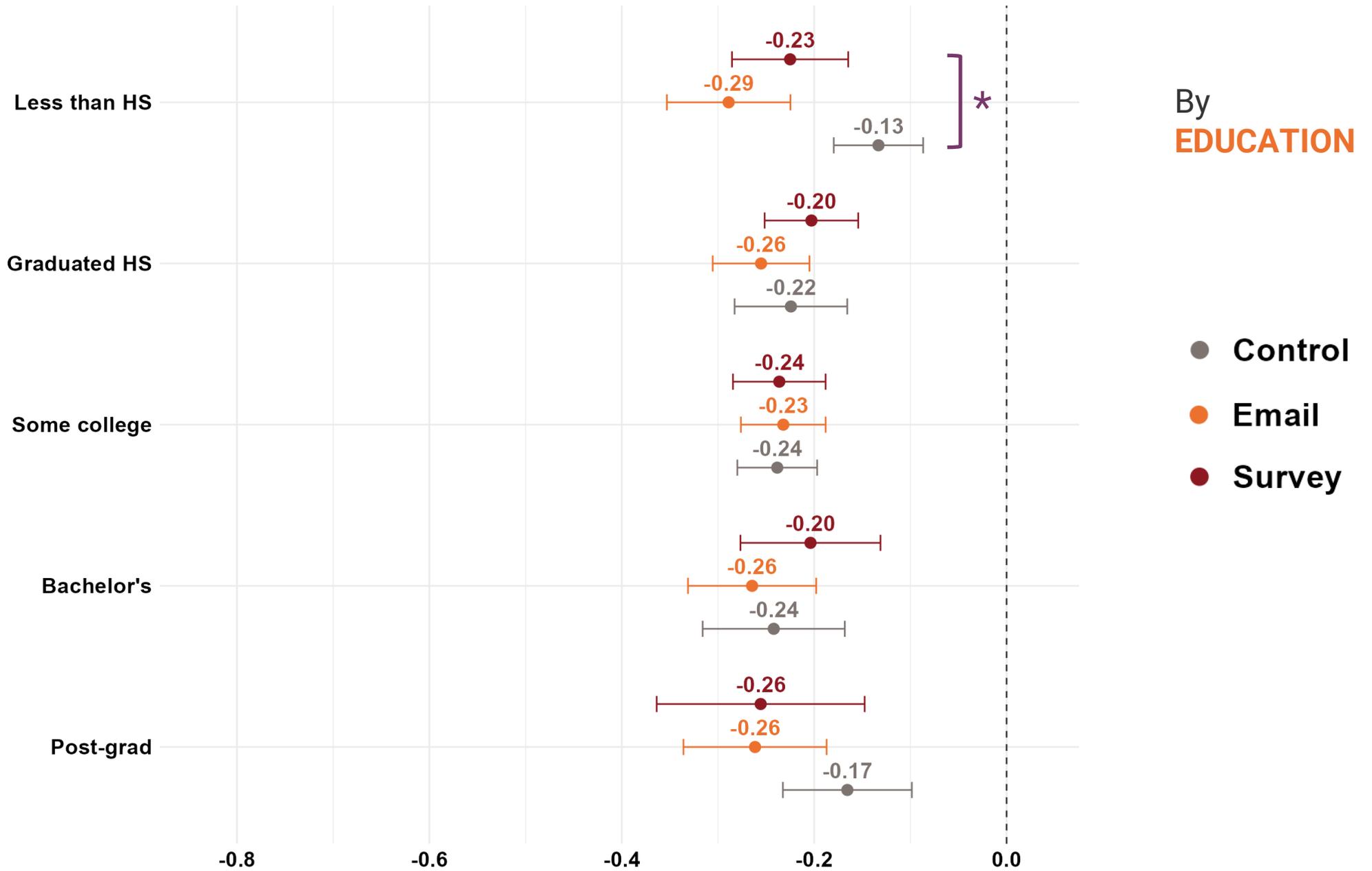
OVERALL



● Control ● Email ● Survey







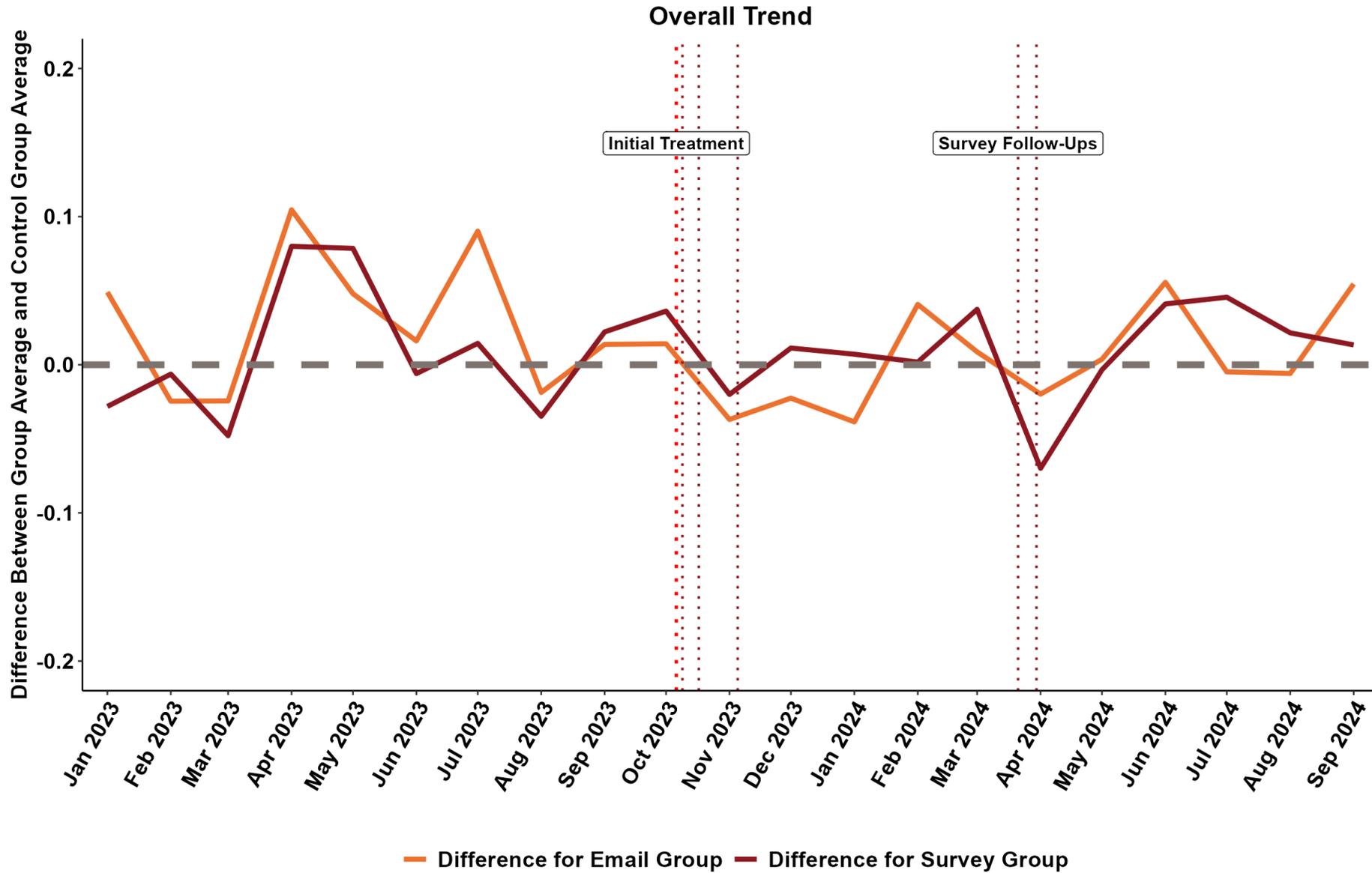
Summary

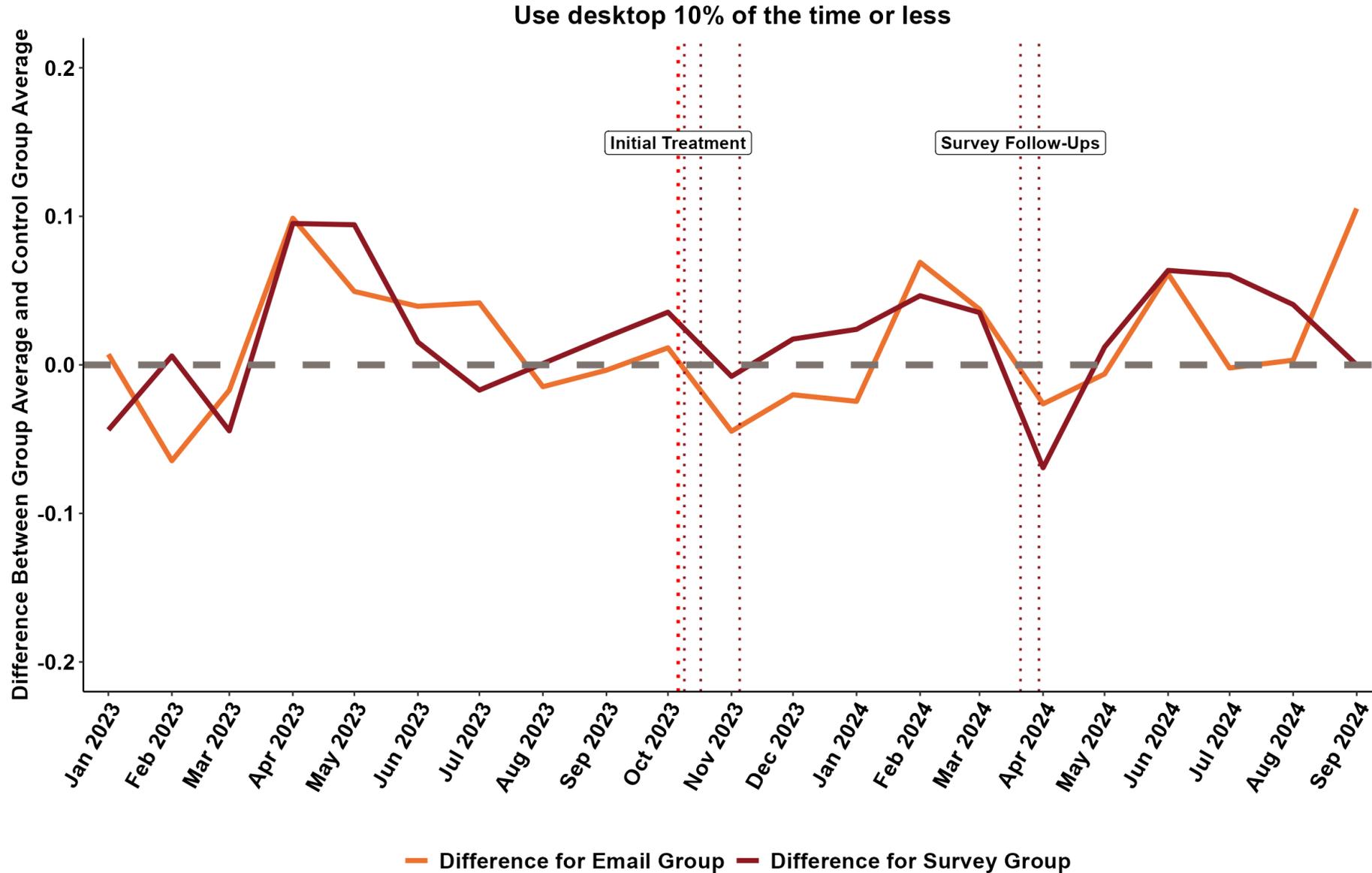
- **We found that:**
 - Overall, the **email intervention** significantly improved data quality
 - Effect was especially pronounced for subjects who were: **aged 18-34, Black, or had less than a high school level of education**
 - There was **no significant difference in survey completion rate**, indicating that the interventions did not cause attrition

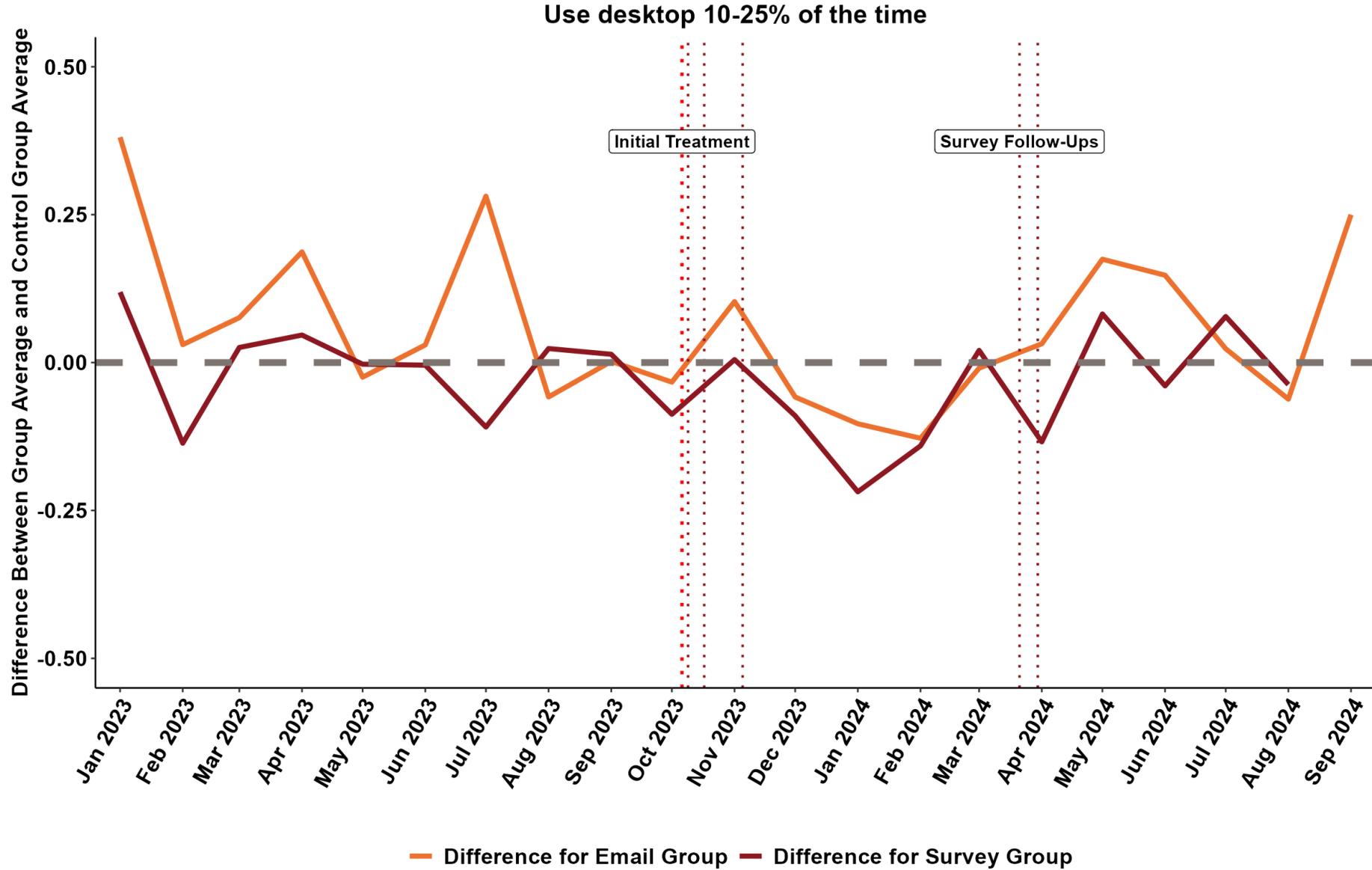
Follow-Up

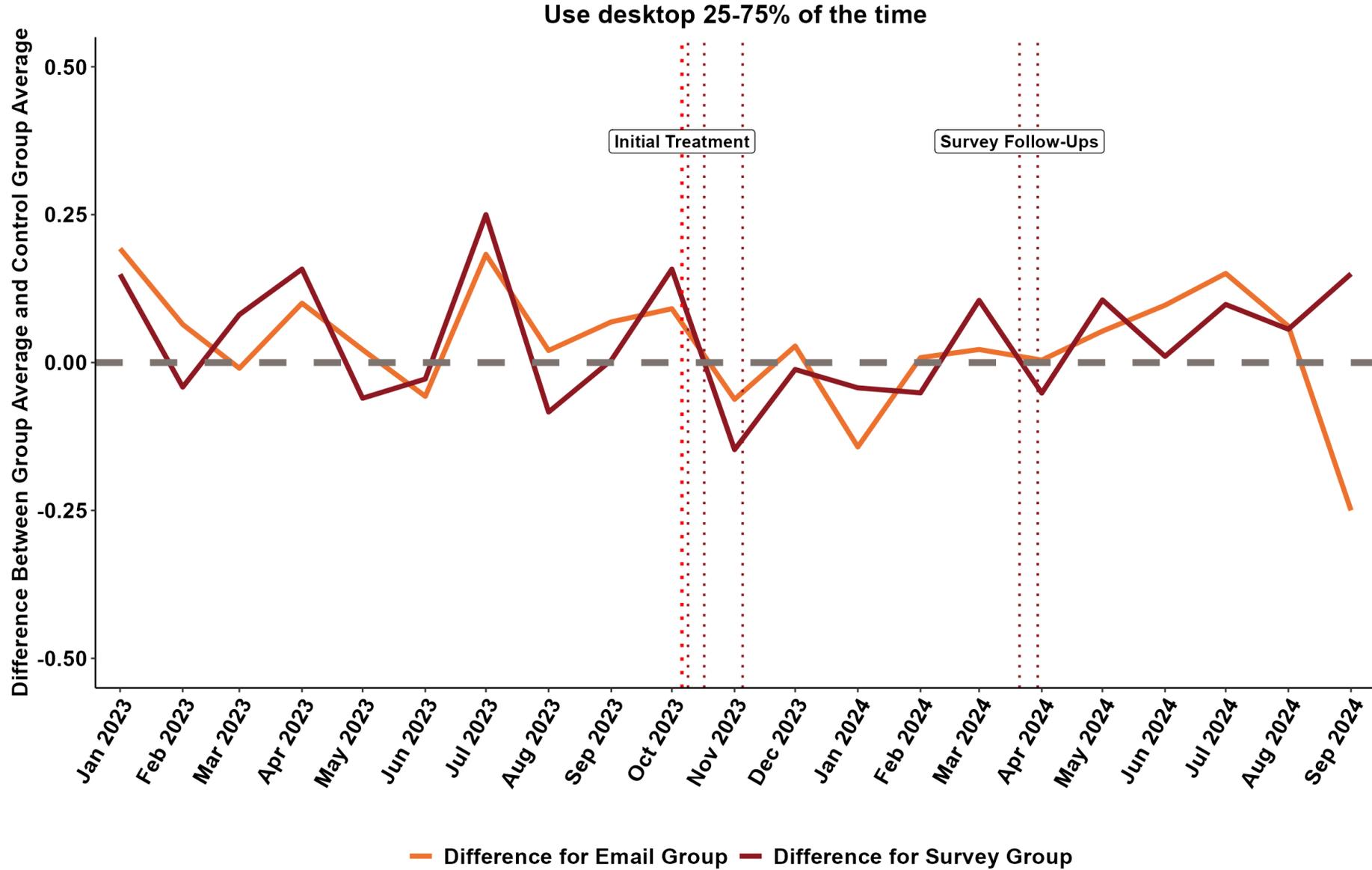
New directions

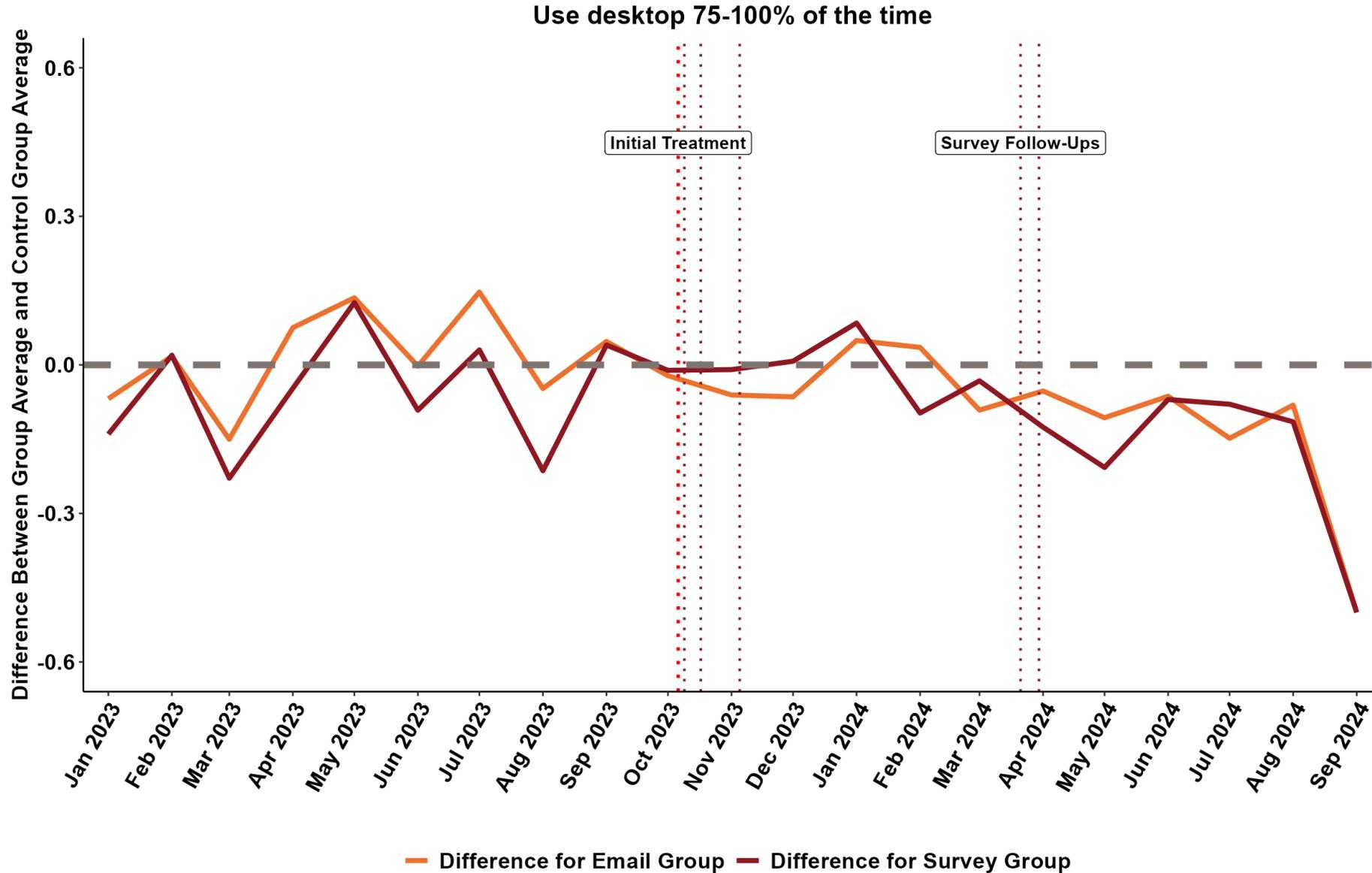
- Following up on our earlier aggregate findings, we explore:
 - How does the impact of these interventions **vary over time**?
 - How does impact **vary according to respondent's primary device type**?



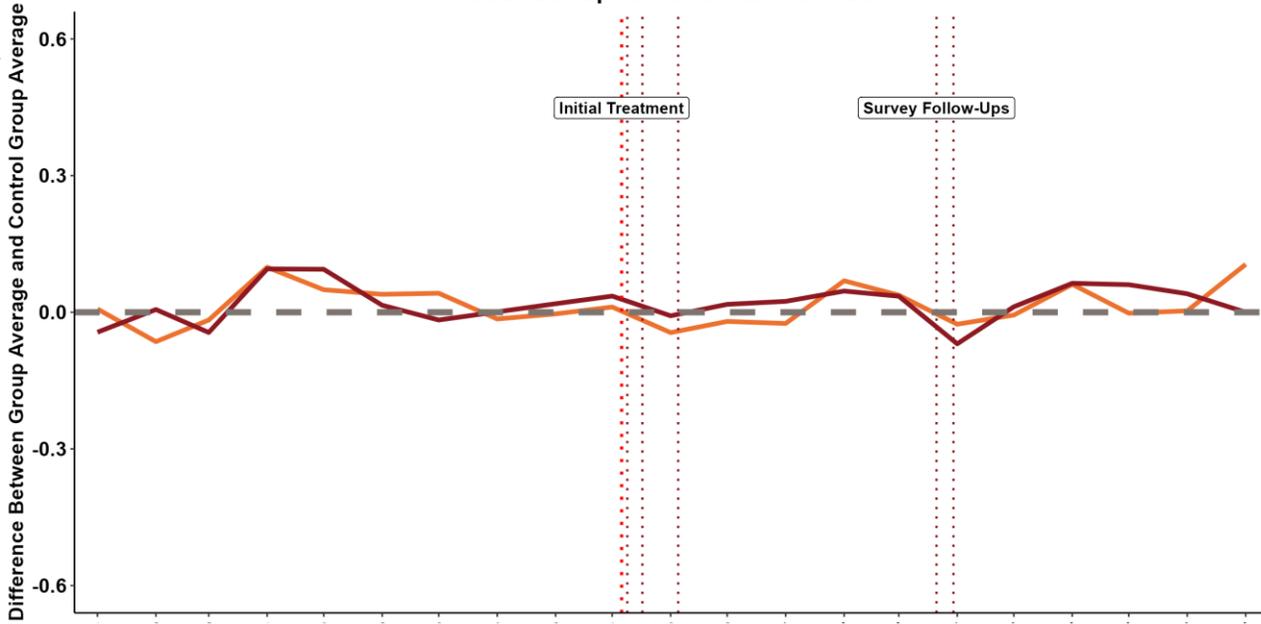




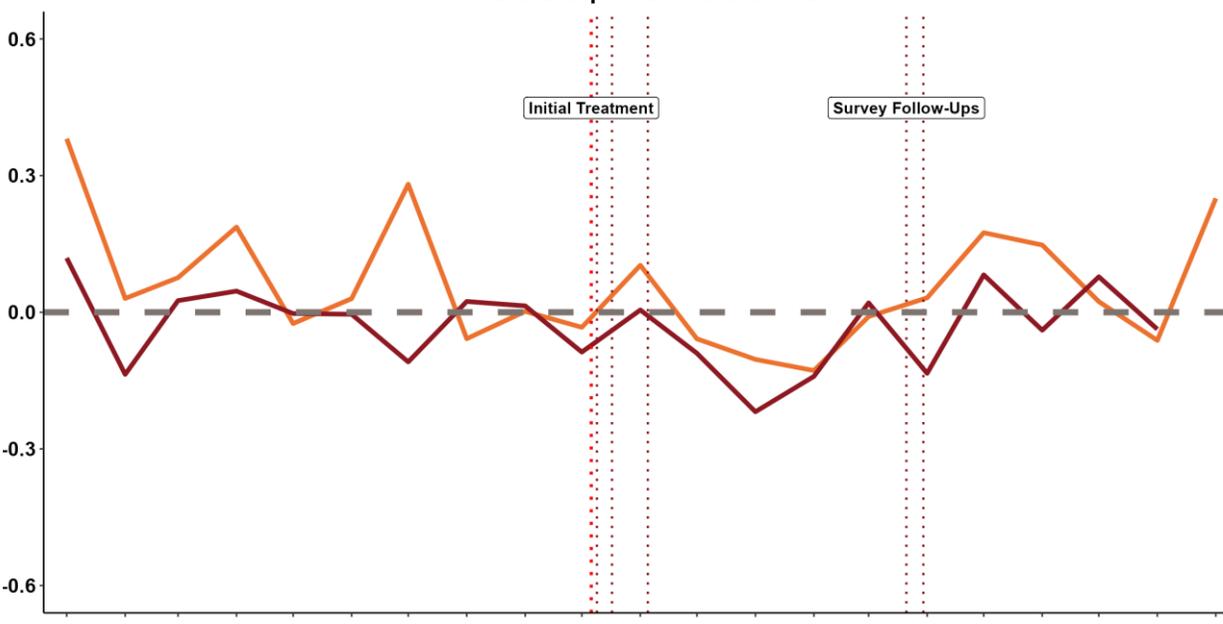




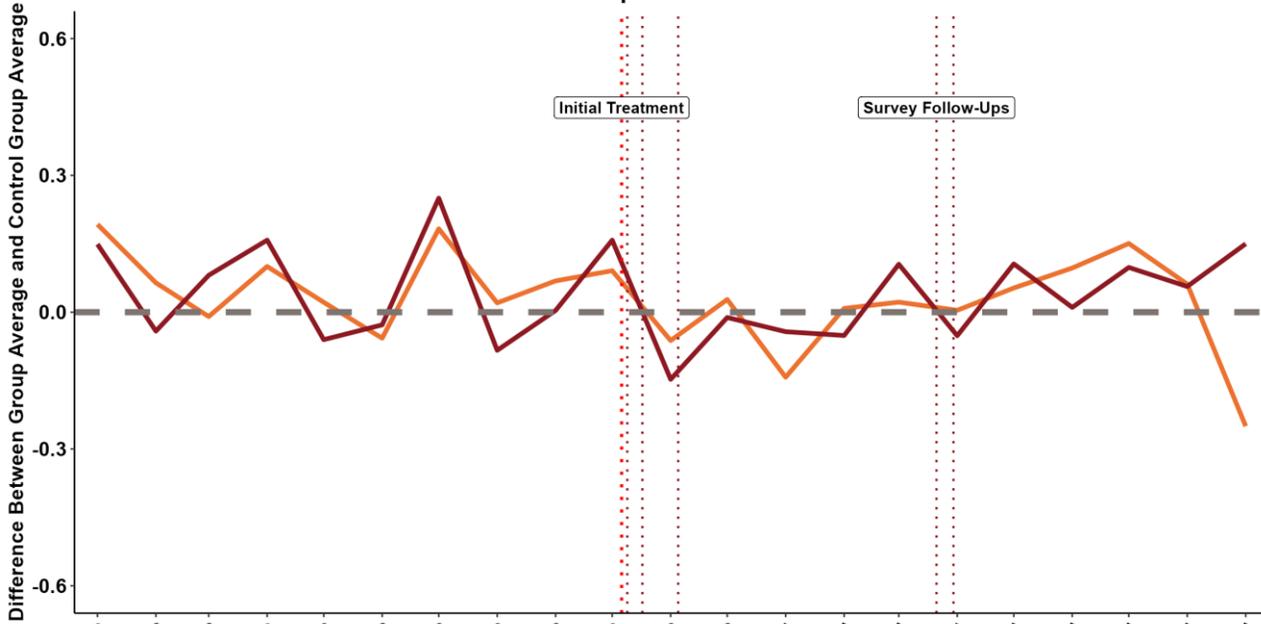
Use desktop 10% of the time or less



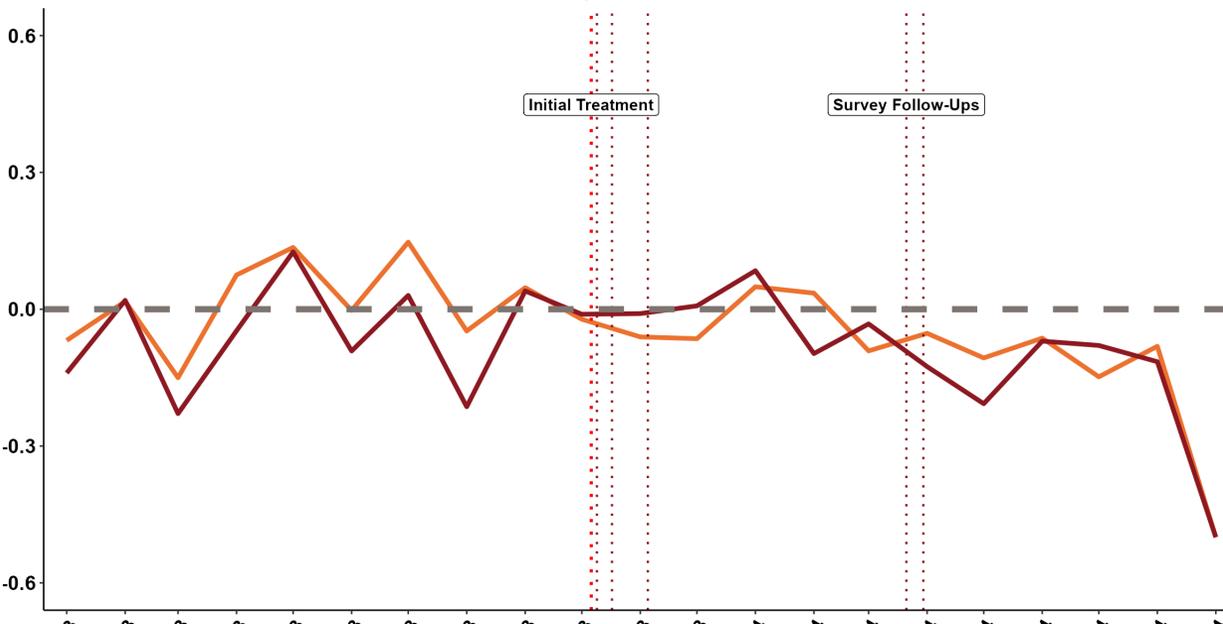
Use desktop 10-25% of the time



Use desktop 25-75% of the time

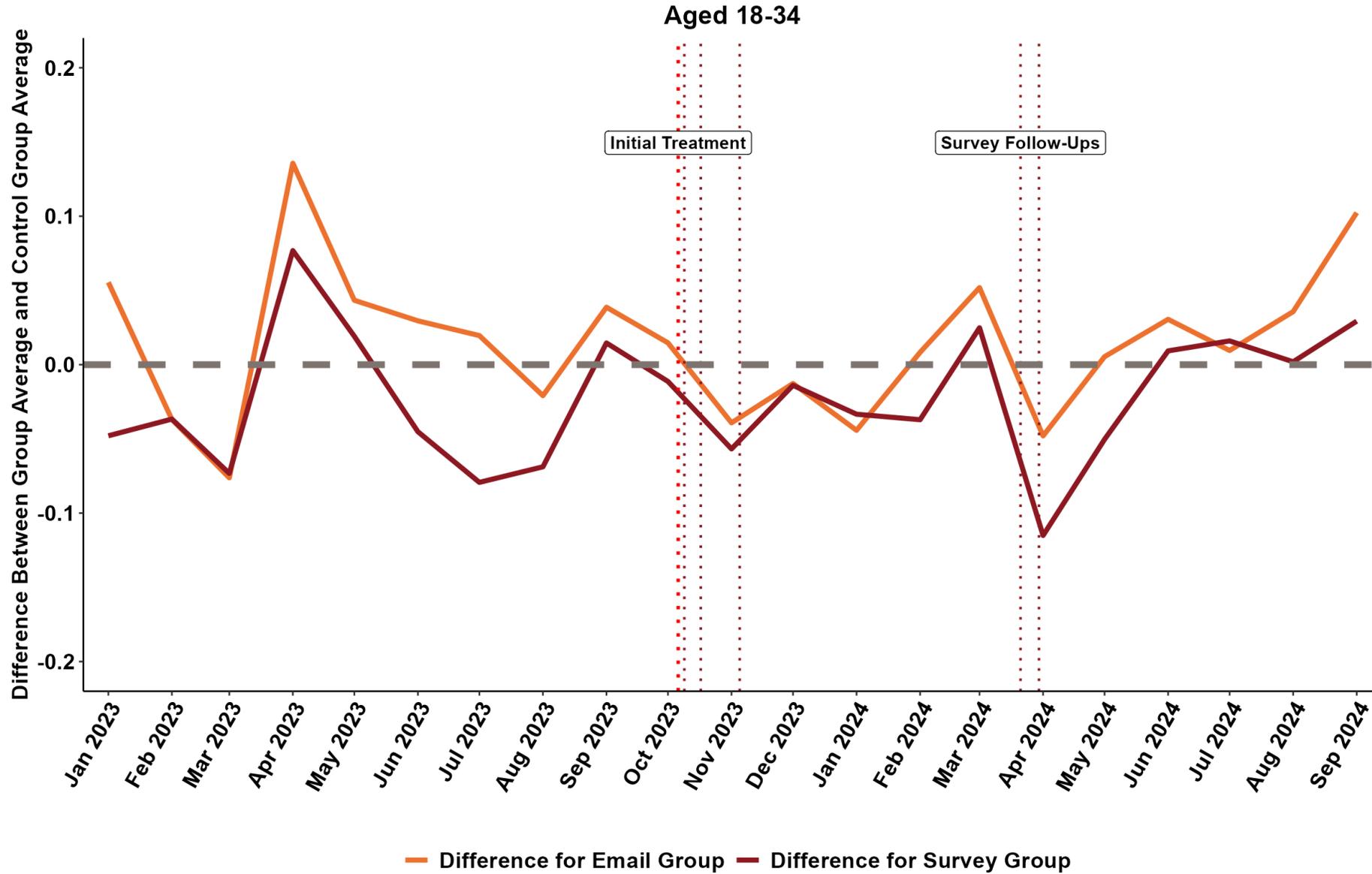


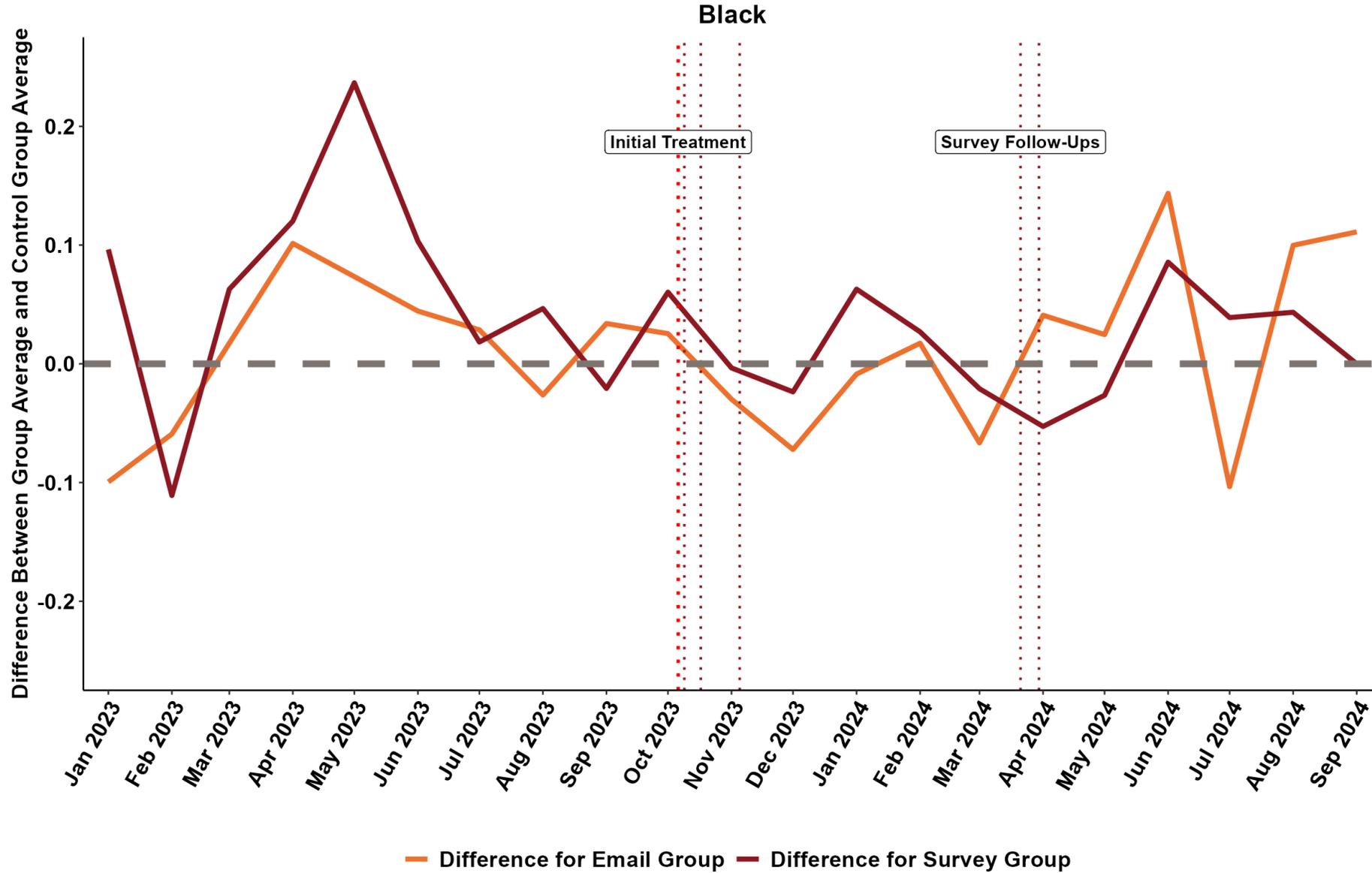
Use desktop 75-100% of the time

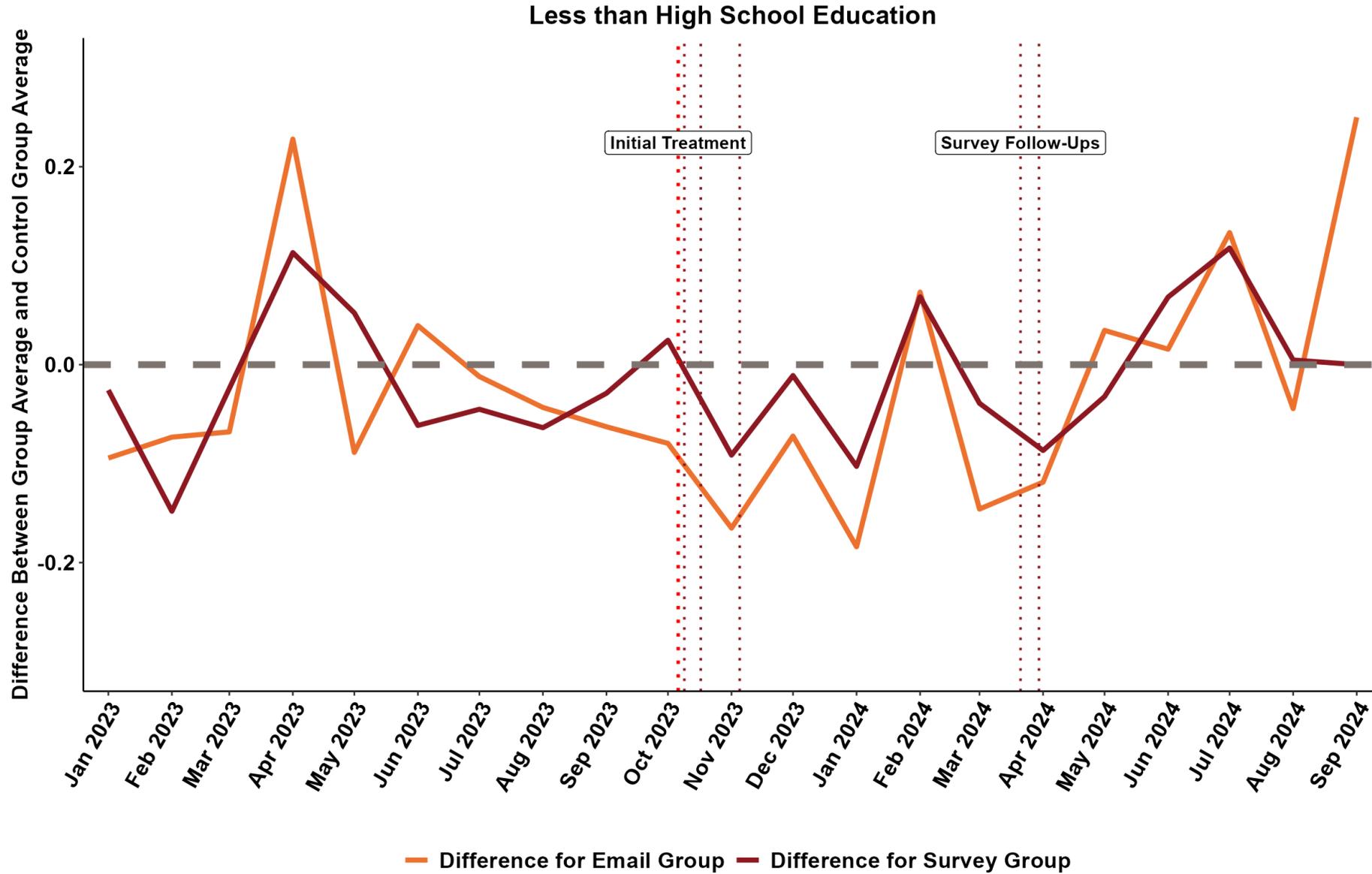


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Discussion

Key Takeaways

- Suboptimal behavior has a **strong seasonal/temporal element** across all groups
- The previously identified statistically significant treatment effects are not always visible **on a month-by-month basis**
- Treatment effect **fades over time**
- Treatment effect varies by **primary device type**

Recommendations

- Treatment effect **fades over time:**
 - **Follow-up treatments**, both in the form of survey reminders and additional emails
 - However, with more follow-ups, we must **exercise greater caution to avoid attrition**
- Treatment effect varies by **primary device type:**
 - Indicates that primary device type is a **salient respondent characteristic for exploration**
 - Consider **tailored interventions that take advantage of device-specific features**
 - e.g. administering the survey intervention through a chatbot interface may be easier on desktop

Thank you!

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Get Your Research Right



