

Digital Advertising and Market Structure: Implications for Privacy Regulation

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Motivation

- Digital advertising has rapidly taken off and become important for firms
 - >75% of ad spend in the US is now digital (eMarketer, 2024)
 - >80% of US SMBs say digital advertising is 'crucial' for their business (Statista, 2023)
- Regulation is now catching up to the industry
 - Last ten years have seen several landmark privacy and consumer data protection laws (GDPR, LGDP in Brazil, DPDPA in India, DMA, DSA,...)
 - Dramatic growth of bills that mention 'privacy' in the US
- But still relatively little research to inform these policies, let alone their implications for firms and managers

Motivation

- One important question in this space: if digital advertising becomes less effective, will there be meaningful consequences for firms and consumers?
- We focus on one piece of the puzzle, and analyze a natural experiment to see how a negative shock to digital advertising effectiveness impacts market structure

Overview: What we do

- We exploit a natural experiment afforded by the introduction of iOS 14.5 (Apple's App Tracking Transparency)
 - Substantially decreased digital advertising effectiveness (Wernerfelt et al., 2024), but exposure varied across industries.
- Use on platform data from Meta and data from the BLS to study effects on
 - (i) Advertising behavior
 - (ii) Firm count
 - (iii) Prices

Main contributions

- (1) We provide large-scale empirical evidence on the effects of digital advertising on market structure outcomes.
 - Results are consistent with a market expansion role for digital advertising: decreases in digital advertising effectiveness reduce investments in advertising, decrease the number of firms, and increase prices.
- (2) Our results also speak to the potentially large consequences of privacy protections.
 - See Dubé et al. (2024) for recent discussion

Caveats

- “The evidence strongly suggests that no single view of advertising is valid in all settings.” (Bagwell, 2007)
 - We analyze one natural experiment in one country
- We cannot make any statements about social welfare
 - Do not get into whether advertising is excessive, optimal privacy policies, etc.

Outline

- Background
- Study Design
- Data
- Main Results
- Conclude

Advertising Theory: Stealing vs. Expansion?

● Business Stealing

- \downarrow ad effectiveness \implies each firm wastes less money on advertising, profits go up, invites entry and lower prices.
- E.g., Dubois et al. (2018), Anderson et al. (2016)

● Market Expansion

- \downarrow ad effectiveness \implies market size drops, fewer firms can survive, less competition and higher prices
- E.g., Sass and Saurman (1995), Shapiro (2018)

- Ultimately, it's an empirical question which effect dominates in our context.
- In the paper we show how many of our findings can be explained by a model of market expansion.

Background: Digital Advertising Context

- Every ad campaign on Meta asks advertisers what their objective is. Why?
 - Example with Page likes
 - Referred to as delivery 'optimization'

Choose a campaign objective

- Awareness
- Traffic
- Engagement
- Leads
- App promotion
- Sales



Sales

Find people likely to purchase your product or service.

Good for:

Conversions ⓘ

Catalog sales ⓘ

Messenger and WhatsApp ⓘ

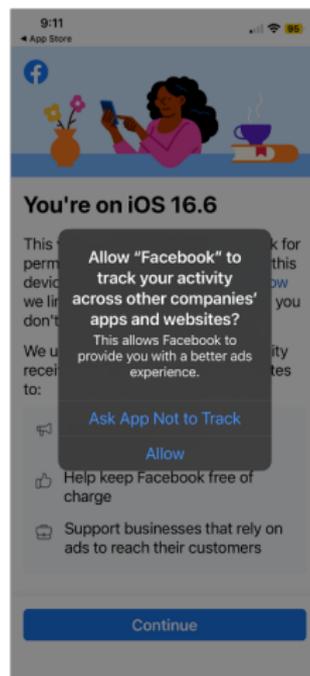
Calls ⓘ

Optimization and Offsite data

- What if you care about sales?
 - Vast majority of sales happen *offsite* (e.g., Safari, Chrome)
 - This is where tracking pixels, 3rd party cookies enter
- Note: Such optimization is very hard (if not infeasible) with TV, radio, print, etc. – arguably a major upside of digital
- Apple's App Tracking Transparency framework (described next) caused major ad platforms to lose a large share of their offsite data.

Apple's App Tracking Transparency Framework

- In April 2021, Apple launched iOS 14.5, that included their App Tracking Transparency (ATT) framework
- iOS users now get a prompt, enabling them to easily opt out of being tracked offsite
- Wernerfelt et al. (2024) show experimentally loss of this data greatly reduces advertising effectiveness (cost per incremental customer)



This was a major hit across the industry

- Google
- Twitter
- Pinterest
- TikTok
- Snapchat
- LinkedIn
- ...

The image displays three screenshots of digital marketing help centers:

- Google Ads Help:** The top screenshot shows the 'About Smart Bidding' page. It features a search bar at the top, a main heading 'About Smart Bidding', and a sub-heading 'Smart Bidding refers to bid strategies that use machine learning to optimize for conversions or conversion value in each and every auction—a feature known as "action-line bidding"'. Below this, there are several links: 'About automated bidding', 'Your guide to Smart Bidding', 'About Smart Bidding', and 'Create a portfolio bid strategy'.
- Twitter Business:** The middle screenshot shows the Twitter Business help center. It has a dark blue header with the text 'Create a website traffic campaign'. Below the header, there is a navigation bar with 'Help Center', 'General', and 'Business'. A search bar contains 'Login or Sign up', and a 'Pinterest home' link is visible on the right.
- TikTok Business Help Center:** The bottom screenshot shows the TikTok Business Help Center. It features a dark blue header with 'TikTok Business Help Center' and a 'Create an ad' button. The main content area is titled 'What's the Website Conversions Objective' and includes a 'Create a Campaign' section. A yellow banner below the main content reads 'Business Help Center' and 'How can we help?'. The bottom navigation bar includes 'Help', 'Marketing Solutions Help', a search bar, and a 'Sign in' button.

Background: Recap

- There are theories that could predict N^* and p^* going either way
- Our empirical strategy is to use the introduction of Apple's ATT framework as a natural experiment for a negative shock to advertising effectiveness
 - This caused large ad platforms to lose some of their most valuable data for ad targeting
- We will use the differential exposure across advertisers and firms as our main source of variation

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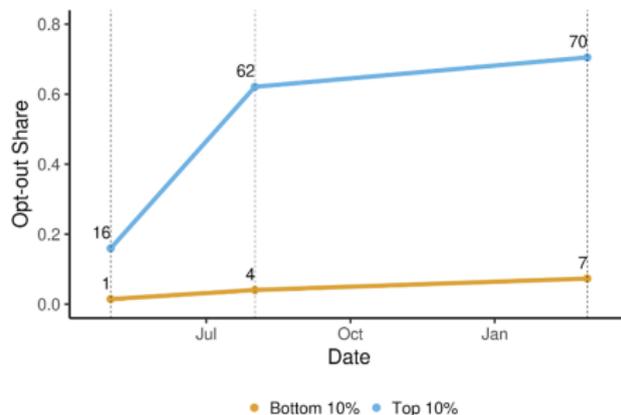
Study Design: Measure of Exposure

- We first come up with a measure of ‘exposure’ to ATT:
 - For every US advertiser on Meta, we observe how much of their ad spend on the platform pre-ATT went to users who would eventually opt out
 - Can aggregate up to industry-level exposure, letting us merge with off platform data at that level

Least Impacted Industries		Most Impacted Industries	
Industry	Impacted Share	Industry	Impacted Share
Gaming - eSports	0.305	Retail - Apparel & Accessories	0.654
Telecom - Telephone Services	0.312	CPG - Apparel & Accessories	0.632
Govt - Govt Owned Media	0.34	CPG - Baby	0.628
Orgs and Assoc - Religious	0.347	Orgs and Assocs - Arts, Heritage, & Edu	0.625
B2B - File Storage, Cloud, & Data	0.366	Retail - Sporting	0.622
Travel - Ride Sharing & Taxi	0.367	Restaurants - Coffee	0.611

Study Design: Exposure across industries

- To illustrate the variation, across advertisers there was a quite large difference in exposure from the most vs. least impacted. (Due not only to opt out rates by share of Android users)
- We measured the users' opt out behavior at two different points post-ATT launch, classifications of advertisers and industries were pretty stable.
- Main point is that there is sizable variation we're working with.



Study Design: Main specification

- We compare outcomes for the industries who were more versus less impacted
 - We focus on the top 10% vs. bottom 10% ($\text{Treat}_i = 1, 0$ resp.)
- Diff-in-diff comparing outcomes between “treated” (more impacted) and “control” (less impacted) industries i before and after the launch of ATT at time t

$$y_{it} = \beta_0 + \beta_1(\text{Treat}_i \times \text{Post}_t) + \theta_t + \lambda_i + \varepsilon_{it}$$

where y_{it} is our outcome variable of interest (e.g., # firms in industry i at time t).

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Outcome Data

- **Advertising data:** Advertiser spend, activity status, campaign details, and demographics from 1.4m US advertisers on Facebook and Instagram.
- **Prices:** BLS' Producer Price Index (PPI). PPI measures the average change over time in the selling prices for all US domestic producers of goods and services. Available at 6 digit NAICS code-month level.
- **Firm count:** BLS' Quarterly Census of Employment and Wages (QCEW). Comprehensive data based on firms eligible for state unemployment insurance. We use the 6 digit NAICS industry codes at the state-quarter level.

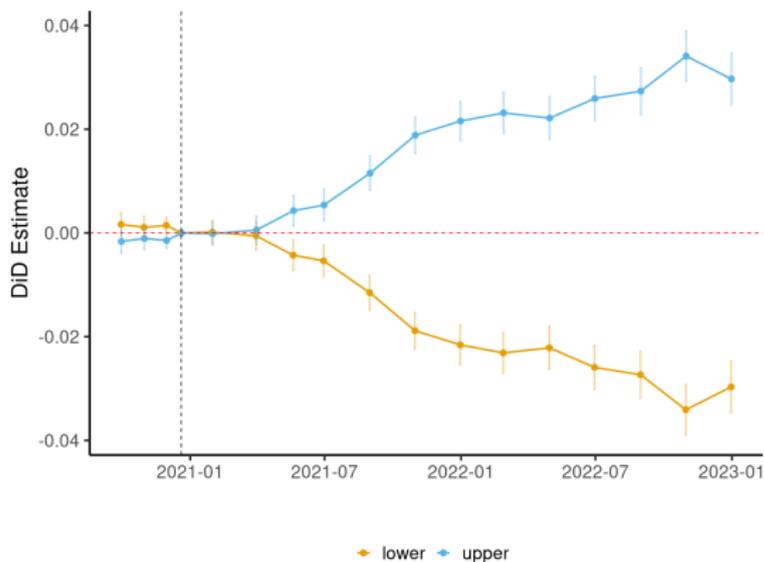
Note: needed to match Meta's internal industry definition with that from the BLS (done manually with audits, see paper).

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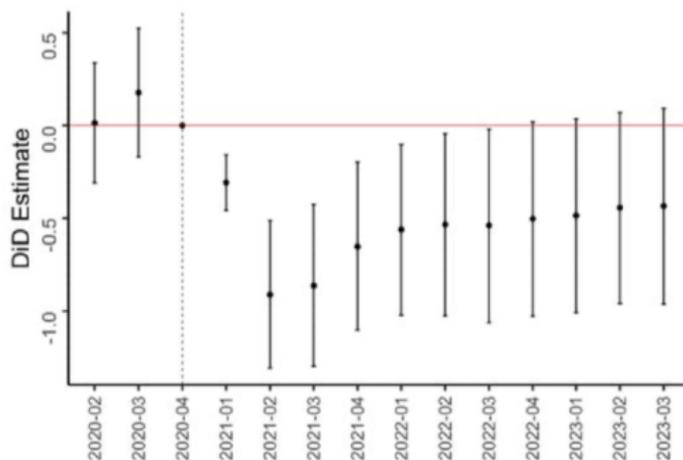
Main Results: Sanity Check

- Did ATT elicit a response from advertisers? ATT made lower funnel (not upper funnel) advertising less effective: do we see a response?
 - 1.4pp reduction in lower funnel campaigns (see also Aridor et al. (2024))



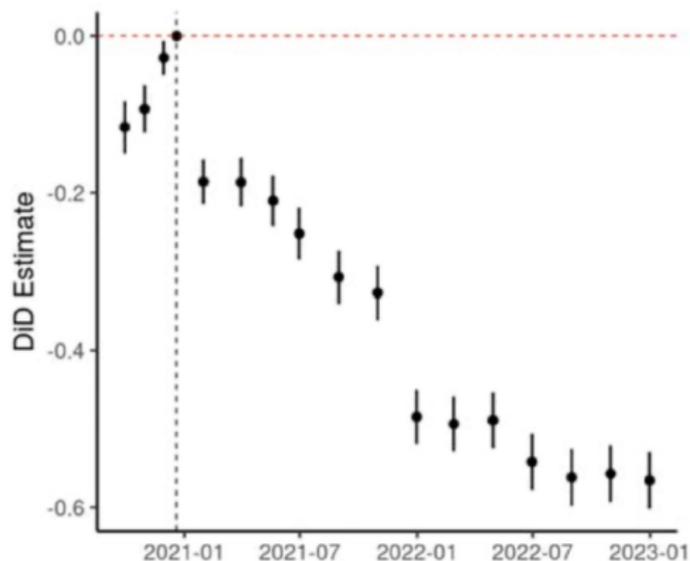
Main Results: Advertising Exit

- Net exit from advertising on Meta increased by about 50%



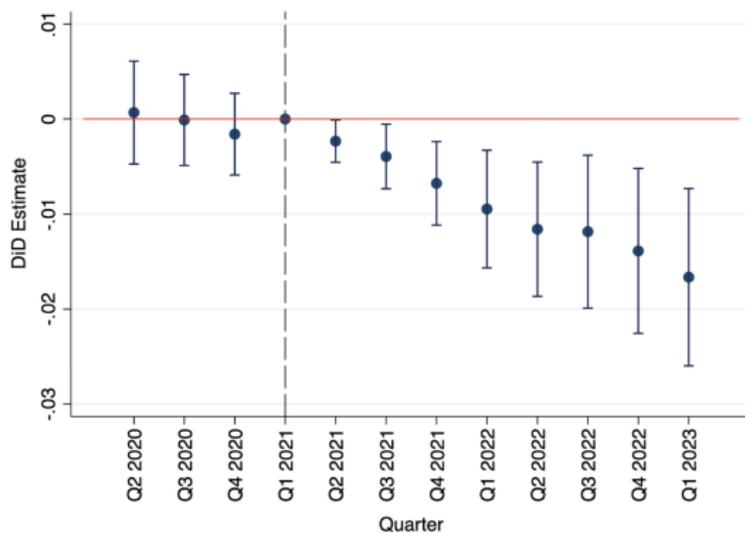
Main Results: Advertising Spend

- Advertising spend decreased by about 28%



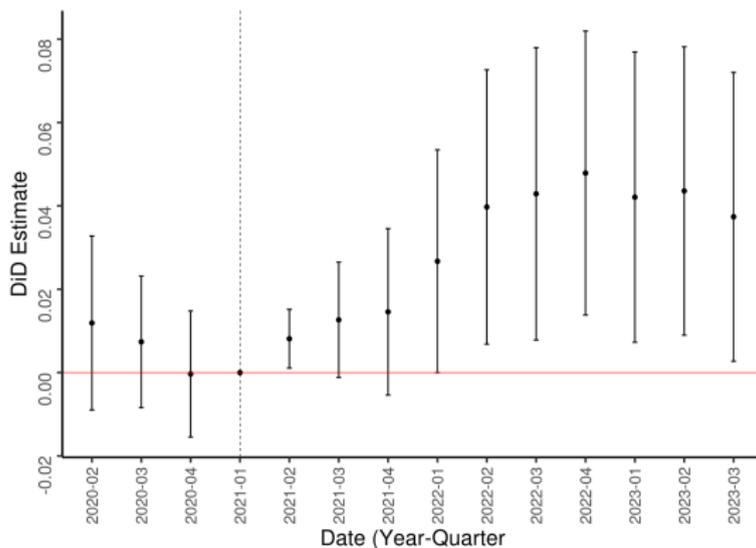
Main Results: Firm exit

- 1% decline in the total number of business establishments



Main Results: Prices

- 2.9% increase in the prices



Main Results: Recap

- We find evidence that industries that experienced greater negative shocks to digital advertising effectiveness saw *decreases in investments in advertising on Meta, reduced firm counts, and higher prices.*
- This is broadly consistent with a market expansion story ('informative' vs. 'persuasive')

Heterogeneity: 'Large' vs. 'Small' advertisers

- We define 'large' as the very big players (roughly $> \$500,000$ /year ad spend) and small as everyone else
- Consistent with past findings, the small guys get hurt more (below: Overall, Small, Large)

Panel B- Advertiser Level Analysis

Share Lower Funnel Ad Campaigns	-0.014*** (0.001)	-0.010*** (0.001)	-0.061*** (0.007)
Pr(Active)	-0.034*** (0.003)	-0.035*** (0.004)	0.056*** (0.007)
Ad Spend	-0.337*** (0.050)	-0.339*** (0.048)	0.850*** (0.150)

Conclusion

- We provide evidence consistent with a market expansion role for digital advertising
 - The magnitudes are further economically meaningful
- Similarly, we find privacy protections can have sizable unintended consequences
 - Important for regulators to weigh tradeoffs