

How Good Is Your Firm at Measuring Advertising ROI



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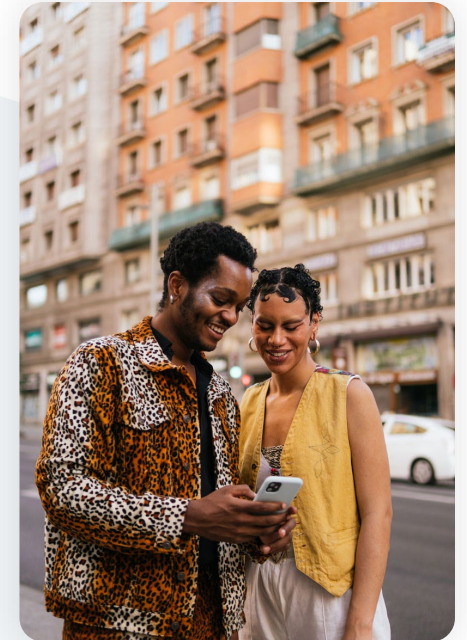
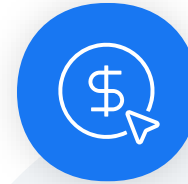
Brittany Swanson
Meta

How good is your firm at measuring
advertising ROI?

ADVERTISING EFFECTS
HAVE ALWAYS BEEN HARD TO MEASURE

“
Half the money I spend on
advertising is wasted; the trouble
is I don't know which half.”

— John Wanamaker (1838 - 1922) Department Store Merchant



How data fuels advertising is changing



Policies are giving people more options to limit how their data is shared with businesses

EXAMPLE

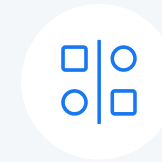
Regulatory policies, such as GDPR, CCPA, and LGPD



People are choosing to opt-out of receiving ads on websites

EXAMPLE

Evolving technologies, such as ad blockers



Platforms are removing identity and grouping the data shared with businesses

EXAMPLE

Platform solutions for new data policies, such as Apple's SKAdNetwork for iOS's ATT and Google's Privacy Sandbox Initiative

INTRODUCTION

Good measurement is increasingly difficult to achieve, yet essential to marketing success

Companies advertise to influence more people to buy their products and increase brand preference.

But when measuring the impact of ads, it's important to remember that correlation does not imply causation.

2,764
Reach

632
Purchases

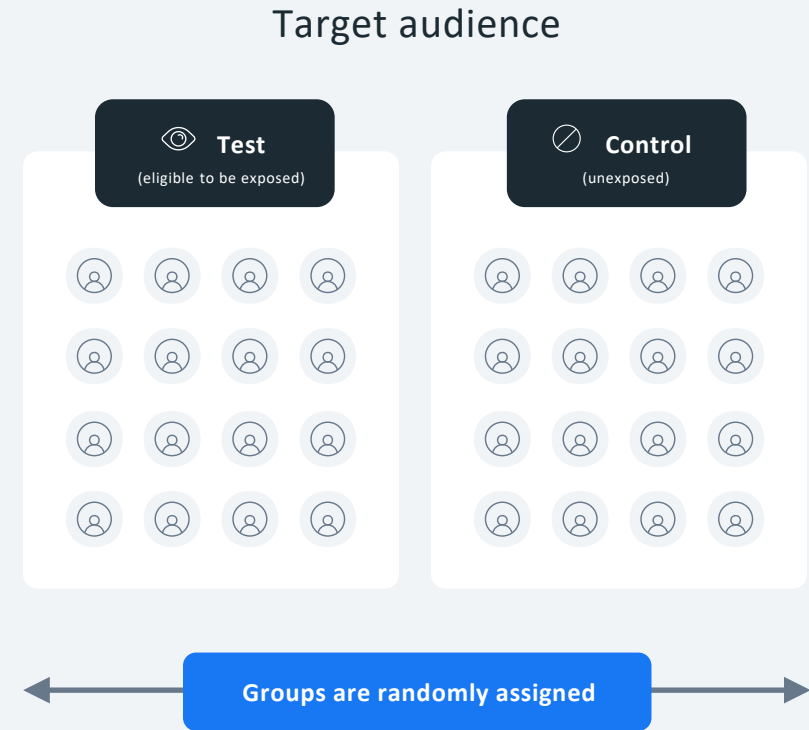
405
Views

And yet, most of the measurement tools advertisers rely on to optimize their marketing investments are measuring correlation, **not causation**

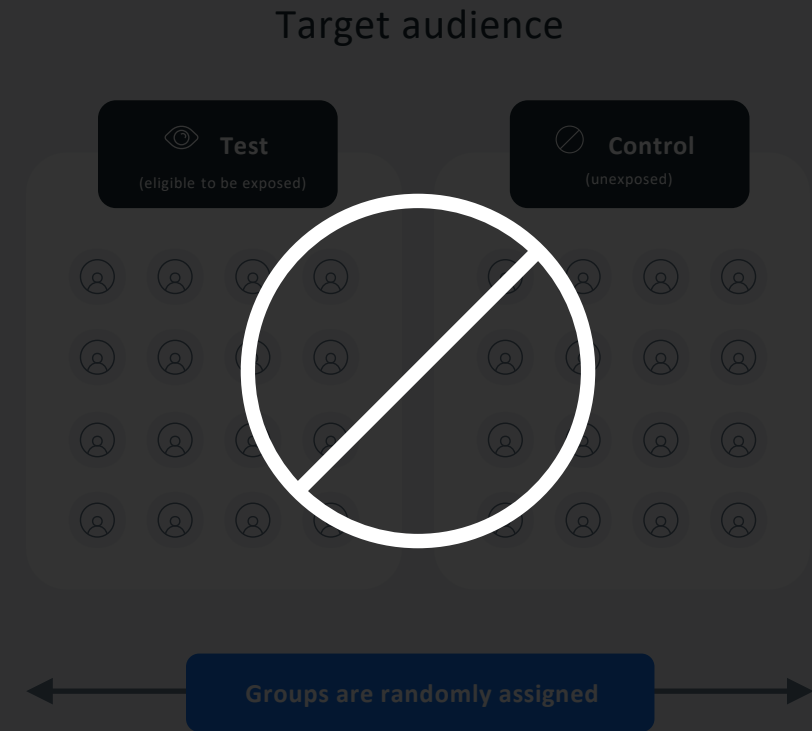


The ladder of incrementality

Randomized control trials (RCTs) are recognized as the “gold standard” to measure incremental effects



But what if you
can't run an RCT?



THE LADDER OF INCREMENTALITY

You can think of incrementality as a ladder of options that get closer to measuring true business value as you climb

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Randomized Experiments

Trials to measure the precise difference between being exposed and not being exposed to an ad campaign.



Quasi-Experiments and Incrementality Models

Techniques that estimate (but don't measure precisely) the incremental effect of being exposed to an ad campaign.



Non-Incremental Models

Systems that don't make an explicit estimate for an ad campaign's effect above a baseline of behavior (i.e., what a person would have done anyways without seeing an ad campaign).



THE LADDER OF INCREMENTALITY

Many different techniques fall into each rung

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Randomized Experiments

- Randomized controlled trials (RCT)
- PSA placebo experiments
- Ghost ads
- Intent to treat
- A/B tests



Quasi-Experiments and Incrementality Models

- Judgment-based controlled experiments
- Natural experiments
- Exposed/unexposed
- Pre/post
- Market mix models
- Model-Based multi-touch attribution



Non-Incremental Models

- Rule-based multi-touch attribution
- Counting (GRPs, clicks, conversions)
- Expert opinion



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Can good data and robust models help improve the incrementality of non-experimental observational measurement methods, reducing the need for RCTs?

Meta teamed up with Northwestern's Kellogg School of Management to answer this question

1,673

RCTs from Facebook's Conversion Lift¹ platform with outcomes measured using conversion pixels

5,000+

user-level characteristics to aid model adjustment

Selected to be representative of tests run between 11/1/19 and 3/1/20 with 1M+ de-identified users in the US

Median test had 7.4M users, 22M impressions → 38B impressions in all

Hypothetical

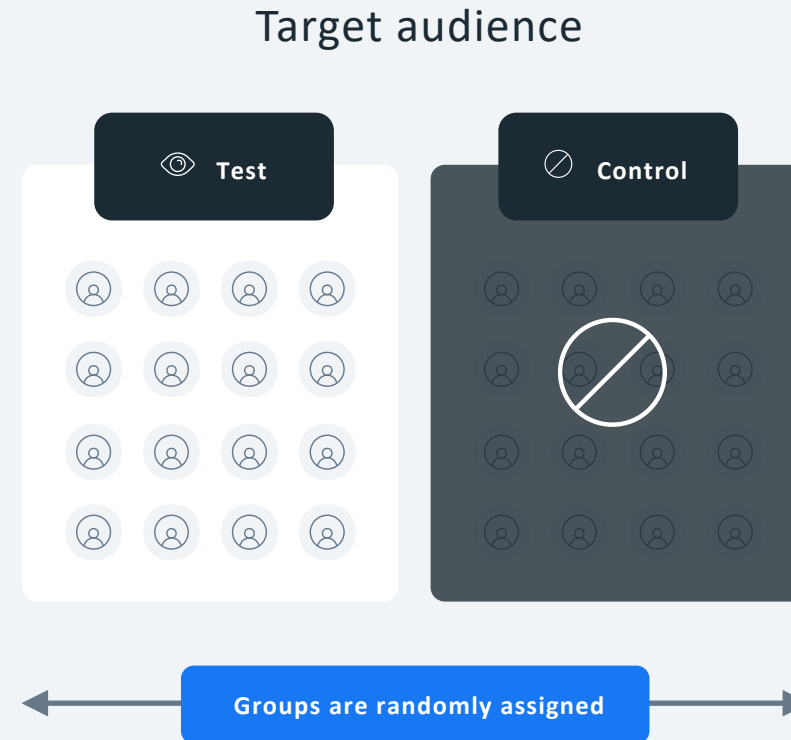
If an advertiser had not implemented a campaign as an RCT, i.e., without a randomized control group, what ad effect would they have estimated using an observational method?

1. <https://pubsonline.informs.org/doi/abs/10.1287/mksc.2022.1413>

THE LADDER OF INCREMENTALITY

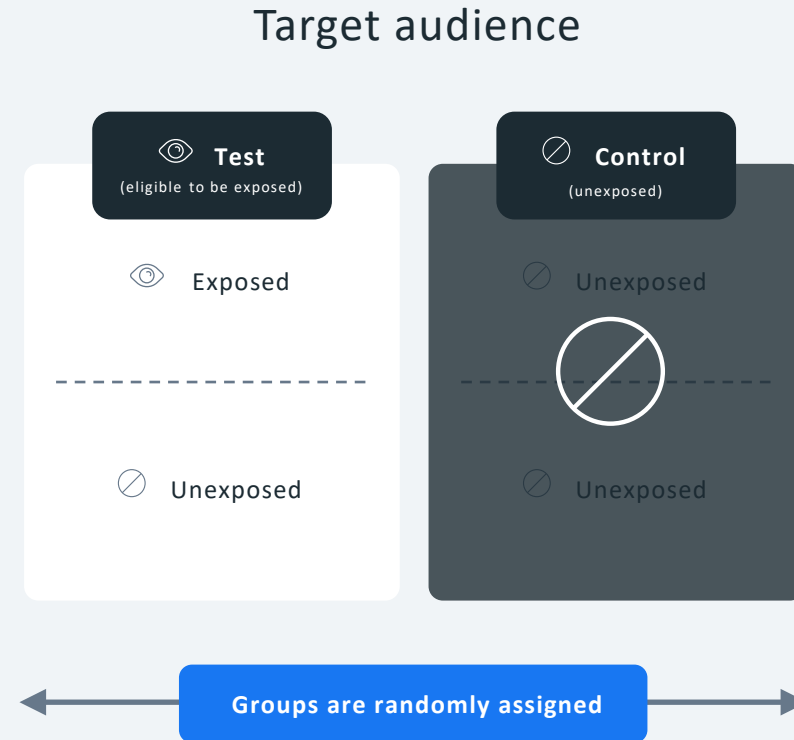
Continue that hypothetical,

Imagine we
don't have
an RCT...



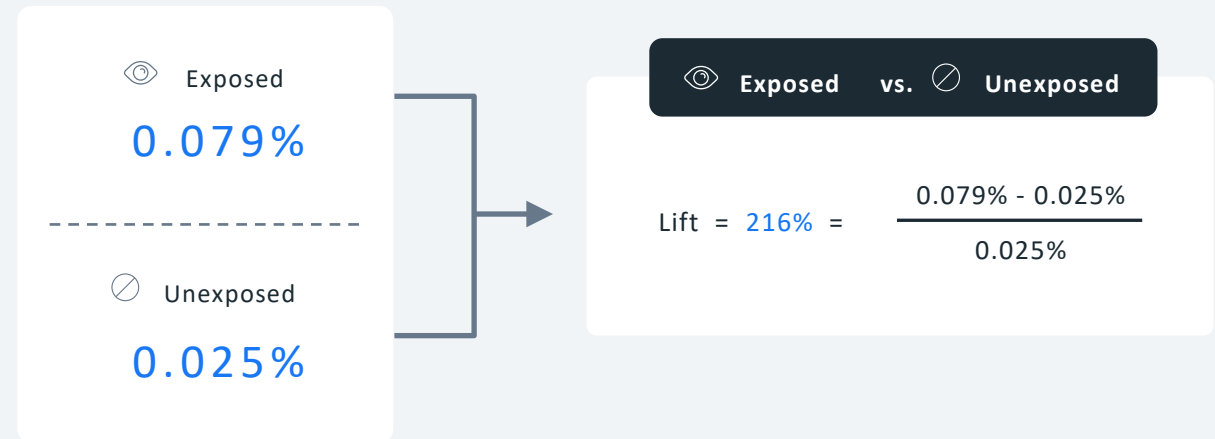
Users in test are
eligible for ads

But we don't have
a control without
the RCT

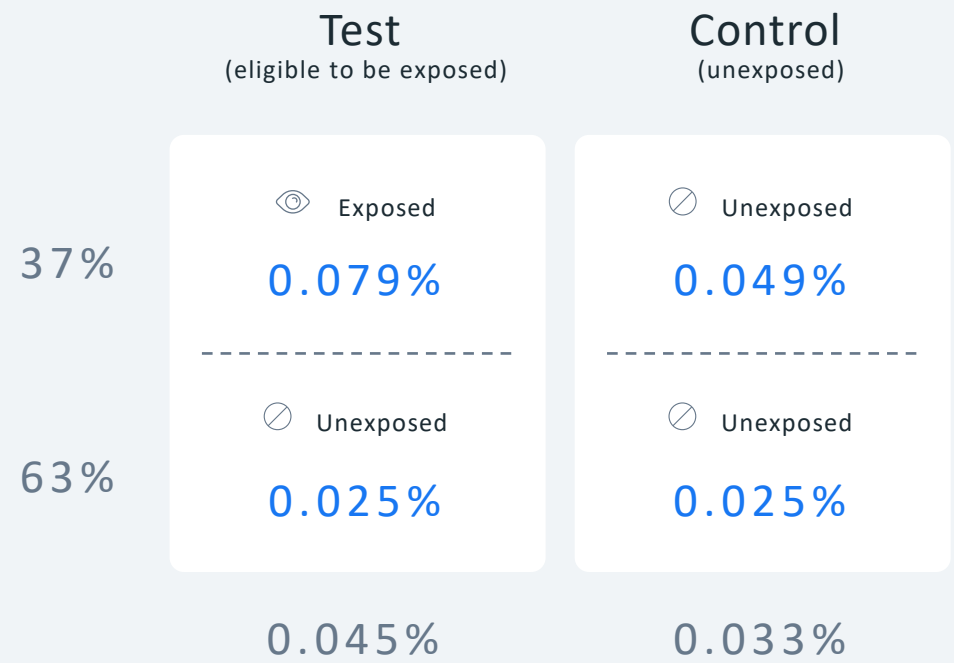


So, simply compare outcomes between people **who saw** versus did **not see** the ad campaign

Target audience

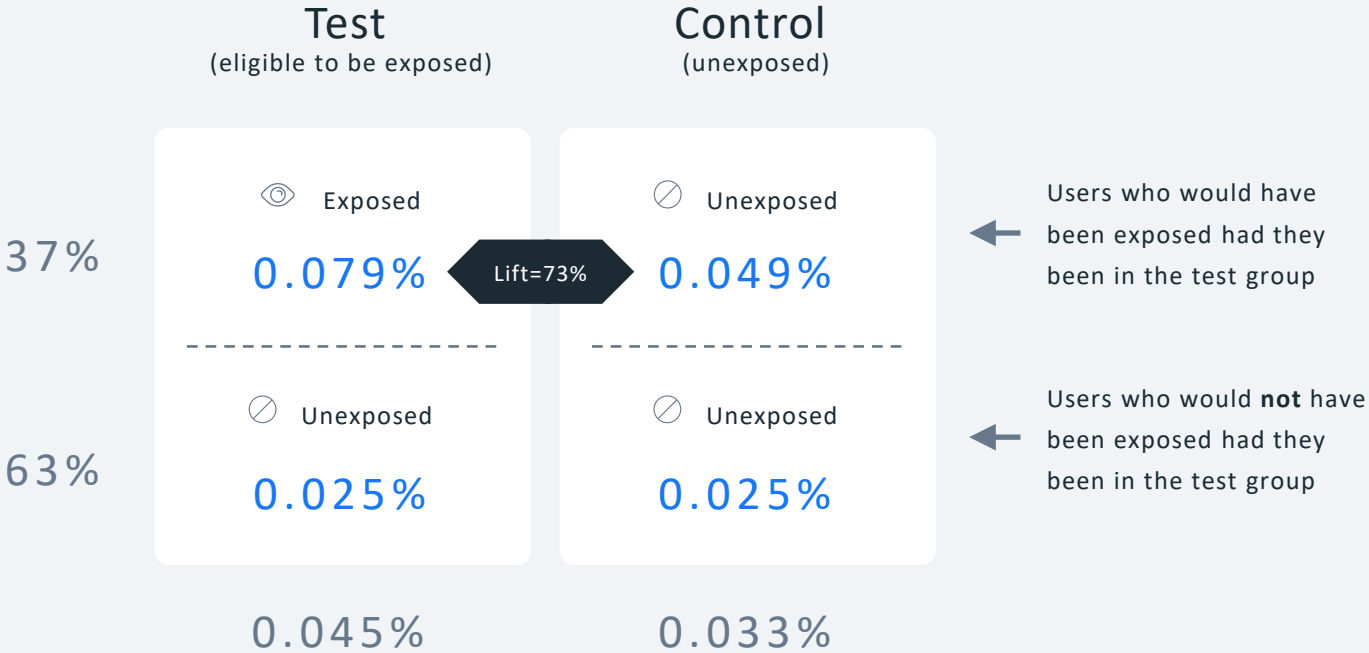


In reality, we did run an RCT and can obtain an estimate using the Test *and* Control

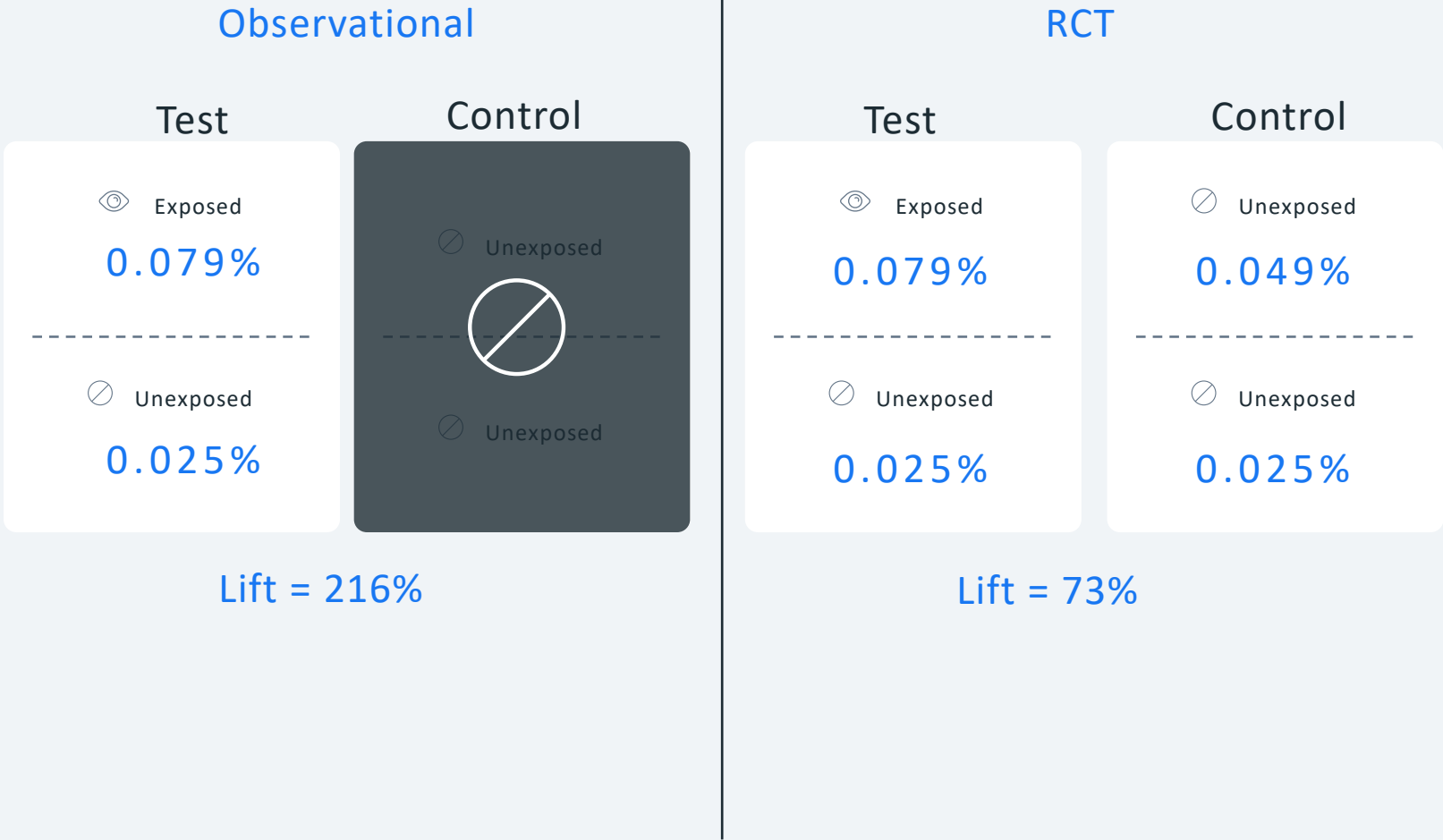


73%


The estimate measured by the RCT



How does the observational estimate compare to the RCT?



Is this a
good idea?

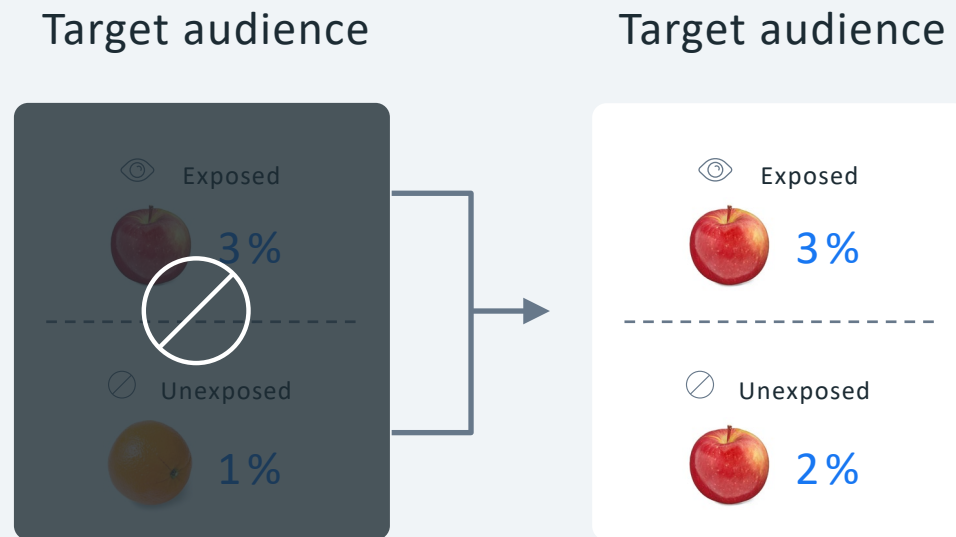


Exposed users
usually differ from
unexposed users

Advanced observational models attempt to correct for this

Goal:

Given characteristics **X** about each user, use a model to adjust for differences between exposed and unexposed users.



THE LADDER OF INCREMENTALITY

We use a significant number of user-level features and different observational models and compare observational results to RCTS

User-level features

1

Prior campaign outcomes

2

Estimated action rates

3

Dense features

4

Sparse features

Observational models



Stratified Propensity
Score Model (SPSM)

(Rosenbaum & Rubin, 1983; Imbens & Rubin, 2015)



Double/Debiased
Machine Learning (DML)

(Chernozhukov et al., 2018)

Summary of results

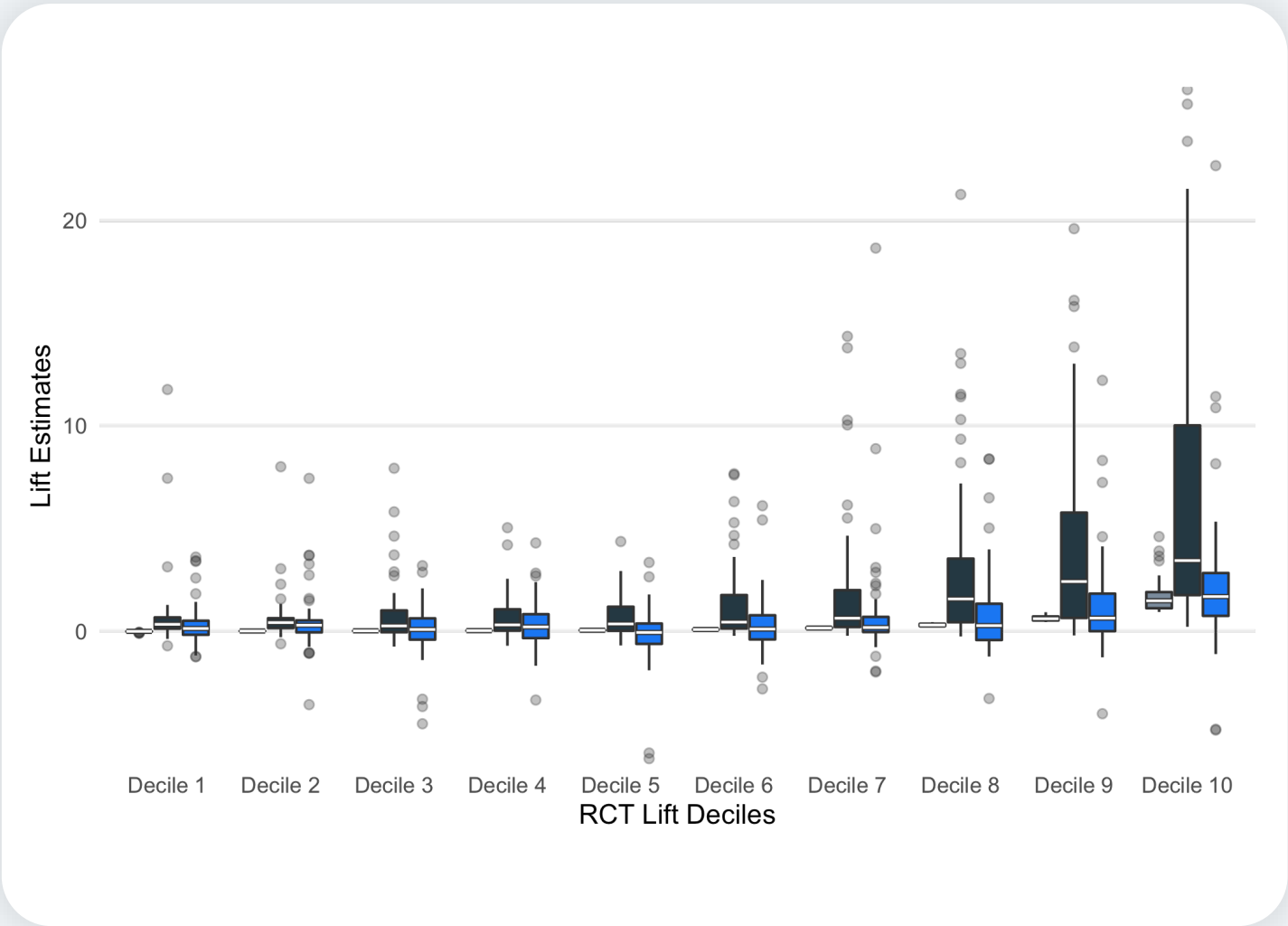
The lift estimates for RCT and DML, our best method, are statistically different in 1258 / 1673
=

75%

of the experiments

Funnel	RCT Median Lift	DML Median Lift	Relative Error
Upper	28%	143%	5.1x
Mid	19%	126%	6.6x
Lower	6%	68%	11.3x

Lower funnel
outcomes



Conclusions



Given the data available, DML generally fails to measure the true effect of advertising accurately



DML does relatively better for prospecting campaigns and those with low baseline conversion rates—but still not accurate



To improve on this, ad platforms would probably need to log data at an extremely granular level (e.g., bid-user)



But this is costly and experimental solutions are already available

However, this doesn't mean non-experimental observational methods aren't useful.

THE LADDER OF INCREMENTALITY

Techniques in the middle of the ladder can be improved upon through calibration with experiments

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Quasi-Experiments and Incrementality Models

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Calibrated with randomized experiments



Uncalibrated

Non-Incremental Models

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By calibrating and moving up the ladder, businesses can better identify ROI

Increasingly, businesses are calibrating MTA or MMM with experiments to evaluate performance.

While not as rigorous as randomized experiments, calibration allows advertisers to advance up the ladder without abandoning the measurement they already use.

2 in 3

of analyzed MMM studies
significantly changed Meta ROI
results after calibration

25%

Average variation in ROI results
after calibration

Group discussion

GROUP DISCUSSION

We built a tool to help assess your measurement systems

The self-assessment calculator provides score cards by channel.

Scores often vary by channel and can help you to explore avenues for improvement.

THREE INPUTS ARE REQUIRED

1

How much money did your organization spend in a year on advertising, marketing and promotions on major paid advertising channels?

2

For each channel, what proportion is measured with the various analysis techniques and methods (e.g., counting methods, rule-based MTA, MMM, pre/post, etc.)

3

How thorough is your organization's process for unifying measurement results across marketing channels into actionable decisions on optimizing marketing spend?



Output:

Report card with a score for how your organization utilizes incrementality-based methods for each channel and across channels

Group thought starters

1. Based on the data Joe collected and his colleague's report, what is the company doing well? What are some areas where improvements could be made? What improvements should he prioritize?
2. What are some blockers that may make it difficult for Joe to change or improve how his company measures success? How can Joe make a compelling case at his organization to drive this change?
3. What types of benchmarks and industry information should Joe take into consideration as he plans to improve measurement at his organization?

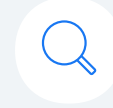
GROUP DISCUSSION

In practice, climbing the ladder requires investment at both the individual and organizational level

Your organization must be bought-in order to execute across these five best practices, especially when it comes to building culture and taking action



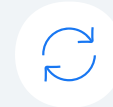
Nurture an Incrementality Culture



Use Observational Methods to Complement Experiments



Validate the Quality of Your Experiments



ABT: Always Be Testing



Take Actions on Causal Measurement Conclusions

For more information on how to climb the ladder of incrementality, check out these resources



The What and Why of Incrementality

Introduces the concept of incrementality and explains why adopting this approach can improve marketing programs



The Ladder of Incrementality

Describes and orders different measurement techniques by how rigorous and accurate they are



Climbing the Ladder of Incrementality

Provides actionable recommendations to improve the accuracy of measurement



fburl.com/incrementality-ladder

