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WHY MICRO TARGETING DOES NOT WORK AND NEVERTHELESS DISRUPTS THE PUBLIC SPHERE

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This online tool reveals your personality based on Facebook 'likes'

University of Cambridge tool estimates key psychological traits, religious and political views, intelligence and happiness using Facebook like data

NEW RESEARCH IN

Physical Sciences

Social Sciences

Computer-based personality judgments are more accurate than those made by humans



Wu Youyou, Michal Kosinski, and David Stillwell

PNAS January 27, 2015 112 (4) 1036-1040; first published January 12, 2015 <https://doi.org/10.1073/pnas.1418680112>

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Article

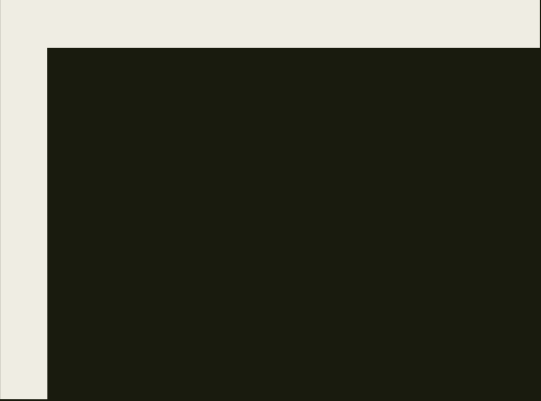
Figures & SI

Info & Metrics

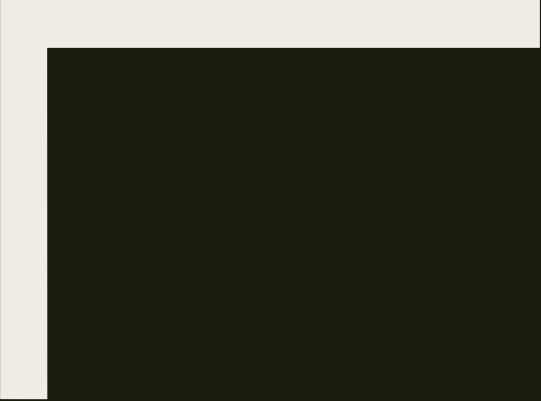
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Significance

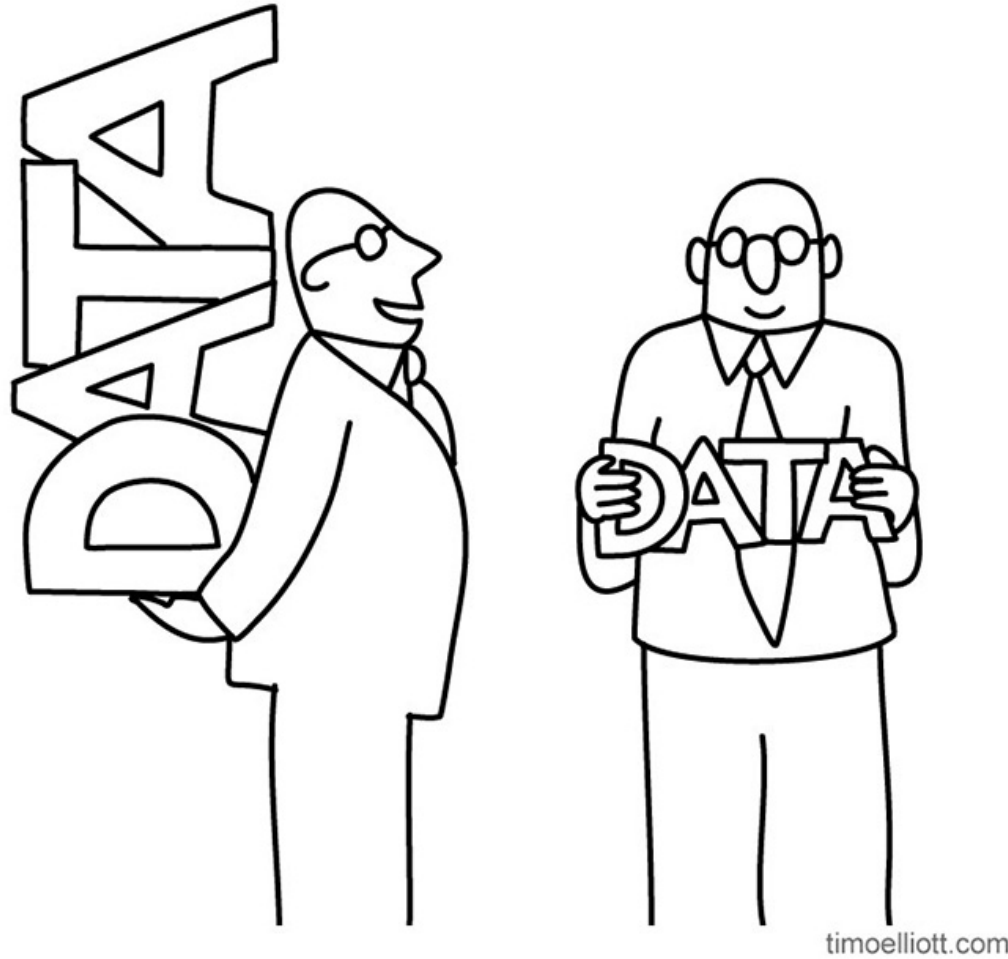
This study compares the accuracy of personality judgment—a ubiquitous and important social-cognitive activity—between computer models and humans. Using several criteria, we show that computers' judgments of people's personalities based on their digital footprints are more accurate and valid than judgments made by their close others or acquaintances (friends, family, spouse, colleagues, etc.). Our findings highlight that people's personalities can be predicted automatically and without involving human social-cognitive skills.



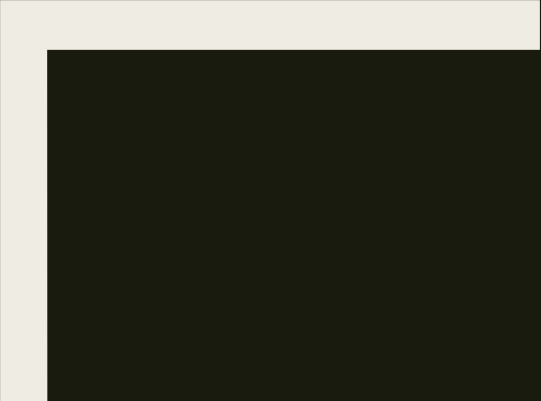
The belief
that commercial and political
micro-targeting 'works'
is just that:
a belief



This belief
is in the realm of
alchemy and astrology



“I think you’ll find that mine is bigger...”



There is no proof



The burden of proof is on those
who make the claims



Who can make such claims?
Those who have the data:
Big Tech Platforms

p-Hacking and False Discovery in A/B Testing

Ron Berman*

Leonid Pekelis†

Aisling Scott‡

Christophe Van den Bulte§

December 11, 2018

Abstract

We investigate to what extent online A/B experimenters “p-hack” by stopping their experiments based on the p-value of the treatment effect, and how such behavior impacts the value of the experimental results. Our data contains 2,101 commercial experiments



The advertising eco-system:

- Advertisers (e.g. Procter & Gambler)
- Web publishers (e.g. NYT)
- Intermediaries (e.g. platforms)
- Consumers

MOBILE



When Procter & Gamble Cut \$200 Million in Digital Ad Spend, It Increased Its Reach 10%

Unilever is also reevaluating its budget

By Lauren Johnson | March 1, 2018

 PREMIUM

DIGIDAY UK

THE GDPR IMPACT

**After GDPR, The New York Times cut off
ad exchanges in Europe — and kept
growing ad revenue**

JANUARY 16, 2019 *by* Jessica Davies

COMMENTARY

Behavioral Advertising's Benefits To Publishers Are Overstated, New Study Suggests

by **Wendy Davis**, Staff Writer @wendyndavis, May 30, 2019

For years, the ad industry has argued that free content online is fueled by online behavioral advertising, or tracking users across the web in order to deduce their interests and serve them with targeted ads.

The argument turns on the assumption that advertisers will pay more for targeted ads than generic ones, and that publishers will therefore garner more money from behaviorally targeted ads.

The claims -- which make some intuitive sense -- appear to have been widely accepted, even making their way into official policy documents. Last year, the Federal Trade Commission suggested in a **staff report** that publishers would be harmed by privacy rules that limited online tracking.



What's next?

1. Assumptions of microtargeting
2. Pitfalls of behaviourism and psychometrics
3. Reproducibility crisis in psychology and science
4. Why microtargeting does not work
5. How it nevertheless disrupts the public sphere

Assumptions of microtargeting

- What is micro-targeting?
- Let's check the common sense (in marketing parlance):
<https://www.cleverism.com/what-is-microtargeting/>

WHAT IS MICROTARGETING?

Microtargeting is one of the latest marketing techniques you can employ to promote your business. It helps organizations identify **consumers' behaviors** and tastes. It has already seen many different uses outside the marketing world.

It is a practice of collecting user's data, including what they buy, their demographics, what they like and what they are most connected to.

Subsequently, companies use this data to segment them into different groups for the sake of content or product marketing.

The basic purpose of microtargeting is to create a marketing strategy which delivers advertisements to a specific faction of society.

It predicts your interests, influences, opinions, and purchasing habits based on your behavioral, geographic, **psychographic** and demographic data.



Assumptions of micro- targeting

- If you can measure behavioural types at population level
- And measure significant correlations
- You can target individuals based on those correlations?

OBJECTIVES OF MICROTARGETING

Different people have different motivational triggers, opinions, and interests. Microtargeting aims at creating a more comprehensive advertisement strategy.

The strategy is based on predictive analyses along with the consumer data collected through different sources.

It is a proven fact that the campaigns that target a certain section of society are much more effective than unrelated advertisements.



Assumptions of micro- targeting

- Proven, or is it?
- Urban myth

Assumptions of microtargeting

- Let's check The Correspondent:
<https://thecorrespondent.com/100/the-new-dot-com-bubble-is-here-its-called-online-advertising/80216261400-7e597058>

the Correspondent

Analysis

4 days ago • Reading time 21 - 26 minutes • [Remind me later](#)

In 2018 \$273bn was spent on digital ads globally. We delve into the world of clicks, banners and keywords to find out if any of it is real. What do we really know about the effectiveness of digital advertising?

The new dot com bubble is here: it's called online advertising

But is any of it real? What do we really know about the effectiveness of digital advertising? Are advertising platforms any good at manipulating us?

You'd be forgiven for thinking the answer to that last question is: yes, extremely good. After all, the market is huge. The amount of money spent on internet ads goes up each year. In 2018, more than \$273bn dollars was spent on digital ads globally, according to research firm eMarketer. Most of those ads were purchased from two companies: Google (\$116bn in 2018) and Facebook (\$54.5bn in 2018).

A former Facebook engineer once said (and he's been quoted a thousand times over):
"The best minds of my generation are thinking about how to make people click on ads."
I spoke to some of those best minds: economists employed and formerly employed by
the most powerful companies in Silicon Valley: Yahoo!, Google, Microsoft, eBay,
Facebook, Netflix, Pandora and Amazon.

"Proprietary transformation functions," one of the consultants had said on the phone when Tadelis reached out. They used proprietary transformation functions, had 25 years of experience, and a long list of prominent clients.

When Tadelis pressed them he realised that “proprietary transformation functions” was only a clever disguise for your garden-variety statistics. You take the weekly expenditure on ads, combine it with the weekly sales, and voila! Fold the mixture into a scatter plot and see what happens. Easy as that! ▼

"This is garbage," Tadelis thought.

Correlation, as any Statistics 101 class will inform you, is not causation. What do these impressive numbers mean if the people who see your ad are the exact same people who were going to use eBay anyway? eBay is no small fry. Surely lots of people looking for shoes end up on the online auction site all by themselves, whether they see an ad or not?

Assumptions of microtargeting

- So, does it work?
- Well, who knows?

Are sales due to the selection effect?

- *Targeting people who were already buying?*

Or due to the advertising effect?

- *Targeting people who would not be buying otherwise?*

Assumptions of microtargeting

- Guess what: nobody knows
- Guess what:
it is NOT at all
'a proven fact that the campaigns
that target a certain section of
society are much more effective than
unrelated advertisements'

Assumptions of microtargeting

Why not?

- We don't know which 'certain section of society' should be targeted
- We have problems measuring effect:
 - *Pay per impression (proves nothing)*
 - *Pay per click (relevant?)*
 - *Pay per conversion (selection effect?)*



Assumptions of microtargeting

So what?

- Goodhart effect
- “If you use a measure as a target, it is no longer a good measure”



Assumptions of microtargeting

Lucas critique

Campbell effect

Goodhart effect

Assumptions of microtargeting

If you use a measure as a target, it is no longer a good measure

- Those using the measure to target **change their behaviour**
- Those targeted **change their behaviour**



Pitfalls of behaviourism and psychometrics

Behaviourism can describe this

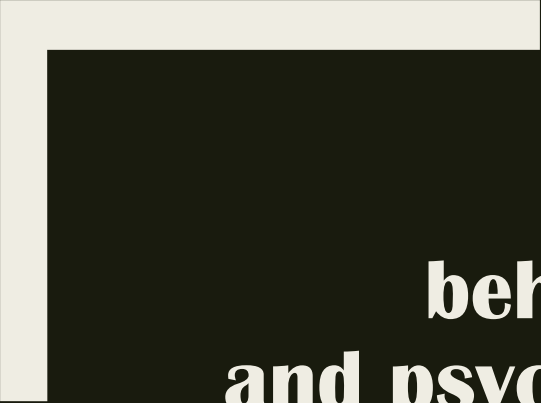
- But not in other terms than
'gaming the system'



Pitfalls of behaviourism and psychometrics

Human behaviour is far more complex

- Better described in terms of agency
- Mutual double anticipation
- Explains foundational unpredictability of human action
- Feature not a bug



Pitfalls of behaviourism and psychometrics

The Big Five: OCEAN

- **Openness**
- **Conscientiousness**
- **Extraversion**
- **Agreeableness**
- **Neuroticism**

This website provides a collection of interactive personality tests with detailed results that can be taken for personal entertainment or to learn more about personality assessment. These tests range from very serious and widely used scientific instruments popular psychology to self produced quizzes. A special focus is given to the strengths, weaknesses and validity of the various systems.

Recommended test for scientific validity

[Big Five Personality Test](#): The general consensus in academic psychology is that there are five fundamental personality traits. This model is assumed in most personality research, and is the basis of many of the most well regarded tests employed by psychologists who maintain close connections with academia. The "big five" tend to not be popular in consumer focused personality assessment or self-help because to many people the feedback of the model seems relatively basic. This test uses public domain scales from the International Personality Item Pool.

Recommended test for personal enjoyment

[Open Extended Jungian Type Scales](#): The system of personality types proposed by Carl Jung (1921) and later refined by C. Myers and I. M. Briggs has become an extremely widely used personality theory in self-help, business management, counselling and spiritual development contexts, but it is not commonly used in academic research where, like all type theories, it is treated sceptically. The system produces 16 personality types on the basis of four dichotomies and is the system used in the Myers Briggs Type Indicator and Keirsey Temperament Sorter instruments, among many others. The OEJS is a free and open source measure of the four dichotomies which yields an equivalent result to the usual tests.

myPersonality was a popular Facebook application created by David Stillwell in 2007 that allowed users to take real psychometric tests and obtain their results instantly. As well as the data from the tests, around 40% of the respondents also opted in to share data from their Facebook profile, resulting in one of the largest social science research databases in history. The application was active until 2012 and collected data from over 6 million volunteers during this time. This data was anonymised and samples of it were shared with registered academic collaborators around the world through the myPersonality project, resulting in over 45 scientific publications in peer-reviewed journals.

Respondents came from various age groups, backgrounds and cultures. They were highly motivated to answer honestly and carefully, as the only gratification they received for their participation was feedback on their results. Their scores are combined with additional information from those who opted in to sharing it with us, including detailed demographic profiles, a record of their social media behaviour, their interests, preferences, opinions, etc.

If you are interested to learn more about the myPersonality project, please contact [David Stillwell](#).



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Notices

Statement from the University of Cambridge about Dr Aleksandr Kogan

We understand from Dr Kogan that he originally created a Facebook app for academic research; however, he states that when the app was repurposed for use by GSR, it was rebranded and released with new terms and conditions, and it was made clear that this was commercial, not academic, research.

Facebook has made a series of allegations surrounding Dr Kogan's use of data. The University of Cambridge takes matters of research integrity and data protection extremely seriously. We have to date found no evidence to contradict Dr Kogan's previous assurances; nevertheless, we have written to Facebook to request all relevant evidence in their possession.

In 2015, Dr Kogan applied to the University for ethical approval to use data collected on behalf of GSR for his academic research. His application was reviewed and subsequently rejected. Dr Kogan was in the process of re-applying when Facebook requested deletion of the data; hence the application was withdrawn.

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 Psychology Press
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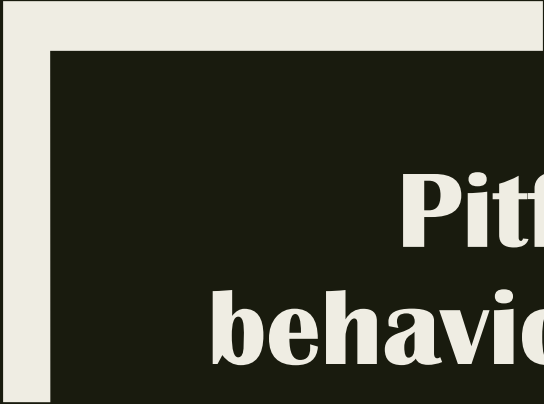
Jack Block, the Big Five, and Personality From the Standpoints of Actor, Agent, and Author

Dan P. McAdams and Keegan Walden

Department of Psychology, Northwestern University, Evanston, Illinois

Jack Block died before his target article could be published. Submitted to *Psychological Inquiry* by his son while Jack was in the hospital, the paper shows many of the earmarks of Jack Block's long and brilliant career as a personality psychologist. Passionate, contrarian, and doggedly analytical, the paper's traits reflect the man's. Jack Block wanted to construct a scientifically rigorous psychology of personality that fully captures the complexity and coherence of human lives as they develop over time. To do so was his life-

ments he has made before. There is nothing elegant or simple about it. It may be that Jack Block did not have time to iron out the paper's rough edges before his life was cut off. But more likely, Jack's swan song, like so many of the influential articles he wrote in his career, was always going to be a gritty and restless communiqué from the battlefield. Jack Block was a fighter. He fought against the accepted truths and the conventional fashions of the discipline. Over the past 15 years, he fought against the Big Five.



Pitfalls of behaviourism and psychometrics

Block enumerated five broad criticisms of the fivefactor model of personality.

He argued that the model:

- a. is lacking in theory,
- b. relies too much on factor analysis,
- c. leaves out important traits,
- d. fails to take into consideration critical developments in trait measurement, and
- e. may be superseded by a two factor approach.

Pitfalls of behaviourism and psychometrics

- "From our own contrarian point of view, a full accounting of personality requires analysis from the standpoints of **actor, agent, and author**.
- By focusing most of their attention on the first of these three, personality psychologists continue to shrink away from the field's historical mandate to study **the full gamut of human personality, in all of its richness and complexity.**"

NEW RESEARCH IN

Physical Sciences

Social Sciences

Computer-based personality judgments are more accurate than those made by humans



Wu Youyou, Michal Kosinski, and David Stillwell

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Info & Metrics

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Significance

This study compares the accuracy of personality judgment—a ubiquitous and important social-cognitive activity—between computer models and humans. Using several criteria, we show that computers' judgments of people's personalities based on their digital footprints are more accurate and valid than judgments made by their close others or acquaintances (friends, family, spouse, colleagues, etc.). Our findings highlight that people's personalities can be predicted automatically and without involving human social-cognitive skills.

Personality traits, like many other psychological dimensions, are latent and cannot be measured directly; various perspectives exist regarding the evaluation criteria of judgmental accuracy (3, 5). We adopted the realistic approach, which assumes that personality traits represent real individual characteristics, and the accuracy of personality judgments may be benchmarked using three key criteria: self-other agreement, interjudge agreement, and external validity (1, 5, 7). We apply those benchmarks to a sample of 86,220 volunteers,* who filled in the 100-item International Personality Item Pool (IPIP) Five-Factor Model of personality (14) questionnaire (15), measuring traits of openness, conscientiousness, extraversion, agreeableness, and neuroticism.

Algorithmic psychometrics and the scalable subject

Social Studies of Science

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journals.sagepub.com/home/sss



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Abstract

Recent public controversies, ranging from the 2014 Facebook ‘emotional contagion’ study to psychographic data profiling by Cambridge Analytica in the 2016 American presidential election, Brexit referendum and elsewhere, signal watershed moments in which the intersecting trajectories of psychology and computer science have become matters of public concern. The entangled history of these two fields grounds the application of applied psychological techniques to digital technologies, and an investment in applying calculability to human subjectivity. Today, a quantifiable psychological subject position has been translated, via ‘big data’ sets and algorithmic analysis, into a model subject amenable to classification through digital media platforms. I term this position the ‘scalable subject’, arguing it has been shaped and made legible by algorithmic psychometrics – a broad set of affordances in digital platforms shaped by psychology and the behavioral sciences. In describing the contours of this ‘scalable subject’, this paper highlights the urgent need for renewed attention from STS scholars on the psy sciences, and on a computational politics attentive to psychology, emotional expression, and sociality via digital media.

Keywords

affect, big data, emotion, Facebook, platforms, psychology, psychometrics, scale, social media, subjectivity

Pitfalls of behaviourism and psychometrics

DATA FALLACIES TO AVOID



CHERRY PICKING

Selecting results that fit your claim and excluding those that don't.



DATA DREDGING

Repeatedly testing new hypotheses against the same set of data, failing to acknowledge that most correlations will be the result of chance.



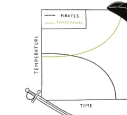
SURVIVORSHIP BIAS

Drawing conclusions from an incomplete set of data, because that data has 'survived' some selection criteria.



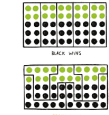
COBRA EFFECT

Setting an incentive that accidentally produces the opposite result to the one intended. Also known as a Perverse Incentive.



FALSE CAUSALITY

Falsely assuming when two events appear related that one must have caused the other.



GERRYMANDERING

Manipulating the geographical boundaries used to group data in order to change the result.



SAMPLING BIAS

Drawing conclusions from a set of data that isn't representative of the population you're trying to understand.



GAMBLER'S FALLACY

Mistakenly believing that because something has happened more frequently than usual, it's now less likely to happen in future (and vice versa).



HAWTHORNE EFFECT

The act of monitoring someone can affect their behaviour, leading to spurious findings. Also known as the Observer Effect.



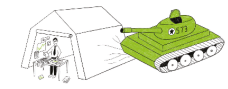
REGRESSION TOWARDS THE MEAN

When something happens that's unusually good or bad, it will revert back towards the average over time.

APPLICATION SUCCESS RATE			
	MALE	FEMALE	
UNIVERSITY 1	18.7%	15.3%	?
UNIVERSITY 2	5.2%	7.8%	
TOTAL	12.8%	14.5%	

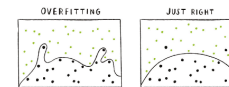
SIMPSON'S PARADOX

When a trend appears in different subsets of data but disappears or reverses when the groups are combined.



MCMANARA FALLACY

Relying solely on metrics in complex situations and losing sight of the bigger picture.



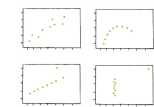
OVERFITTING

Creating a model that's overly tailored to the data you have and not representative of the general trend.



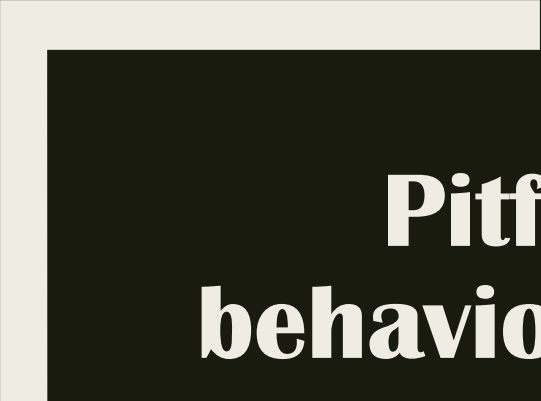
PUBLICATION BIAS

Interesting research findings are more likely to be published, distorting our impression of reality.



DANGER OF SUMMARY METRICS

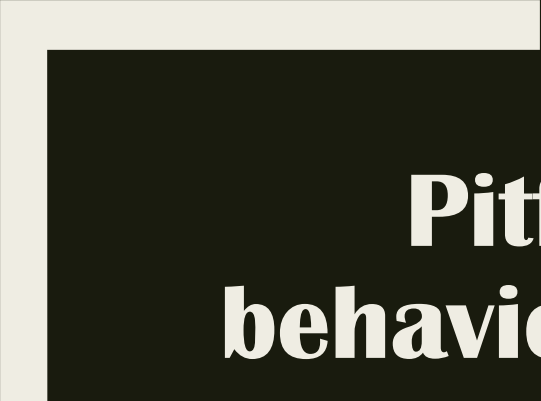
Only looking at summary metrics and missing big differences in the raw data.



Pitfalls of behaviourism and psychometrics

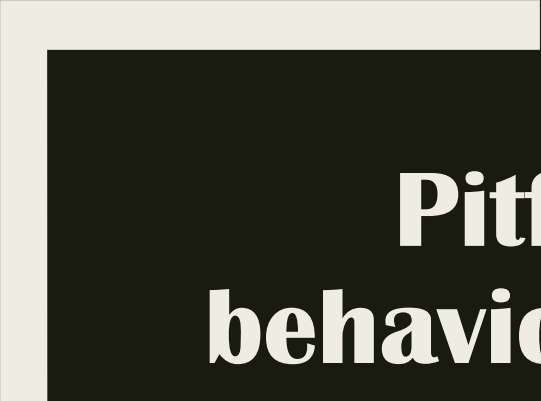
Relationships in data:

- **mere correlations**
(the more data, the more spurious correlations)
- **causal relationships**
(mathematical and empirical verification)
- **conceptual relationships**
(analytical and performative)



Pitfalls of behaviourism and psychometrics

- Analytical:
 - *correlation between bachelors and unmarried men*
 - *correlation between danger and risk (conceptual overlap)*
 - *correlation between fear and danger (conceptual overlap)*




Pitfalls of behaviourism and psychometrics

- Performative:
 - *'I declare you husband and wife'*
 - *This (whatever linked variables) 'counts as' an emotion of fear*
 - *This (whatever linked variables) 'counts as' a PhD in law (in Canada)*


Pitfalls of behaviourism and psychometrics

- Performative:
 - *This (whatever linked variables) 'counts as' a pedestrian crossing*
 - Qualification
(performative speech act)
precedes quantification



Reproducibility crisis in psychology and science

- Reproducibility crisis is related to problematic use of statistics
- ML applications are very vulnerable to data fallacies, p-hacking, data dredging, null-hypothesis delusion
- This goes even more for behavioural targeting



Reproducibility crisis in psychology and science

- Micro-targeting imports the horrors of that crisis:
 - *False claims*
 - *Irresponsible trade-offs*
 - *Unreliable applications*

Statistical Rituals: The Replication Delusion and How We Got There

Gerd Gigerenzer

First Published June 14, 2018 | Research Article |  Check for updates

<https://doi.org/10.1177/2515245918771329>

[Article information](#) ▾

 Altmetric 381 

Abstract

The “replication crisis” has been attributed to misguided external incentives gamed by researchers (the *strategic-game hypothesis*). Here, I want to draw attention to a complementary internal factor, namely, researchers’ widespread faith in a statistical ritual and associated delusions (the *statistical-ritual hypothesis*). The “null ritual,” unknown in statistics proper, eliminates judgment precisely at points where statistical theories demand it. The crucial delusion is that the p value specifies the probability of a successful replication (i.e., $1 - p$), which makes replication studies appear to be superfluous. A review of studies with 839 academic psychologists and 991 students shows that the replication delusion existed among 20% of the faculty teaching statistics in psychology, 39% of the professors and lecturers, and 66% of the students. Two further beliefs, the illusion of certainty (e.g., that statistical significance proves that an effect exists) and Bayesian wishful thinking (e.g., that the probability of the alternative hypothesis being true is $1 - p$), also make successful replication appear to be certain or almost certain, respectively. In every study reviewed, the majority of researchers (56%–97%) exhibited one or more of these delusions. Psychology departments need to begin teaching statistical thinking, not rituals, and journal editors should no longer accept manuscripts that report results as “significant” or “not significant.”

Keywords

[replication](#), [p-hacking](#), [illusion of certainty](#), [p value](#), [null ritual](#)

The essence of this hybrid theory is the null ritual (Gigerenzer, 2004):

1. Set up a null hypothesis of “no mean difference” or “zero correlation.” Do not specify the predictions of your own research hypothesis.
2. Use 5% as a convention for rejecting the null hypothesis. If the test is significant, accept your research hypothesis. Report the test result as $p < .05$, $p < .01$, or $p < .001$, whichever level is met by the obtained p value.
3. Always perform this procedure.

The null ritual does not exist in statistics proper. This point is not always understood; even its critics sometimes confuse it with Fisher’s theory of null-hypothesis testing and call it “null-hypothesis significance testing.” In fact, the ritual is an incoherent mishmash of ideas from Fisher on the one hand and Neyman and Pearson on the other, spiked with a characteristically novel contribution: the elimination of researchers’ judgment.

The Earth Is Round ($p < .05$)

Jacob Cohen

After 4 decades of severe criticism, the ritual of null hypothesis significance testing—mechanical dichotomous decisions around a sacred .05 criterion—still persists. This article reviews the problems with this practice, including its near-universal misinterpretation of p as the probability that H_0 is false, the misinterpretation that its complement is the probability of successful replication, and the mistaken assumption that if one rejects H_0 one thereby affirms the theory that led to the test. Exploratory data analysis and the use of graphic methods, a steady improvement in and a movement toward standardization in measurement, an emphasis on estimating effect sizes using confidence intervals, and the informed use of available statistical methods is suggested. For generalization, psychologists must finally rely, as has been done in all the older sciences, on replication.

sure how to test H_0 , chi-square with Yates's (1951) correction or the Fisher exact test, and wonders whether he has enough power. Would you believe it? And would you believe that if he tried to publish this result without a significance test, one or more reviewers might complain? It could happen.

Almost a quarter of a century ago, a couple of sociologists, D. E. Morrison and R. E. Henkel (1970), edited a book entitled *The Significance Test Controversy*. Among the contributors were Bill Rozeboom (1960), Paul Meehl (1967), David Bakan (1966), and David Lykken (1968). Without exception, they damned NHST. For example, Meehl described NHST as “a potent but sterile intellectual rake who leaves in his merry path a long train of ravished maidens but no viable scientific offspring” (p. 265). They were, however, by no means the first to do so. Joseph Berkson attacked NHST in 1938, even before it sank its deep roots in psychology. Lancelot Hogben's

What's wrong with NHST? Well, among many other things, it does not tell us what we want to know, and we so much want to know what we want to know that, out of desperation, we nevertheless believe that it does! What we want to know is “Given these data, what is the probability that H_0 is true?” But as most of us know, what it tells us is “Given that H_0 is true, what is the probability of these (or more extreme) data?” These are not the same, as has been pointed out many times over the years by the contributors to the Morrison–Henkel (1970) book, among

Why micro-targeting does not work (as claimed)

- Problems with null hypothesis testing:
 - *If A then probably not B*
 - *If B then not A?*
- If one is Belgium one is probably not a member of the Belgium Senate
- One is a member of the Belgium Senate, so one is not Belgium?

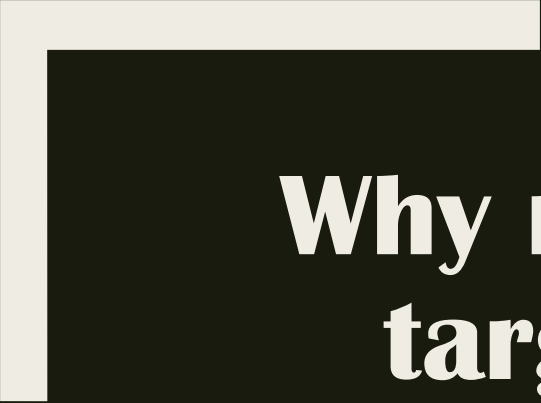
Why micro-targeting does not work (as claimed)

- Problems with null hypothesis testing:
 - *If A then probably not B*
 - *If B then not A?*

A = null hypothesis

B = the data

- If A is true, probability that B = 4%, is often thought to imply that:
- If B, probability that A is true = 4%, **quod non**



Why micro-targeting does not work (as claimed)

- Problems with null hypothesis testing:
 - *If A then probably not B*
 - *If B then not A?*

A = null hypothesis

B = the data

- If B, what is the probably that A is true? (this is what you want to know)

Assumptions of microtargeting

Why is there no proof that micro-targeting works?

- We don't know which 'certain section of society' should be targeted
- We have problems measuring effect:
 - *Pay per impression (proves nothing)*
 - *Pay per click (relevant?)*
 - *Pay per conversion (selection effect?)*



The public sphere

- AdDrivenContent
- Algorithms optimising for increasing ad revenue



The public sphere

- Hannah Arendt:
I am not worried that behaviourism is true,
but that it will become true



The public sphere

**Not everything that can be counted counts
Not everything that counts can be counted**



The public sphere

**Not everything that can be controlled matters
Not everything that matters can be controlled**

Fake news on Twitter during the 2016 U.S. presidential election

Nir Grinberg^{1,2*}, Kenneth Joseph^{3*}, Lisa Friedland^{1*},
Briony Swire-Thompson^{1,2}, David Lazer^{1,2†}

The spread of fake news on social media became a public concern in the United States after the 2016 presidential election. We examined exposure to and sharing of fake news by registered voters on Twitter and found that engagement with fake news sources was extremely concentrated. Only 1% of individuals accounted for 80% of fake news source exposures, and 0.1% accounted for nearly 80% of fake news sources shared. Individuals most likely to engage with fake news sources were conservative leaning, older, and highly engaged with political news. A cluster of fake news sources shared overlapping audiences on the extreme right, but for people across the political spectrum, most political news exposure still came from mainstream media outlets.

Discussion

This study estimated the extent to which people on Twitter shared and were exposed to content from fake news sources during the 2016 election season. Although 6% of people who shared URLs with political content shared content from fake news sources, the vast majority of fake news shares and exposures were attributable to tiny fractions of the population. Though previous work has shown concentration of volume both in political conversations on Twitter (21) and in fake news consumption (9), the extreme levels we observed are notable. For the average panel member, content from fake news sources constituted only 1.18% of political exposures, or about 10 URLs during the last month of the election campaign. These averages are of similar magnitude to estimates from previous work (8, 9), which is noteworthy given the vastly different study populations and methodologies. As in these studies, we found that the vast majority of political exposures, across all political groups, still came from popular non-fake news sources. This is reassuring in contrast to claims of political echo chambers (22) and fake news garnering more engagement than real news during the election (3).

← Thread



jack 🌍🌍🌍🇺🇸
@jack

We've made the decision to stop all political advertising on Twitter globally. We believe political message reach should be earned, not bought. Why? A few reasons... 🧵

9:05 PM · Oct 30, 2019 · [Twitter for iPhone](#)

104.4K Retweets 430.7K Likes



jack 🌍🌍🌍🇺🇸 @jack · Oct 30
Replying to @jack

A political message earns reach when people decide to follow an account or retweet. Paying for reach removes that decision, forcing highly optimized and targeted political messages on people. We believe this decision should not be compromised by money.

651

6.5K

56.3K



jack 🌍🌍🌍🇺🇸 @jack · Oct 30

While internet advertising is incredibly powerful and very effective for commercial advertisers, that power brings significant risks to politics, where it can be used to influence votes to affect the lives of millions.

255

4.8K

42.5K





The public sphere

Micro-targeting operates at two levels:

- *Fake news (posted without payment)*
- *Fake ads (posted against payment)*
- Fake news is amplified due to:
 - *Ad-driven content, as algorithms optimise for increased ad revenue*
- The difference is questionable:
 - *Advertorials blur the difference*
 - *Stopping political advertising does not solve the problem of ad-driven content*

Karlin Lillington: Ban all indiscriminate data gathering

Online advertising 'an almost completely unregulated environment' with no transparency

🕒 Thu, Nov 7, 2019, 05:30



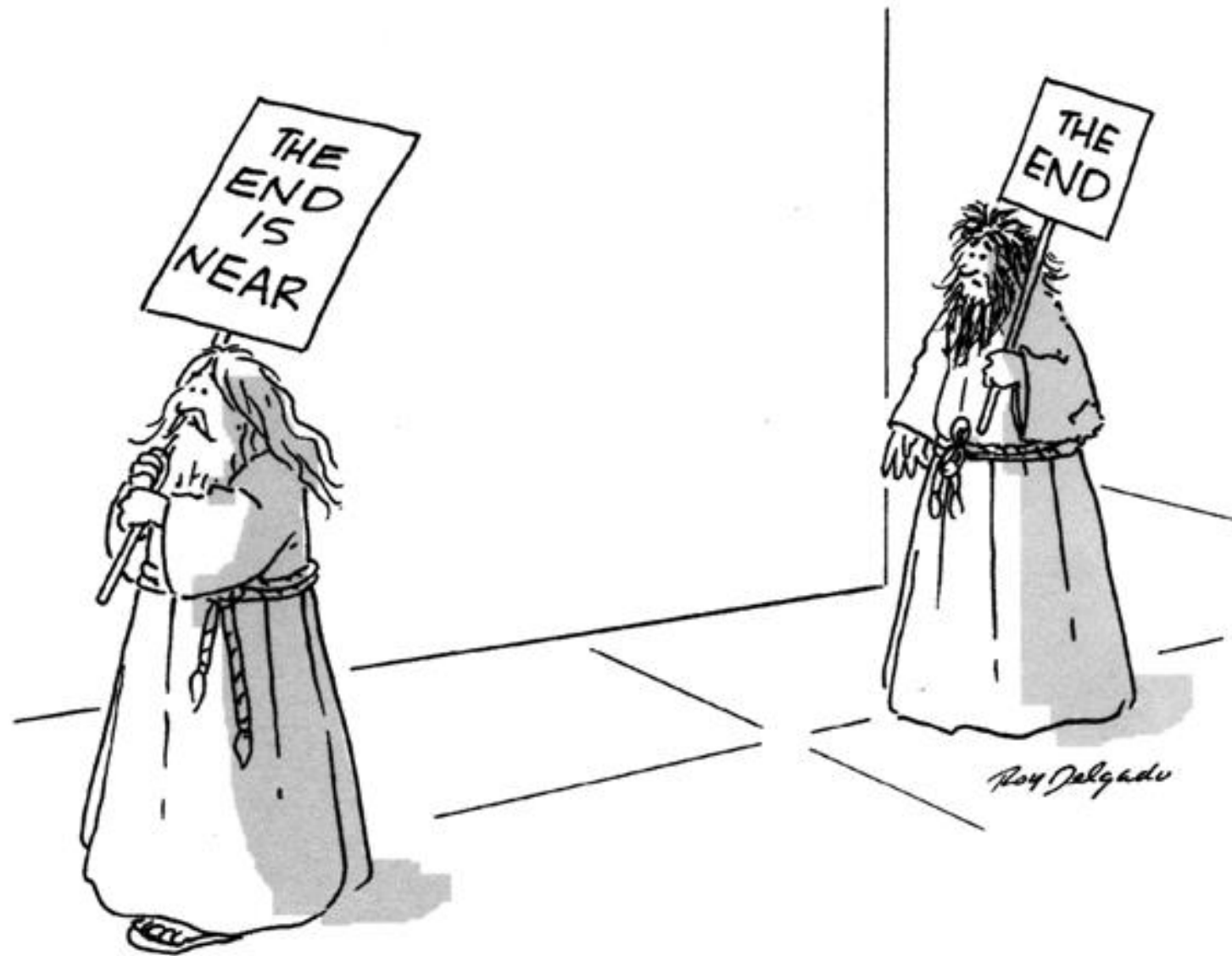
Karlin Lillington

 Follow



Twitter chief executive Jack Dorsey testifies before the Senate Intelligence Committee in Washington. Twitter is to stop accepting political advertising globally on its platform. Photograph: Jim Watson/AFP via Getty Images

There's a far better approach to all of this. Just disallow all indiscriminate data gathering, ban sales of data to third parties, and ban microtargeting. Full stop. Everywhere online. For everything.



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hyperlinks

- <https://www.cleverism.com/what-is-microtargeting/>
- <https://openpsychometrics.org/>
- <https://www.psychometrics.cam.ac.uk/productservices/mypersonality>
- <https://www.geckoboard.com/learn/data-literacy/statistical-fallacies/>

See also slides Europe Lecture

- https://www.montesquieu-instituut.nl/9353262/g/ppt/mireille.pdf?pk_campaign=txtmmpdf-1809&pk_kwd=1809
- For the lecture: https://www.montesquieu-instituut.nl/id/vklpnxbnxnr8/democratie_in_de_21e_eeuw_de_impact_van