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8
 9 IN THE UNITED STATES DISTRICT COURT
 10 FOR THE NORTHERN DISTRICT OF CALIFORNIA
 11 SAN JOSE DIVISION

12
 13 **NetChoice, LLC,**

14 Plaintiff,

15 v.

16 **Rob Bonta, in his official capacity as**
 17 **Attorney General of the State of California,**

18 Defendant.

Case No. 5:22-cv-08861-BLF

**DECLARATION OF
 SERGE EGELMAN, PH.D. IN SUPPORT
 OF DEFENDANT’S OPPOSITION TO
 PLAINTIFF’S MOTION FOR A SECOND
 PRELIMINARY INJUNCTION**

Date: January 23, 2025
 Time: 9:00 a.m.
 Dept: 3
 Judge: The Honorable Beth Labson
 Freeman
 Trial Date: Not scheduled
 Action Filed: 12/14/2022

1 I, Serge Egelman, Ph.D., declare and state as follows:

2 1. I submit this declaration in support of Defendant’s Opposition to Plaintiff’s Motion for a
3 Second Preliminary Injunction.

4 **BACKGROUND & QUALIFICATIONS**

5 2. I am the Research Director of the Usable Security & Privacy Group at the International
6 Computer Science Institute (ICSI), which is a non-profit research institute affiliated with
7 the University of California, Berkeley. I also hold a position as a research scientist within
8 the Electrical Engineering and Computer Sciences (EECS) Department at the University
9 of California, Berkeley. I received my Ph.D. from Carnegie Mellon University’s School of
10 Computer Science. My research has been cited over 13,000 times, and my h-index—the
11 most common metric for scientific impact¹—is over 50.²

12 3. I have been performing research into online privacy for over twenty years. My research
13 focuses on the interplay of online privacy, computer security, and human factors. In short,
14 I study: consumer privacy and security decision making; consumer privacy preferences;
15 privacy and security expectations; and how those expectations comport with reality (e.g.,
16 by performing technical analyses of online services and other software to examine privacy
17 and security practices). This research involves both technical knowledge to build tools for
18 use in measurement studies (e.g., observational studies of how user data is collected and
19 shared in practice), as well as a deep understanding of how to apply social science
20 methodologies (e.g., human subjects research, surveys, interviews, randomized controlled
21 trials, etc.). I have served as an invited expert for several web standards efforts that
22 pertained to privacy and security, and have received over a dozen awards from the
23 research community (including: privacy research awards from two European data
24 protection authorities, AEPD in Spain and CNIL in France; the USENIX Security
25 Symposium Distinguished Paper Award, from one of the top academic computer security
26 conferences; the Caspar Bowden Award for Outstanding Research in Privacy Enhancing

27 ¹ J.E. Hirsch. “An Index to Quantify an Individual's Scientific Research Output.” *Proc.*
28 *Natl. Acad. Sci. U.S.A.* 102 (46) 16569-16572, <https://doi.org/10.1073/pnas.0507655102> (2005).

² <https://scholar.google.com/citations?user=WN9t4n0AAAAJ&hl=en>

1 Technologies; and seven paper awards from the ACM Special Interest Group on
2 Computer-Human Interaction [SIGCHI], the top human-computer interaction conference).
3 I have also been repeatedly invited to speak at the FTC’s annual “PrivacyCon” event
4 based on my laboratory’s research.

5 4. Over the past decade, my laboratory has been studying the mobile app ecosystem, which
6 has included building tools to detect when personal information is accessed by mobile
7 apps and the third parties with whom they share it. We have used these tools in peer-
8 reviewed published research studies about consumer privacy, including examining mobile
9 apps’ compliance with various privacy regulations and platform policies.

10 5. One research study performed by my laboratory demonstrated that a majority of child-
11 directed Android apps appeared to be violating COPPA,³ which led to major policy shifts
12 by both Google and Apple, makers of the two leading mobile platforms. I have since been
13 invited to give keynotes at several international conferences on child development and
14 technology as an expert on online privacy as it pertains to children. I have also testified
15 before the U.S. Senate on how COPPA can be improved to match the realities of modern
16 technology, and have been asked to provide feedback on draft legislation from members
17 of both houses of Congress.

18 6. My *curriculum vitae*, which sets forth my experience and credentials more fully, is
19 attached as Exhibit A.

20 7. I have testified as an expert in the following cases:

- 21 • *Garner v. Amazon.com, Inc.*, Case No. 2:21-cv-00750 (W.D. Wash.)
- 22 • *Lopez et al. v. Apple, Inc.*, Case No. 4:19-cv-04577-JSW (N.D. Cal.)
- 23 • *Martinez et al. v. D2C, LLC d/b/a UNIVISION NOW*, Case No. 1:23-cv-21394-RNS
24 (S.D. Fla.).
- 25 • *Bloom v. Zuffa LLC*, Case No. 2:22-cv-00412-RFB-BNW (D. Nev.)

26 _____
27 ³ Irwin Reyes, Primal Wijesekera, Joel Reardon, Amit Elazari Bar On, Abbas
28 Razaghpanah, Narseo Vallina-Rodriguez, and Serge Egelman. “*Won’t Somebody Think of the Children?*” *Examining COPPA Compliance at Scale*. Proceedings on Privacy Enhancing Technologies (PoPETS), 2018(3):63–83.

- 1 • *Clark, et. al. v. Yodlee, Inc.*, Case No: 3:20-cv-05991-SK (N.D. Cal.).
- 2 • *Czarnionka v. The Epoch Times Association, Inc.*, Case No. 1:22-cv-6348 (S.D.N.Y.)
- 3 • *Frasco v. Flo Health, et al.*, Case No. 3:21-cv-00757 (N.D. Cal.).
- 4 • *Hart v. TWC Prod. & Tech. LLC*, 526 F. Supp. 3d 592, Case No. 20-cv-03842-
5 JST (N.D. Cal. 2021)
- 6 • *District of Columbia v. Town Sports International, LLC*, Case No. 2020 CA 003691
7 B (D.C. Sup. Ct. 2020)
- 8 • *In re Vizio, Inc., Consumer Privacy Litig.*, 238 F. Supp. 3d 1204, (C.D. Cal. 2017)
- 9 • *In re LinkedIn User Privacy Litigation*, 932 F. Supp. 2d 1089 (N.D. Cal. 2013)
- 10 • *In re Netflix Privacy Litigation*, Case No.: 5:11-CV-00379 EJD (N.D. Cal. 2012)

- 11 8. I am being compensated in the above-entitled case at an hourly rate of \$400/hour for
12 preparing this declaration. My compensation is not in any way dependent on the outcome
13 of this or any related proceeding.
- 14 9. The opinions in this declaration are my expert opinions, which are based on my education
15 and training, my peer-reviewed published research and the research of others, my
16 knowledge of relevant technologies (including my reading of the public technical
17 documents offered by NetChoice’s members about their capabilities), as well as my
18 reading of the legislation.
- 19 10. I have reviewed AB 2273, California’s Age Appropriate Design Code (AADC) Act. In my
20 expert opinion, this law is necessary to address realities of modern technology that have
21 resulted in the exploitation of minors; its provisions are reasonable and technically
22 feasible to adopt (i.e., the technologies necessary to comply are already in widespread use
23 by NetChoice’s members), and I believe that they are substantially similar to policies in
24 other jurisdictions within which NetChoice members operate. The law only applies to
25 services that are likely to be used by children (rather than all online services), and only
26 requires that companies take steps to limit harm to children, allowing them and their
27 parents to make more informed decisions about their online activities and the
28 dissemination of their personal information. Services not likely to be used by children are

1 unlikely to be impacted by this legislation; child-directed services can comply by simply
2 limiting privacy-invasive tracking and considering potential harms to children. Similar
3 laws already exist in other sectors, which society has accepted: that convenience stores
4 cannot sell tobacco products and alcohol to minors is not viewed as tyrannical overreach
5 or limitations on “freedom to innovate,” but instead as a commonsense safeguard.

6 **COLLECTION & USE OF PERSONAL INFORMATION ONLINE**

- 7 11. The “free” Internet is subsidized through the collection of users’ personal information for
8 both advertising and analytics purposes. In the case of advertising, this means showing
9 Internet users ads that are specifically tailored to their inferred interests. In the case of
10 analytics, this means observing how users interact with the service in order to maximize
11 its profitability (e.g., strategically placing in-app purchase opportunities based on users’
12 in-app behaviors, identifying the users most likely to buy expensive items based on their
13 inferred demographics, manipulating users into spending more time using a service, etc.).
14 In other cases, this may mean straight up selling the user data to third parties so that those
15 third parties may perform these and other yet-unknown activities.
- 16 12. Because so much of the Internet is supported by advertisements, one key metric that
17 online services use is known as “engagement,” which refers to the amount of time that
18 consumers spend using a service or the frequency of interactions that consumers have with
19 that service. That is, the more time consumers spend using a service that displays ads, the
20 more ads that consumers are likely to be shown, and therefore the more revenue that the
21 service can derive by charging advertisers to show those ads. Similarly, the more personal
22 information that consumers share with a service, the more likely those consumers are to
23 see “relevant”⁴ ads, and therefore the more likely they are to click those ads.

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26 ⁴ Based on my experience, I am not convinced that highly targeted ads based on
27 consumers’ personal information provide benefits for anyone beyond the advertising companies:
28 the liabilities associated with the collection of this data, the fact that a lot of the ads are
mistargeted, that consumers are opposed to their data being used in this manner, and that a large
portion of the revenue is consumed by middlemen suggests that on balance, contextual ads
provide more benefits to consumers and publishers than behaviorally-targeted ads.

1 13. Thus, many services collect analytics data to measure engagement and then use this data
2 to develop features that are likely to lead to greater levels of engagement (i.e., more time
3 spent using the service or more monetizable personal information divulged to the service).

4 14. Advertisements are targeted at users based on inferences about those users' interests.
5 Individual users' interests are inferred based on data automatically collected from them:
6 the services they use, how they use them, from where they use them, and so forth. In
7 short, online and offline activities are tracked, which allows companies to maintain
8 detailed profiles of individual user behavior, which in turn is used to predict users'
9 interests, preferences, and even demographics. The collected information may be used to
10 predict a consumer's religion, health conditions, sexual orientation, or political affiliation.
11 Some of this information may be revealed by the phone's location (e.g., where a user
12 lives, who they live with, where they work, etc.), or even by just the name of the app that
13 is being used (e.g., a Bible app revealing religion or a dating app revealing sexual
14 orientation).

15 15. For example, Meta, a NetChoice member, uses the personally-identifiable information that
16 it collects to build dossiers about users' interests, preferences, activities (both online and
17 offline), location trails, and even social relations. These dossiers are then used to
18 determine which advertisements that Meta is paid for displaying to show to which users.⁵
19 For example, when I accessed the "Ads Manager" interface to go through the steps of
20 posting a targeted advertisement on Facebook (as any advertiser would), I was given the
21 option to target an advertisement based on demographics, interests, and prior observed
22 behaviors (Figures 1–3).

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28 ⁵ Meta. "Audience Ad Targeting." *Meta Ads*, <https://www.facebook.com/business/ads/ad-targeting>. Accessed: October 22, 2024.

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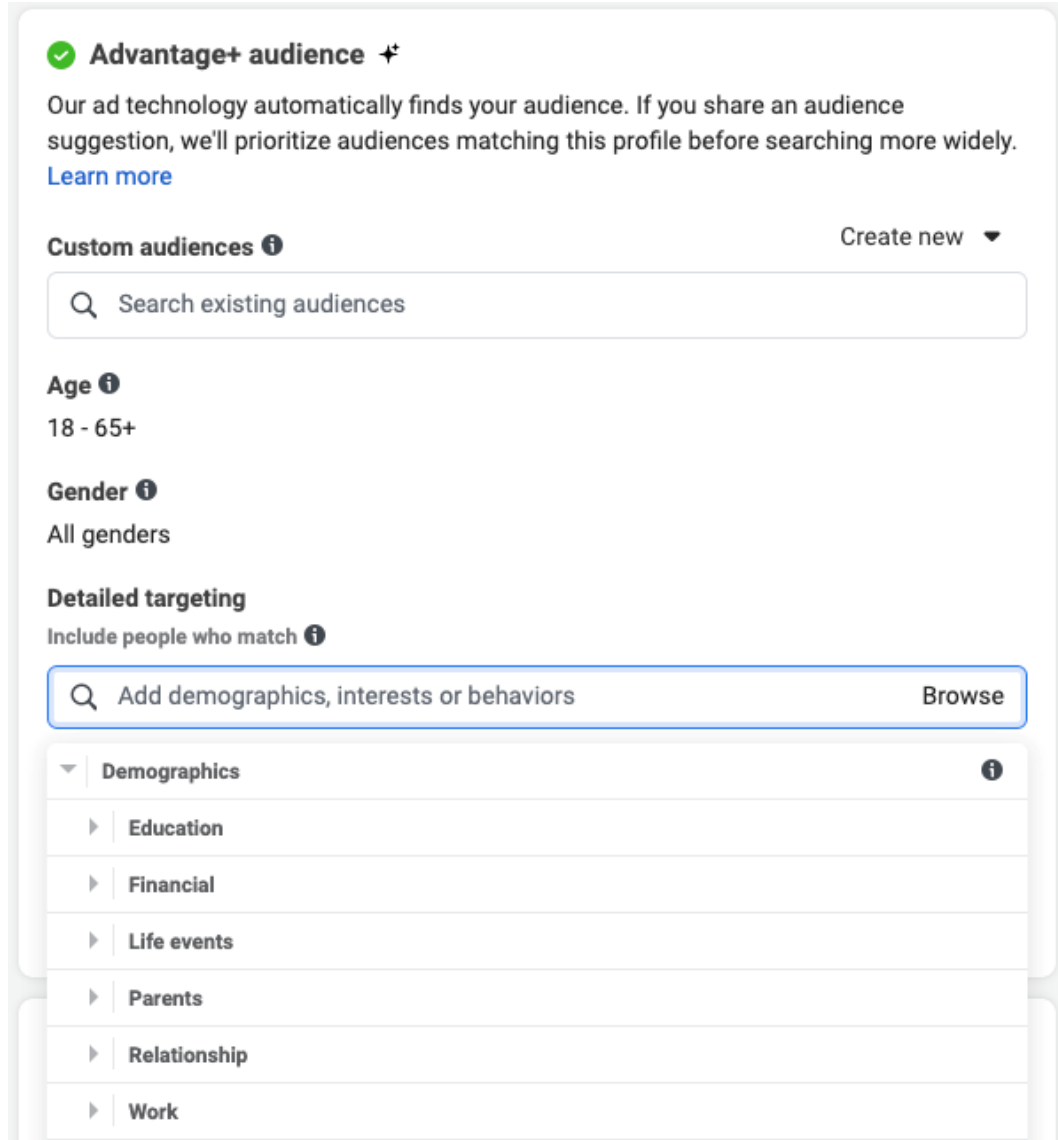


Figure 1: Facebook Ads Manager showing demographic targeting criteria.
Source: <https://adsmanager.facebook.com/>.

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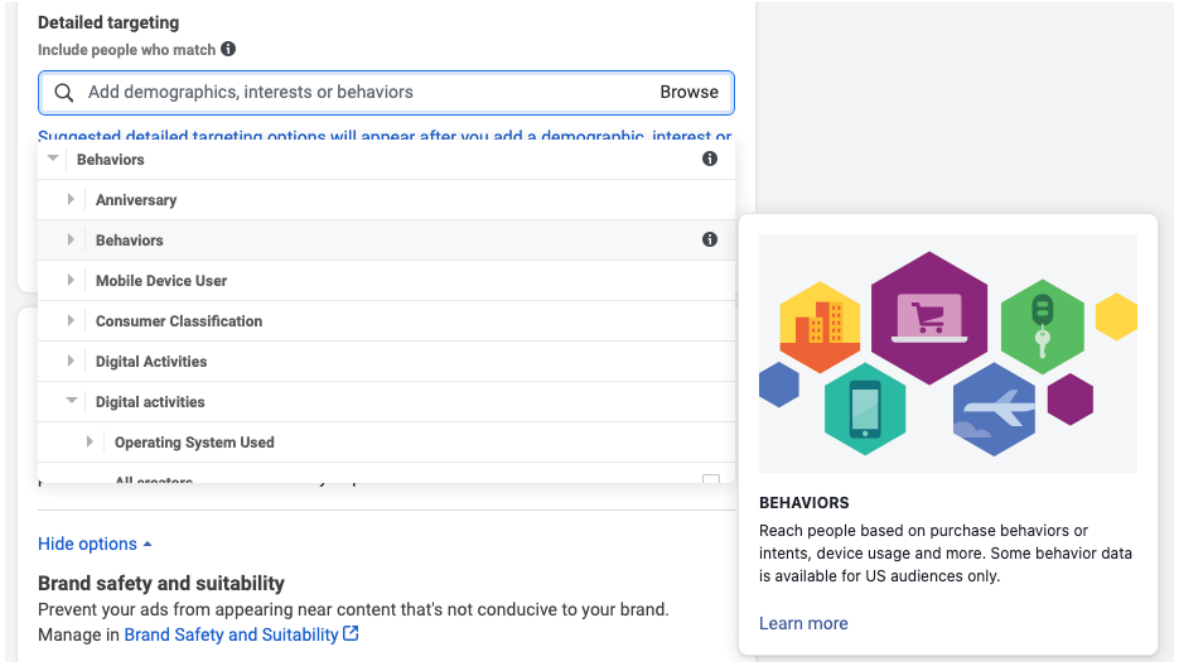


Figure 2: Facebook Ads Manager showing behavioral targeting criteria.
Source: <https://adsmanager.facebook.com/>.

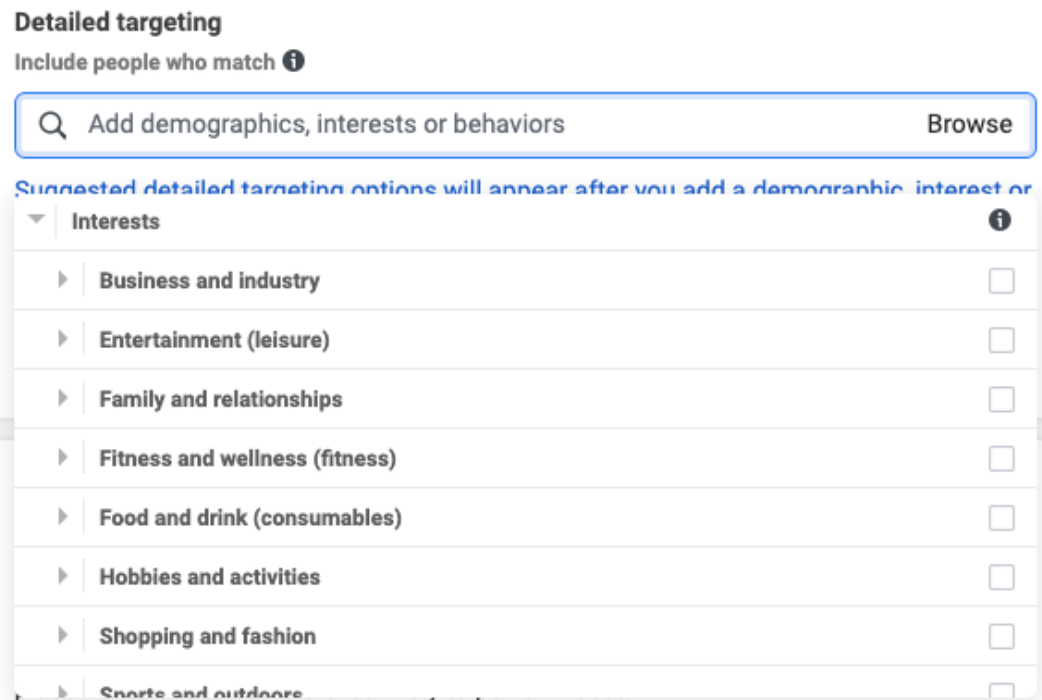


Figure 3: Facebook Ads Manager showing interest-based targeting criteria.
Source: <https://adsmanager.facebook.com/>.

1 16. Specifically, expanding these options shows very fine-grained ad targeting criteria that
2 Meta's advertisers can use to target ads at specific Facebook users.⁶ While some of these
3 data points are based on information that Facebook users have posted to Facebook, many
4 others are based on information gleaned from third parties or on metadata gleaned from
5 users' usage behaviors. In other words, despite the fact that Facebook is not in the
6 business of selling products to consumers, it allows advertisers to target ads based on
7 items that Facebook users have recently purchased from third-party websites and physical
8 stores. All together, these fine-grained ad targeting criteria include sensitive personal
9 information such as:

- 10 • Demographics
 - 11 i. Age
 - 12 ii. Gender
 - 13 iii. Geographic location
 - 14 iv. Education level
 - 15 v. Field of study (e.g., major)
 - 16 vi. School
 - 17 vii. When someone went to college
 - 18 viii. Income
 - 19 ix. Life events (e.g., anniversary, travel, birthday, new job, new relationship,
20 recent engagement, newlyweds, recently moved, etc.)
 - 21 x. Parental status (including ages of children)
 - 22 xi. Relationship status
 - 23 xii. Employer
 - 24 xiii. Field of employment
 - 25 xiv. Job title
- 26 • Interests
 - 27 i. Specific businesses and industries
 - 28 ii. Entertainment interests (e.g., games played, movies and television watched,
music interests, reading preferences, etc.)
 - iii. Family interests (e.g., parenting, marriage, etc.)
 - iv. Fitness interests (e.g., bodybuilding, exercise preferences, yoga, etc.)
 - v. Food and drink preferences (including restaurant preferences)
 - vi. Hobbies (e.g., arts and music, current events, politics, home and garden,
pets, travel, vehicles, etc.)
 - vii. Shopping and fashion
 - viii. Sports and outdoors
 - ix. Technology
- Behaviors
 - i. Recent purchases

⁶ Meta. "Facebook Ads Manager." <https://adsmanager.facebook.com/>. Accessed: October 22, 2024.

- 1 ii. Device usage/ownership
- 2 iii. Specific software used
- 3 iv. Online activities
- 4 v. Travel history
- 5 vi. Transit behaviors (e.g., commuters, users of public transit, etc.)

6 17. Online services are able to offer advertisers such fine-grained ad targeting options due to
 7 the breadth of the data they collect from individual Internet users. For example, when
 8 people create Facebook profiles and use Facebook, they share a wealth of personal
 9 information with Meta: their names, addresses, contact information, gender, preferences
 10 (e.g., via use of the “like” button and membership in affinity groups), relationship
 11 information, birthdates, and many other types of information. To quote Mark Zuckerberg
 12 on peoples’ willingness to provide Facebook sensitive information unquestioningly,
 13 “[p]eople just submitted it. I don’t know why. They ‘trust me.’ Dumb f[***]s.”⁷

14 18. Tracking of users’ online behaviors is made possible by “persistent identifiers.” An
 15 identifier is any piece of information that allows an individual—or device—to be uniquely
 16 identified. “Persistent” identifiers are identifiers that tend to not change over time.⁸ For
 17 example, motor vehicles have persistent identifiers in the form of license plates: a license
 18 plate uniquely identifies a vehicle and vehicles tend to have the same license plate over
 19 time. If someone records all the license plates at a particular place over time, they can
 20 determine how many times in that period any individual vehicle was there (and thus infer
 21 their operators’ activities). Similarly, if license plates are recorded at many different
 22 locations and that data is combined, one could reconstruct the movements of individual
 23 vehicles. Thus, combining a persistent identifier with information about where that
 24 identifier was observed (e.g., a website or mobile app) allows a data recipient to
 25 reconstruct an individual’s activities. Using this knowledge, one could infer information
 26 about a person’s routines, preferences, demographics, and even relations and social
 27 connections by tracking their persistent identifiers. It is for this reason that persistent

27 ⁷ Laura Raphael. “Mark Zuckerberg Called People Who Handed Over Their Data ‘Dumb
 28 F****.’” *Esquire*, March 19, 2018, <https://www.esquire.com/uk/latest-news/a19490586/mark-zuckerberg-called-people-who-handed-over-their-data-dumb-f/>. Accessed: October 22, 2024.

⁸ <https://www.nnlm.gov/guides/data-glossary/persistent-unique-identifier>

1 identifiers, including ones that identify personal devices—because personal devices tend
2 to be used by one individual and, just as a car’s owner is assumed to be its driver absent
3 strong evidence to the contrary, the device’s owner can be assumed to be that user—are
4 categorized as personal information under various privacy laws (e.g., CCPA,⁹ COPPA,¹⁰
5 HIPAA,¹¹ GDPR,¹² GLBA¹³).

6 19. Online advertisements need not use consumers’ personal information: while the
7 *behavioral* or *targeted* advertising described in the prior paragraphs relies on collecting
8 personal information to infer users’ interests, *contextual* advertising does not. Contextual
9 advertising refers to choosing ads based on what the user is doing in the moment. In other
10 words, it is based on the type of website or online service that the user is currently
11 visiting, which is where the ad is to appear, and not on a collected profile or tracking
12 information about that user. For example, a mattress review website does not need to
13 collect personal information to know that visitors might be receptive to ads for mattresses
14 or bedding. By definition, contextual advertising does not require the collection of
15 consumers’ personal information, because it does not rely on the tracking of their online
16 activities.

17 20. In addition to questionable economic benefits, over half a century of published research
18 on consumer behavior and preferences has demonstrated that consumers are opposed to
19 this type of tracking by businesses. For example, when Westin performed consumer
20 surveys on public privacy perceptions going back to the 1970s,¹⁴ he consistently found
21 that a majority of the U.S. public are either “very” or “somewhat” concerned with how
22 their personal information is collected and used by businesses. In 2001, one study found

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24 ⁹ Cal. Civ. Code § 1798.140(15).

25 ¹⁰ 15 U.S.C § 6501(8)(F).

26 ¹¹ 45 C.F.R. § 164.514(b)(2)(i).

27 ¹² GDPR Art. 4 (1).

28 ¹³ 16 C.F.R. § 313.3. See also, e.g.,

<https://www.federalregister.gov/documents/2021/12/09/2021-25736/standards-for-safeguarding-customer-information>

¹⁴ Ponnurangam Kumaraguru and Lorrie Faith Cranor. “Privacy indexes: a survey of Westin's studies.” *Carnegie Mellon University Tech Report CMU-ISRI-5-138*, 2005.

1 that as many as 64% of consumers refused to shop online due to privacy concerns.¹⁵ A
2 Pew survey from 2020 found that more than half of Americans have refused to use certain
3 products or services due to privacy concerns.¹⁶ In the past two decades, as more and more
4 aspects of daily life have moved online, many consumers have also simply become
5 resigned to having their information used in objectionable ways.¹⁷ A 2019 Pew survey of
6 consumers found that 62% of Americans do not believe it is possible to “go through daily
7 life without companies collecting data about them,” 79% are very or somewhat concerned
8 about this, and 81% believe the risks of collecting this data outweigh the benefits.¹⁸

9 21. While consumers are overwhelmingly opposed to this type of tracking and the profiling
10 and resale of their information that it supports (one study of U.S. consumers found that up
11 to 86% do not want ads that are tailored based on their online activities),¹⁹ consumers
12 nonetheless continue to engage with services that appear to conflict with their stated
13 privacy preferences. This is known as the “privacy paradox.” Some stakeholders like to
14 point out this disconnect and use it to disingenuously claim that it means that consumers
15 do not “really” care about privacy. But the published research on the privacy paradox
16 demonstrates that this argument is incorrect, and that there are several rational
17 explanations for the privacy paradox, which include lack of awareness of data collection
18 methods, poor usability, mismatched incentives, and perceived lack of agency.

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21 ¹⁵ M J. Culnan and Milne, G. R. “The Culnan-Milne Survey on Consumers & Online
22 Privacy Notices: Summary of Responses.” In *Interagency Public Workshop (Ed.) Get Noticed:
23 Effective Financial Privacy Notices*, Washington, D.C., 2001.

24 ¹⁶ Andrew Perrin, “Half of Americans have decided not to use a product or service
25 because of privacy concerns.” *Pew Research Center*, August 14, 2020.
26 <https://www.pewresearch.org/fact-tank/2020/04/14/half-of-americans-have-decided-not-to-use-a-product-or-service-because-of-privacy-concerns/>

27 ¹⁷ Nora A. Draper and Joseph Turow. “The corporate cultivation of digital
28 resignation.” *New media & society* 21.8 (2019): 1824-1839.

¹⁸ Pew Research Center. “Americans and Privacy: Concerned, Confused and Feeling Lack
of Control Over Their Personal Information.” Nov. 15, 2019.

<https://www.pewresearch.org/internet/2019/11/15/americans-and-privacy-concerned-confused-and-feeling-lack-of-control-over-their-personal-information/>

¹⁹ J. Turow, J. King, C. J. Hoofnagle, A. Bleakley, and M. Hennessy (2009). “Americans
Reject Tailored Advertising and Three Activities That Enable It.”
<https://doi.org/10.2139/ssrn.1478214>

1 22. In many cases, consumers simply do not understand when they are making decisions that
2 will impact their privacy. For example, in a series of studies that I co-authored,²⁰ we
3 presented subjects with different search engine interfaces, including one that annotated
4 search results with privacy information; subjects were instructed to use the search engine
5 to buy items from merchants of their choice. While all subjects expressed strong privacy
6 preferences in a survey administered prior to the study (i.e., subjects were specifically
7 screened for strong privacy preferences, so that we could explicitly test whether interface
8 design impacted their ability to act on those preferences), we observed that without
9 information about privacy practices presented in an easily-accessible manner, subjects
10 made purchases from the cheapest merchants. When search results were annotated with
11 privacy ratings, subjects were significantly more likely to make purchases from merchants
12 with more agreeable privacy policies (i.e., better aligned with participants' stated privacy
13 preferences), even paying more money to do so. These and other studies demonstrate that
14 people often act in ways that seem contrary to their stated privacy preferences when they
15 are not fully aware of a business's privacy practices (e.g., due to well-documented
16 problems with the "notice and consent" framework, such as expecting consumers to read
17 and understand privacy policies, which I describe in subsequent sections).

18 23. In other cases, convoluted user interfaces make it difficult for consumers to understand
19 how to make privacy-protective decisions. This poor usability often results in consumers
20 sharing personal information without ever being aware of it. For example, while studies
21 have shown that consumers have concerns about sharing personal information with the
22 wrong audiences on social media, they nonetheless continue to overshare,²¹ which has

23 ²⁰ Janice Y. Tsai Serge Egelman, Lorrie Cranor, and Alessandro Acquisti. "The effect of
24 online privacy information on purchasing behavior: An experimental study." *Information systems
25 research* 22, no. 2 (2011): 254-268; Serge Egelman, Janice Tsai, Lorrie Faith Cranor, and
26 Alessandro Acquisti. "Timing is everything? The effects of timing and placement of online
27 privacy indicators." In *Proceedings of the SIGCHI Conference on Human Factors in Computing
28 Systems*, pp. 319-328. 2009; Julia Gideon, Lorrie Cranor, Serge Egelman, and Alessandro
Acquisti. "Power strips, prophylactics, and privacy, oh my!." In *Proceedings of the Second
Symposium on Usable privacy and security*, pp. 133-144. 2006.

²¹ Maritza Johnson, Serge Egelman, and Steven M. Bellovin. 2012. "Facebook and
privacy: it's complicated." In *Proceedings of the Eighth Symposium on Usable Privacy and*

(continued...)

1 been shown to be the result of difficult-to-use privacy settings interfaces (or mismatches
 2 between the design of those interfaces and users' mental models).²² One early study on the
 3 use of Facebook found that while participants expressed strong privacy preferences, they
 4 nonetheless shared sensitive information because more than one-in-five did not
 5 understand what Facebook's privacy settings did or how to use them, and therefore did not
 6 change them from the overly-permissive defaults.²³ In a study of file-sharing software,
 7 researchers discovered that due to convoluted privacy settings interfaces, many users were
 8 inadvertently sharing their entire hard drives.²⁴ In a study of tools provided by the
 9 advertising industry to opt out of behavioral advertising on websites, the researchers
 10 observed:

11 “Participants found many tools difficult to configure, and tools' default settings were often
 12 minimally protective. Ineffective communication, confusing interfaces, and a lack of
 13 feedback led many participants to conclude that a tool was blocking [online behavioral
 14 advertising] when they had not properly configured it to do so. Without being familiar with
 15 many advertising companies and tracking technologies, it was difficult for participants to
 16 use the tools effectively.”²⁵

- 17 24. Incentives are also important when studying privacy tradeoffs. Privacy decisions are not
 18 made in a vacuum: that consumers engage with services that violate their privacy
 19 preferences is often an indictment of the lack of market choice rather than an indication
 20 that consumers are behaving hypocritically. Similarly, privacy is often not the only

21 Security (SOUPS '12). Association for Computing Machinery, New York, NY, USA, Article 9,
 22 1–15. <https://doi.org/10.1145/2335356.2335369>

23 ²² Jennifer King, Airi Lampinen, and Alex Smolen. 2011. “Privacy: is there an app for
 24 that?” In *Proceedings of the Seventh Symposium on Usable Privacy and Security (SOUPS '11)*.
 25 Association for Computing Machinery, New York, NY, USA, Article 12, 1–20.
 26 <https://doi.org/10.1145/2078827.2078843>

27 ²³ Alessandro Acquisti and Ralph Gross. “Imagined communities: Awareness, information
 28 sharing, and privacy on the Facebook.” In *Privacy Enhancing Technologies: 6th International
 Workshop, PET 2006, Cambridge, UK, June 28-30, 2006, Revised Selected Papers 6*, pp. 36-58.
 Springer Berlin Heidelberg, 2006.

29 ²⁴ Nathaniel S. Good and Aaron Krekelberg. 2003. “Usability and privacy: a study of
 30 Kazaa P2P file-sharing.” In *Proceedings of the SIGCHI Conference on Human Factors in
 31 Computing Systems (CHI '03)*. Association for Computing Machinery, New York, NY, USA,
 32 137–144. <https://doi.org/10.1145/642611.642636>

33 ²⁵ Pedro Leon, Blase Ur, Richard Shay, Yang Wang, Rebecca Balebako, and Lorrie
 34 Cranor. “Why Johnny can't opt out: a usability evaluation of tools to limit online behavioral
 35 advertising.” In *Proceedings of the SIGCHI conference on human factors in computing systems*,
 36 pp. 589-598. 2012.

1 consideration: if the costs of protecting one’s privacy are unreasonably high (e.g., time
2 invested learning to correctly use privacy settings, monetary costs, abstaining from social
3 life, etc.), many consumers will engage with privacy-violative services because they
4 cannot afford the alternatives. For example, I value my free time, but that I still show up
5 to work does not make me a hypocrite. Similarly, when faced with the choice between
6 protecting their privacy or engaging with their peers online, many younger people will
7 choose the latter, despite the known privacy risks. Many studies have shown that despite
8 the known privacy risks, many young people continue to use social media due to the fear
9 of missing out.²⁶

10 25. Finally, many consumers simply do not believe they have agency when it comes to
11 making online privacy decisions: because many believe that their privacy preferences will
12 not be honored no matter the actions that they take, many choose to engage with privacy-
13 violative services to extract benefits, believing that they will end up paying the privacy
14 costs regardless. A 2015 consumer survey concluded the following:

15 “[A] majority of Americans are resigned to giving up their data—and that is why many
16 appear to be engaging in tradeoffs. Resignation occurs when a person believes an
17 undesirable outcome is inevitable and feels powerless to stop it. Rather than feeling able to
18 make choices, Americans believe it is futile to manage what companies can learn about
19 them. Our study reveals that more than half do not want to lose control over their
20 information but also believe this loss of control has already happened.”²⁷

19 26. A study specifically on young people and the privacy paradox observed:

20 “Based on focus group interviews, we considered how young adults’ attitudes about privacy
21 can be reconciled with their online behavior. The “privacy paradox” suggests that young
22 people claim to care about privacy while simultaneously providing a great deal of personal

22 ²⁶ Vittoria Franchina, Mariek Vanden Abeele, Antonius J. Van Rooij, Gianluca Lo Coco,
23 and Lieven De Marez. “Fear of missing out as a predictor of problematic social media use and
24 phubbing behavior among Flemish adolescents.” *International journal of environmental research
25 and public health* 15, no. 10 (2018): 2319; Dmitri Rozgonjuk, Cornelia Sindermann, Jon D. Elhai,
26 and Christian Montag. “Fear of Missing Out (FoMO) and social media’s impact on daily-life and
27 productivity at work: Do WhatsApp, Facebook, Instagram, and Snapchat Use Disorders mediate
28 that association?.” *Addictive Behaviors* 110 (2020): 106487; Ine Beyens, Eline Frison, and Steven
29 Eggermont. “‘I don’t want to miss a thing’: Adolescents’ fear of missing out and its relationship
30 to adolescents’ social needs, Facebook use, and Facebook related stress.” *Computers in Human
31 Behavior* 64 (2016): 1-8.

32 ²⁷ Joseph Turow, Michael Hennessy, and Nora Draper. “The tradeoff fallacy: How
33 marketers are misrepresenting American consumers and opening them up to
34 exploitation.” *Available at SSRN 2820060* (2015).

1 information through social media. Our interviews revealed that young adults do understand
 2 and care about the potential risks associated with disclosing information online and engage
 3 in at least some privacy-protective behaviors on social media. However, they feel that once
 4 information is shared, it is ultimately out of their control. They attribute this to the opaque
 5 practices of institutions, the technological affordances of social media, and the concept of
 6 networked privacy, which acknowledges that individuals exist in social contexts where
 7 others can and do violate their privacy.”²⁸

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27. Similarly, users continue to use apps that they find “creepy” due to a sense of learned helplessness: they do not believe that they have the power to control who receives their personal information when they participate in the digital economy.²⁹

TOOLS FOR LIMITING COLLECTION & USE OF PERSONAL INFORMATION

28. **Privacy Policies.** Internet users have few tools to control their online privacy. Since the dawn of the Internet age, the primary framework for managing online privacy has been the “notice and consent” framework, whereby online services post privacy policies (“notice”) and consumers can choose whether to engage with services based on their understanding of those policies (“consent”). Unfortunately, this framework is fundamentally detached from reality: decades of research have demonstrated that consumers do not read these privacy policies, consumers do not understand what the policies mean (when they do read them), and worse, privacy policies often do not accurately describe their services’ behaviors.

29. In one study in which participants were asked to explicitly confirm that they read and agreed to a website’s privacy policy, 80% clicked a box to affirm that they had done so despite not actually accessing or reading the policy.³⁰ This number likely represents a lower bound, given the presence of “demand characteristics” (i.e., participants were in a

²⁸ Eszter Hargittai, and Alice Marwick. “‘What can I really do?’ Explaining the privacy paradox with online apathy.” *International journal of communication* 10 (2016): 21.

²⁹ Irina Shklovski, Scott D. Mainwaring, Halla Hrunnd Skúladóttir, and Höskuldur Borgthorsson. “Leakiness and creepiness in app space: perceptions of privacy and mobile app use.” In *Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '14)*. Association for Computing Machinery, New York, NY, USA, 2347–2356. <https://doi.org/10.1145/2556288.2557421>

³⁰ Nili Steinfeld. “‘I agree to the terms and conditions’: (How) do users read privacy policies online? An eye-tracking experiment.” *Computers in Human Behavior* 55 (2016): 992-1000.

1 laboratory setting and therefore were likely to pay more attention to the instructions than
2 they would have in the real world), as well as the fact that most online services do not
3 present users with interstitial messages demanding that they read and agree to their
4 privacy policies: most privacy policies are accessed through discreet links outside the
5 user's field of focus. Another study found that privacy-concerned users were influenced
6 by the mere presence of a privacy policy link, despite few reading the policies.³¹ This
7 suggests that the mere presence of a privacy policy erroneously signals "good" privacy
8 practices.

9 30. Nonetheless, if users do opt to read privacy policies, it is often a significant time
10 investment. In 2008, McDonald and Cranor showed that if users read the privacy policies
11 for every website they accessed, they would need to spend up to 300 hours doing so
12 annually (based on average policy lengths, number of websites visited, and reading
13 speeds).³² Of course, their estimate is based on data from 2008 that showed the average
14 Internet user visits around 1,500 unique websites annually; 15 years later, the number of
15 websites has proliferated, as has the amount of time that consumers spend online, which
16 suggests that the time investment to read and understand privacy policies has only
17 increased.

18 31. It is also not clear that the time investment to read privacy policies is worthwhile for most
19 consumers: several studies have shown that the privacy policies found on popular
20 websites are written at the college level and therefore may not be understood by a
21 significant proportion of the population (much less children).³³ This also ignores the fact

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23 ³¹ Jensen, Carlos, Colin Potts, and Christian Jensen. "Privacy practices of Internet users:
Self-reports versus observed behavior." *International Journal of Human-Computer Studies* 63,
no. 1-2 (2005): 203-227.

24 ³² Aleecia M. McDonald and Lorrie Faith Cranor. "The cost of reading privacy policies."
I/S: A Journal of Law and Policy for the Information Society, 4 (2008): 543.

25 ³³ Yuanxiang Li *et al.* "Online privacy policy of the thirty Dow Jones corporations:
Compliance with FTC Fair Information Practice Principles and readability assessment."
26 *Communications of the IIMA* 12.3 (2012): 5; Carlos Jensen and Colin Potts. "Privacy policies as
27 decision-making tools: an evaluation of online privacy notices." *Proceedings of the SIGCHI
Conference on Human Factors in Computing Systems*. 2004; George R. Milne, Mary J. Culnan,
28 and Henry Greene. "A longitudinal assessment of online privacy notice readability." *Journal of
Public Policy & Marketing* 25.2 (2006): 238-249.

1 that such privacy policies are subject to change at any time without notice, and thus the
2 time investment may all be for naught.

3 32. Even when policies are noticed, read, and understood, they generally do not explain a
4 service’s data practices in sufficient detail for consumers to make informed decisions. For
5 example, despite various laws requiring that services post privacy policies, there are rarely
6 requirements that force those services to name the specific third parties with whom they
7 share data—they are usually only required to specify the broad categories of data
8 recipients. Even though those third parties may have their own data practices that are
9 documented in their own privacy policies, it is nearly impossible for consumers to inform
10 themselves about those practices if they are unable to locate those additional privacy
11 policies because they do not know the identities of the companies. Similarly, it is nearly
12 impossible for consumers to understand the privacy practices of large companies that offer
13 multiple services, as their privacy policies are often written in a manner that aggregates
14 their practices across all of their offered services (e.g., Google’s privacy policy,³⁴ which
15 has nearly 6,000 words and is written at a college reading level, describes their data
16 collection practices across all of their services and does not convey what data may be
17 collected by Google Maps vs. Gmail vs. Docs vs. Search).

18 33. **Blocking Cookies and Fingerprinting.** In addition to reading privacy policies, there are
19 some technologies that consumers can use in futile attempts to better protect their privacy.
20 “Cookies” are data that websites store in consumers’ web browsers, which are then
21 transmitted back to websites when visited in the future. This allows a website to recognize
22 a user over time, without having to log in again (as well as allowing the website to
23 “remember” other settings, such as a default language). Because cookies have been
24 historically abused for invasive tracking and profiling,³⁵ modern web browser software
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26 _____
27 ³⁴ <https://policies.google.com/privacy?hl=en-US>

28 ³⁵ J. R. Mayer and J. C. Mitchell, “Third-Party Web Tracking: Policy and Technology,”
2012 IEEE Symposium on Security and Privacy, San Francisco, CA, USA, 2012, pp. 413-427,
doi: 10.1109/SP.2012.47.

1 allows users to delete stored cookies or to block cookies set by third-party trackers
2 altogether.

3 34. However, deleting or blocking cookies is no longer an effective strategy, as tracking now
4 occurs using other means that consumers cannot control.³⁶ For example, unique
5 “fingerprints”—the aggregation of several data points to create a unique identifier—can
6 be constructed based on seemingly-benign information that is automatically transmitted to
7 online services without user consent: software versions (e.g., the web browser and
8 operating system), language settings, time zones, screen resolution, battery levels, etc.³⁷
9 Even what fonts are installed on a computer, which are available to websites, can be used
10 to uniquely identify a website visitor.³⁸ Apps on mobile devices have additional data
11 points available for constructing unique fingerprints to identify their users, all without the
12 use of cookies and with few actions that users can take to prevent this from occurring.
13 Perversely, whether a user has configured privacy settings away from the defaults is often
14 used as a data point for further tracking (i.e., while some web browsers can transmit a
15 user-configurable “do not track” signal to websites, many websites choose not to honor
16 this and instead use it as another way to identify and track users).³⁹

17 35. Every device connected to the Internet has an Internet Protocol (IP) address, which is used
18 to route information to and from it. While IP addresses must be transmitted to send and

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20 ³⁶ N. Nikiforakis, A. Kapravelos, W. Joosen, C. Kruegel, F. Piessens and G. Vigna,
21 “Cookieless Monster: Exploring the Ecosystem of Web-Based Device Fingerprinting,” *2013*
22 *IEEE Symposium on Security and Privacy*, Berkeley, CA, USA, 2013, pp. 541-555, doi:
10.1109/SP.2013.43; R. Upathilake, Y. Li, and A. Matrawy, “A classification of web browser
23 fingerprinting techniques,” *2015 7th International Conference on New Technologies, Mobility*
24 *and Security (NTMS)*, Paris, France, 2015, pp. 1-5, doi: 10.1109/NTMS.2015.7266460.

³⁷ See, e.g., <https://amiunique.org/>; Peter Eckersley. “How unique is your web browser?”
25 In *Privacy Enhancing Technologies: 10th International Symposium, PETS 2010, Berlin,*
26 *Germany, July 21-23, 2010. Proceedings 10*, pp. 1-18. Springer Berlin Heidelberg, 2010;

³⁸ David Fifield and Serge Egelman. “Fingerprinting web users through font metrics.”
27 *Financial Cryptography and Data Security: 19th International Conference, FC 2015, San Juan,*
28 *Puerto Rico, January 26-30, 2015, Revised Selected Papers 19*. Springer Berlin Heidelberg, 2015.

³⁹ Geoffrey A. Fowler, “Think you’re anonymous online? A third of popular websites are
‘fingerprinting’ you.” *The Washington Post*, October 31, 2019.
[https://www.washingtonpost.com/technology/2019/10/31/think-youre-anonymous-online-third-
popular-websites-are-fingerprinting-you/](https://www.washingtonpost.com/technology/2019/10/31/think-youre-anonymous-online-third-popular-websites-are-fingerprinting-you/); Michael Simon, “Apple is removing the Do Not Track
toggle from Safari, but for a good reason.” *Macworld*, February 6, 2019.
<https://www.macworld.com/article/232426/apple-safari-removing-do-not-track.html>.

1 receive data, they can also be used to track users over time. Since devices behind a
2 firewall (e.g., a household WiFi router) will appear to the outside world to share the same
3 IP address, the collection of IP addresses is often used as a way of performing “cross-
4 device tracking,” which allows data recipients to infer when the same individual has
5 moved from using a mobile device to a desktop computer to a smart TV; it also allows
6 data recipients to infer when multiple individuals reside within the same household. For
7 example, Meta’s privacy policy states that they collect “information about the network
8 you connect your device to, including your IP address” to target advertisements and
9 provide “business services” to unnamed partners.⁴⁰ There is little that consumers can do to
10 prevent this, without substantially degrading their online experiences. Worse, there is no
11 way for consumers to know when this type of tracking is even occurring.

12 36. **Machine-Readable Privacy Policies.** Over 20 years ago, due to the privacy concerns
13 regarding cookies, online tracking, and the acknowledgement that natural language
14 privacy policies are woefully inadequate, several proposals were put forth to create
15 machine-readable privacy policies. The idea behind these proposals was that consumers
16 could use an interface to save their privacy preferences within their web browsers (or
17 other software under their control), websites could post machine-readable policies, and
18 then web browsers could act on consumers’ behalf to either alert them when encountering
19 a website with a disagreeable privacy policy (determined by the browser’s automatic
20 parsing of a website’s machine-readable policy), or take some other action (e.g.,
21 automatically negotiating a better policy, blocking cookies or other transmissions, etc.).
22 One of these proposals became a web standard: the Platform for Privacy Preferences
23 Project (P3P),⁴¹ was a web standard developed by the World Wide Web Consortium. (I
24 served on the standards committee as an invited expert.)

25 37. The P3P standard gained traction, with many industry stakeholders adopting it by posting
26 “P3P policies” on their websites so that web browsers could automatically parse them and

27
28 ⁴⁰ <https://www.facebook.com/privacy/policy/>

⁴¹ <https://en.wikipedia.org/wiki/P3P>

1 alert users when they encountered websites that violated those users' stated privacy
 2 preferences. Microsoft's Internet Explorer (IE) browser was the first major web browser
 3 to adopt P3P, and by default, IE would block third-party tracking cookies unless the
 4 website posted a P3P policy (and then would block third-party cookies in accordance with
 5 the user's stated privacy preferences). In response, many companies (e.g., Amazon,
 6 Facebook, and Google; all NetChoice members) posted P3P policies that did not actually
 7 describe their privacy practices, but nonetheless tricked the IE browser into accepting their
 8 tracking cookies, due to the presence of a valid P3P header.⁴² One study of over 33,000
 9 websites observed that more than one third were transmitting P3P policies that appeared to
 10 be designed to circumvent IE's cookie blocking (and did not accurately describe their
 11 sites' actual privacy practices).⁴³ (The same study found that many of these websites were
 12 certified participants in TRUSTe's⁴⁴ EU Safe Harbor industry self-regulation program,
 13 and concluded that such certified sites were no more likely to comply with the P3P
 14 standard than websites not certified.) Some of these P3P policies can still be found today
 15 when accessing the websites that include trackers from NetChoice members.⁴⁵ For
 16 example, as of March 28, 2023, Google Ads⁴⁶ transmits a P3P policy header, but the body
 17 of the policy is as follows:

18 CP="This is not a P3P policy! See g.co/p3phelp for more info."

19 38. Thus, I have come to the conclusion that voluntary online standards that aim to give
 20 consumers more control over their privacy are futile, as they are likely to be coopted by
 21 the companies that profit.

22 ⁴² Lorrie Faith Cranor, "Necessary but not sufficient: Standardized mechanisms for
 23 privacy notice and choice." *J. on Telecomm. & High Tech. L.* 10 (2012): 273.

24 ⁴³ Pedro Giovanni Leon, Lorrie Faith Cranor, Aleecia M. McDonald, and Robert
 25 McGuire. "Token attempt: the misrepresentation of website privacy policies through the misuse
 of p3p compact policy tokens." In *Proceedings of the 9th annual ACM workshop on Privacy in
 the electronic society (WPES '10)*. Association for Computing Machinery, New York, NY, USA,
 93–104. <https://doi.org/10.1145/1866919.1866932>

26 ⁴⁴ TRUSTe is now known as "TrustArc."

27 ⁴⁵ Lorrie Faith Cranor, "Internet Explorer privacy protections also being circumvented by
 Google, Facebook, and many more." *Technology Academics Policy*, February 18, 2021.

28 https://www.techpolicy.com/Cranor_InternetExplorerPrivacyProtectionsBeingCircumvented-by-Google.aspx

⁴⁶ <https://adservice.google.com/adsid/google/ui>

SPECIAL CONCERNS REGARDING CHILDREN’S PRIVACY

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2 39. Data monetization without appropriate oversight is even more concerning when the data
3 comes from children, who are unlikely to understand that this is happening, much less
4 consent to it, but who could potentially face enormous impacts due to future usage of this
5 data. This data may be used for manipulative marketing campaigns, but also may feed
6 biased and unaccountable algorithms that use it to make decisions about a child’s future,
7 not to mention outright malicious uses of the data (e.g., non-custodial parents purchasing
8 location data to geolocate a child).

9 40. In 2016 my research team decided to look at how well mobile apps directed at children
10 appeared to be complying with COPPA, which has been in effect since 2000. We wrote
11 bespoke instrumentation for the Android platform that allows us to run mobile apps and
12 monitor exactly what personal information those apps access and with whom they share
13 it.⁴⁷ We also used our instrumentation to determine whether transmissions containing
14 personal information were performed securely and confidentially.

15 41. Starting in late 2016, we began downloading as many free apps in the “Designed for
16 Families” (DFF) program as we could find, which ended up being just under 6,000 apps.⁴⁸
17 The DFF program is a section of the Play Store, Google’s centralized Android app market,
18 which is exclusively for apps that are directed to children. Mobile app developers must

19 ⁴⁷ P. Wijesekera, A. Baokar, A. Hosseini, S. Egelman, D. Wagner, and K. Beznosov.
20 “Android permissions remystified: A field study on contextual integrity.” In *Proceedings of the*
21 *24th USENIX Security Symposium (USENIX Security 15)*, pages 499–514, Washington, D.C.,
22 Aug. 2015. USENIX Association; P. Wijesekera, A. Baokar, L. Tsai, J. Reardon, S. Egelman, D.
23 Wagner, and K. Beznosov. “The feasibility of dynamically granted permissions: aligning mobile
24 privacy with user preferences.” In *Proceedings of the 2017 IEEE Symposium on Security and*
25 *Privacy*, Oakland ’17. IEEE Computer Society, 2017; P. Wijesekera, J. Reardon, I. Reyes, L.
26 Tsai, J.-W. Chen, N. Good, D. Wagner, K. Beznosov, and S. Egelman. “Contextualizing privacy
27 decisions for better prediction (and protection).” In *Proceedings of the 2018 CHI Conference on*
28 *Human Factors in Computing Systems*, CHI ’18, pages 1–13, New York, NY, USA, 2018.
Association for Computing Machinery; J. Reardon, A. Feal, P. Wijesekera, A. E. B. On, N.
Vallina-Rodriguez, and S. Egelman. “50 Ways to Leak Your Data: An Exploration of Apps’
Circumvention of the Android Permissions System.” In *Proceedings of the 24th USENIX Security*
Symposium, USENIX Security ’19, Berkeley, CA, USA, 2019. USENIX Association; We wrote
our tools for Google’s Android platform only because it is open source: having the source code
for the operating system allowed us to modify it for this purpose; at the time, we didn’t look at
Apple’s iOS simply because we didn’t have the source code to add the same level of
instrumentation.

⁴⁸ Reyes *et al.*, *supra* note 3.

1 participate in the program when they upload their app and disclose to Google that it is
2 directed at children. As part of the program, they must affirm to Google that their app is in
3 compliance with COPPA. Our goal was to evaluate whether that appeared to be the case
4 in practice.

5 42. Of the child-directed apps that we tested, more than half appeared to be violating COPPA
6 in one way or another: 5% collected location or other contact information and 19%
7 collected personal information without verifiable parental consent and shared it with third
8 parties whose public disclosures indicated they would use them for prohibited purposes
9 (e.g., behavioral advertising); 40% transmitted personal information insecurely.
10 Separately, 39% appeared to be violating Google’s platform policies (i.e., an example of
11 industry self-regulation) surrounding the collection of persistent identifiers for advertising
12 and analytics purposes.⁴⁹

13 43. We also examined mobile apps that had been certified by the COPPA Safe Harbor
14 programs, meaning that the app developer claimed to participate in a private FTC-
15 approved compliance-certification program.⁵⁰ (We found it extraordinarily difficult to
16 identify which mobile apps had actually been certified; none of the programs we contacted
17 were willing to share lists of apps with us, and most of their websites did not provide this
18 information.) Of the 237 apps we found that claimed to be Safe Harbor certified, 64%
19 appeared to violate Google’s policies on transmitting identifiers for advertising/analytics
20 purposes, 33% transmitted personal information to prohibited third parties, and 32%
21 transmitted personal information insecurely. We concluded that the apps that we
22 examined, which claimed to be certified as COPPA-compliant by Safe Harbor programs,
23 were no more likely to protect children’s personal information than apps that had not been
24 certified by these programs.⁵¹ (This result is consistent with prior research on adverse
25 selection in industry self-regulatory certification programs.)⁵²

26 ⁴⁹ *Ibid.*

27 ⁵⁰ 16 C.F.R. § 312.11.

28 ⁵¹ Reyes *et al.*, *supra* note 3.

⁵² Benjamin Edelman. “Adverse selection in online ‘trust’ certifications.” In *Proceedings of the 11th International Conference on Electronic Commerce*, pp. 205-212. 2009.

1 44. Thus, based on this research, I have come to the conclusion that voluntary industry self-
2 regulatory programs are ineffective and do not lead to better outcomes for consumers.

3 45. Similarly, through this research, I identified several additional gaps in regulation (beyond
4 the inadequacy of the Safe Harbor programs), that I recommended be fixed in my U.S.
5 Senate testimony.⁵³ Particularly relevant here are COPPA’s “internal operations”
6 exemption⁵⁴ and “actual knowledge” standard.⁵⁵ (These and other recommendations were
7 recently published in a law review article.)⁵⁶

8 46. Generally, websites and other online services must obtain verifiable parental consent
9 before disclosing children’s personal information to third parties, unless it is to support the
10 service’s internal operations and is not used for any other purpose. However, from a
11 technical standpoint, most internal operations do not strictly require the collection of
12 persistent identifiers that can be used to track children’s activities across different
13 services. In fact, both major platforms provide guidelines on how software developers can
14 perform these activities *without* collecting advertising identifiers or non-resettable device
15 identifiers.⁵⁷ For example, by definition, “contextual advertising” involves showing
16 consumers ads *without* using data previously collected about them, and therefore no
17 personal information is needed to show contextual ads. To prevent one user from being
18 shown the same ad repeatedly (known as “frequency capping”), a session-based or
19 installation-based identifier should be used, such that the collected data cannot be used to
20 track the user across other services.

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22
23 ⁵³ U.S. Congress. Hearing of the Subcommittee on Consumer Protection, Product Safety,
24 and Data Security of the Committee on Commerce, Scient, and Transportation. Hearing on
25 “Protecting Kids Online: Internet Privacy and Manipulative Marketing.” Testimony of Serge
26 Egelman, 2021. [https://www.commerce.senate.gov/services/files/0DC78E9D-88B2-4D54-8F4A-
27 AE7B4C7D0EF6](https://www.commerce.senate.gov/services/files/0DC78E9D-88B2-4D54-8F4A-AE7B4C7D0EF6)

28 ⁵⁴ 15 U.S.C. § 6501(4)(A).

⁵⁵ 15 U.S.C. § 6501(4)(B).

⁵⁶ Egelman, S., 2023. “Informing Future Privacy Enforcement by Examining 20+ Years of
COPPA.” *Harvard Journal of Law & Technology*, 37(3).

⁵⁷ Google, “Best Practices for Unique Identifiers.” April 6, 2023.
<https://developer.android.com/training/articles/user-data-ids>; Apple, “User Privacy and Data
Use.” 2023. <https://developer.apple.com/app-store/user-privacy-and-data-use/>.

1 47. Nonetheless, in the course of my research, I have noticed that many privacy policies
2 associated with child-directed services use the phrase “internal operations” when
3 describing the flow of children’s personal information to third parties. In many of these
4 cases, these third parties are advertisers whose public disclosures indicate that they may
5 use the data for COPPA-prohibited purposes. Thus, I have concluded that for many
6 developers, the phrase “internal operations” appears to be a shibboleth used to justify
7 privacy-invasive practices.

8 48. Secondly, COPPA’s “actual knowledge” standard, by which it must be shown that an
9 individual within these third-party organizations knew that they received data from
10 children under 13, incentivizes data recipients to simply look the other way if and when
11 they receive children’s personal information, even when those third-party transmissions
12 also include the names of the apps or websites that are transmitting them the data. Many
13 of these data recipients are advertising and/or analytics companies that publicly advertise
14 their abilities to target ads based on inferring the demographics of the users of the services
15 sending them data. Furthermore, there are many commercial services that purport to
16 provide the target demographics of a given mobile app or a website, and thus determining
17 whether or not a service is directed at children is readily ascertainable.

18 49. For example, ironSource is a targeted advertising company that we observed receiving
19 personal information from child-directed apps.⁵⁸ Their privacy policy stated they did not
20 knowingly receive personal information from children under 13, a point which was
21 reiterated to my laboratory in a letter from their general counsel.⁵⁹ In my response, I
22 pointed out that all developers wishing to use ironSource’s services must provide a
23 company name at sign-up, and we observed companies with the following names sending
24 them personal information: “Arial & Babies,” “Androbaby,” “Babies Funny World,”
25 “BabyBus Kids Games,” “For Little Kids,” “GameForKids,” and “KidsUnityApps.” From
26

27 ⁵⁸ Reyes *et al.*, *supra* note 3.

28 ⁵⁹ Serge Egelman, “We get letters.” The AppCensus Blog, May 10, 2018.
<https://web.archive.org/web/20240415123554/https://blog.appcensus.io/2018/05/10/we-get-letters/>.

1 these developer names provided to ironSource, the resulting data was likely coming from
2 children. However, ironSource can deny actual knowledge, so long as no human within
3 the company looks at the data that they are soliciting from developers who use their
4 services.

5 **DARK PATTERNS**

6 50. Another way that consumers are manipulated online is through the use of “dark patterns.”
7 The term “dark patterns” refers to the strategic use of user interface designs to manipulate
8 consumers into acting against their interests:

9 “Dark patterns are user interfaces whose designers knowingly confuse users, make it
10 difficult for users to express their actual preferences, or manipulate users into taking
11 certain actions. They typically exploit cognitive biases and prompt online consumers to
12 purchase goods and services that they do not want or to reveal personal information they
13 would prefer not to disclose.”⁶⁰

14 51. As background, in computing, the “user interface” refers to the mechanisms with which
15 the user directly interacts when using software, hardware, or other systems; it is the point
16 of contact between the human and the computer. The user interface might refer to the
17 physical controls that a user uses to control a device (e.g., a mouse and keyboard) or it
18 may refer to graphical elements found within software that the user must interact with to
19 control that software (e.g., buttons, menus, text, etc.).

20 52. “Human-computer interaction” (HCI) is a discipline within computer science that has
21 existed for over half a century and studies how people interact with computers. “[HCI]
22 draws on the fields of computer science, psychology, cognitive science, and organisational
23 and social sciences in order to understand how people use and experience interactive
24 technology.”⁶¹ This involves studying how people use both software- and hardware-based
25 user interfaces. Practitioners in the field are prevalent throughout industry: online
26 services’ revenue is so highly dependent on their customers’ abilities to correctly navigate

27 ⁶⁰ Jamie Luguri, Lior Jacob Strahilevitz, “Shining a Light on Dark Patterns,” *Journal of*
Legal Analysis, Volume 13, Issue 1, 2021, Pages 43–109, <https://doi.org/10.1093/jla/laaa006>

28 ⁶¹ Cairns P, Cox AL, eds. “Frontmatter.” In: *Research Methods for Human-Computer*
Interaction. Cambridge University Press; 2008:i-vi.

1 their user interfaces that it is standard practice for companies to employ entire teams
2 dedicated to the design of their products' user interfaces.

3 53. Behavioral economists and others who study decision making have long known that *how* a
4 choice is presented to a person has a profound impact on the choice that person ultimately
5 makes.⁶² For example, someone is more likely to purchase a yogurt labeled as being “95%
6 fat free” than if the same yogurt were labeled as “5% fat,” even though both labels convey
7 the same information. This type of intentional manipulation of how choices are presented
8 to consumers to influence their decision making is known as “nudging”:

9 “A nudge, as we will use the term, is any aspect of the choice architecture that alters people's
10 behavior in a predictable way without forbidding any options or significantly changing their
11 economic incentives. To count as a mere nudge, the intervention must be easy and cheap to
12 avoid. Nudges are not mandates. Putting fruit at eye level counts as a nudge. Banning junk
13 food does not.”⁶³

14 54. User interface design is a powerful tool that directly influences how users use the system
15 to make various decisions. That is, user interfaces can and do embody nudges: design
16 suggestions that guide users towards a recommended course of action without
17 constraining their choices. As noted earlier, the term “dark patterns” (sometimes known as
18 “deceptive patterns”) refers to user interface nudges that are specifically designed to
19 manipulate users into acting against their own interests. To be clear, “dark patterns” does
20 not refer to a specific design, but a type of conduct:

21 “Internet users are inundated with attempts to persuade, including digital nudges like
22 defaults, friction, and reinforcement. When these nudges fail to be transparent, optional,
23 and beneficial, they can become ‘dark patterns.’”⁶⁴

24 55. This type of “manipulative design” exists in the physical world, as well. Consider this
25 example of traveling through airports from Brignull’s *Deceptive Patterns*:⁶⁵

26 ⁶² Thaler, Richard; Sunstein, Cass (2008). *Nudge: improving decisions about health, wealth, and happiness*. Yale University Press.

27 ⁶³ *Id.*

28 ⁶⁴ Fagan P. “Clicks and tricks: The dark art of online persuasion.” *Curr Opin Psychol*. 2024 Aug;58:101844. doi: 10.1016/j.copsyc.2024.101844. Epub 2024 Jul 10. PMID: 39029271.

⁶⁵ Harry Brignull (2023). *Deceptive Patterns: Exposing the Tricks Tech Companies Use to Control You*. Testimonium Ltd.

1
2 “To understand how businesses can employ design to manipulate users for profit, let’s start
3 with a physical example: travelling through an airport. When you travel through London
4 Gatwick Airport, you’re advised to ‘arrive at least two hours before your flight to allow
5 plenty of extra time to check-in and pass through security.’⁶⁶ But after you go through
6 security at Gatwick, you’re not allowed to go directly to the departure lounge. You’re forced
7 to do something that has nothing to do with your trip, and it consumes your attention, energy
8 and time. You have no choice in the matter – even if you’re running late.”⁶⁷

9 “The London Gatwick mandatory retail experience.”⁶⁸

10 “In the industry, this is known as a ‘forced path’ store layout.⁶⁹ It’s really just a shop that’s
11 a long, winding corridor, packed into a rectangular footprint in the same way your gut is
12 packed into your belly – travellers are forced in one end and come out the other. The curved
13 path serves a useful function for the business – it forces retail displays into the centre of the
14 traveller’s vision, making it almost impossible for them to avoid looking at the stuff on sale
15 as they navigate their way through the area.”⁷⁰

16 56. Researchers have examined the types of dark patterns found in use by various popular
17 online services and have built taxonomies.⁷¹ For example, “fake scarcity” (Figure 4) is a
18 type of dark pattern in which the consumer is told that only a limited supply of a product
19 remains in order to make them more likely to purchase the product due to “fear of missing
20 out”:

21 “Fake scarcity works by creating an artificial sense of limited availability around a product
22 or service, pushing users to act quickly out of fear of missing out. This is achieved by
23 displaying misleading messages about low stock levels or high demand. By tapping into the
24 scarcity cognitive bias, this deceptive pattern preys on users’ natural tendency to assign
25 more value to items that appear rare or exclusive, pushing them into making hasty
26 purchasing decisions without fully evaluating their options.”⁷²

27 ⁶⁶ <https://www.gatwickairport.com/faqs/flights-and-airlines/>

28 ⁶⁷ Harry Brignull (2023). *Deceptive Patterns: Exposing the Tricks Tech Companies Use to Control You*. Testimonium Ltd.

⁶⁸ *Id.*

⁶⁹ Santos, D. (2018, October 9). Customer Paths and Retail Store Layout — Part 3. Aislelabs. <https://www.aislelabs.com/blog/2018/09/26/customer-paths-and-retail-store-layout-part-3>.

⁷⁰ Harry Brignull (2023). *Deceptive Patterns: Exposing the Tricks Tech Companies Use to Control You*. Testimonium Ltd.

⁷¹ Harry Brignull. “Types of Deceptive Pattern.” *Deceptive Patterns*, <https://www.deceptive.design/types>.

⁷² Harry Brignull. “Fake Scarcity.” *Deceptive Patterns*, <https://www.deceptive.design/types/fake-scarcity>.

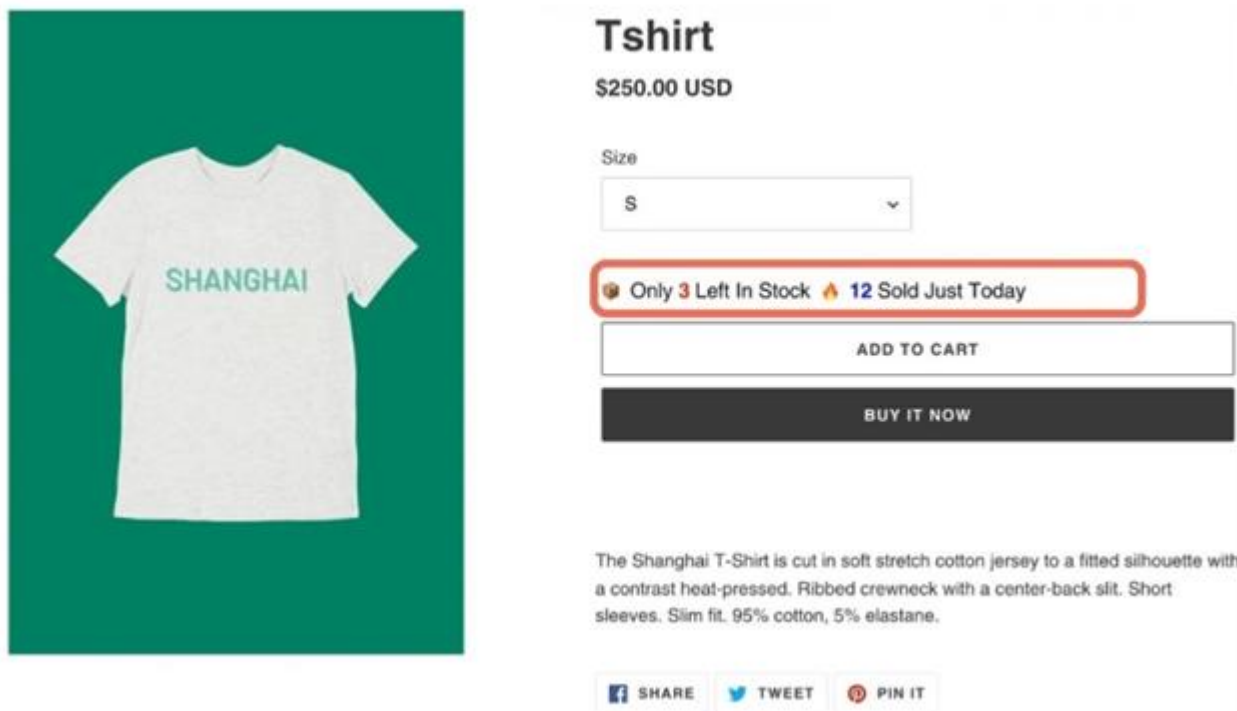


Figure 4: Example of the "fake scarcity" dark pattern, in which low supply and high demand are being represented to the website visitor.
 Source: <https://www.deceptive.design/types/fake-scarcity>

57. In many cases, dark patterns are outright deceptive: researchers found that many of the websites employing the specific dark pattern above (i.e., communicating fake scarcity) simply used random or fixed numbers, and thus were making misrepresentations.⁷³

58. Dark patterns are deployed across many commercial websites and other online services and are often used to encourage consumers to spend additional money or time or to give up privacy.⁷⁴

CALIFORNIA’S AGE-APPROPRIATE DESIGN CODE ACT

⁷³ Arunesh Mathur, Gunes Acar, Michael J. Friedman, Eli Lucherini, Jonathan Mayer, Marshini Chetty, and Arvind Narayanan. 2019. “Dark Patterns at Scale: Findings from a Crawl of 11K Shopping Websites.” Proc. ACM Hum.-Comput. Interact. 3, CSCW, Article 81 (November 2019), 32 pages. <https://doi.org/10.1145/3359183>

⁷⁴ Fagan P. “Clicks and tricks: The dark art of online persuasion.” Curr Opin Psychol. 2024 Aug;58:101844. doi: 10.1016/j.copsyc.2024.101844. Epub 2024 Jul 10. PMID: 39029271; Arielle Pardes. “How Facebook and Other Sites Manipulate Your Privacy Choices.” Wired, August 20, 2020, <https://www.wired.com/story/facebook-social-media-privacy-dark-patterns/>.

1 59. From my understanding of the California Age-Appropriate Design Code Act (AADC), I
2 believe that several of the privacy problems I have identified in my research will be
3 addressed, and that technology to comply with the AADC is already in widespread use
4 (including by NetChoice’s members).

5 60. I understand that the AADC may regulate algorithms. An algorithm is simply a sequence
6 of operations: there is often an input, calculations are performed on that input, and then
7 the results of those calculations are provided as output. Within the context of online
8 services, algorithms are used for everything from recommending content to users to
9 inferring a user’s preferences and traits for purposes such as targeted advertising. There is
10 no such thing as a “neutral” algorithm: algorithms are designed for specific purposes. One
11 algorithm might be designed to show ads that maximize ad revenue, whereas another
12 might be designed to optimize engagement through content recommendations; other
13 algorithms might be used for more mundane tasks, such as sorting items chronologically
14 or alphabetically. For example, in determining the tweets that appear in a user’s feed (of
15 the hundreds of millions sent per day), the online service formerly known as Twitter
16 weighed factors such as the number of likes, retweets, social relations, recency, perceived
17 topic relevance, and use of embedded media, among other factors.⁷⁵

18 61. While some algorithms might make objective decisions (e.g., correctly sorting a list of
19 items by date), others are subjective and therefore less straightforward to audit for
20 correctness (e.g., recommending content and choosing advertisements to display).⁷⁶
21 Algorithms are increasingly being used to make decisions about individuals that can have
22 profound consequences, such as extending credit, housing, insurance, employment, or
23 school admissions; in many cases there is little transparency or recourse surrounding these
24 decisions, as they are made automatically and opaquely, and they may also use incorrect
25

26 ⁷⁵ Josiah Hughes, “How the Twitter Algorithm Works [2023 Guide].” Hootsuite,
27 December 14, 2022. <https://blog.hootsuite.com/twitter-algorithm/>

28 ⁷⁶ Zeynep Tufekci, “Algorithmic Harms beyond Facebook and Google: Emergent
Challenges of Computational Agency,” *Colorado Technology Law Journal* 13, no. 2 (2015): 203-
218.

1 or biased data.⁷⁷ Most adults do not understand if, when, and how these decisions are
2 being made, children less so.

3 62. Algorithms that are optimized for increasing user engagement can also result in harm to
4 consumers. For example, there was public outrage when the public learned that Facebook
5 was using its content recommendation algorithms to intentionally cause emotional distress
6 among its users. (Facebook researchers found that emotionally-charged posts were more
7 likely to lead to user engagement; Facebook thus has an incentive to use its algorithms to
8 prioritize showing users posts that are likely to evoke emotional responses.)⁷⁸ More recent
9 research has shown that misinformation leads to greater levels of engagement for social
10 media platforms: “(i) misinformation sources evoke more outrage than do trustworthy
11 sources; (ii) outrage facilitates the sharing of misinformation at least as strongly as sharing
12 of trustworthy news; and (iii) users are more willing to share outrage-evoking
13 misinformation without reading it first.”⁷⁹

14 63. The AADC regulates the use of so-called “dark patterns.” Dark patterns are design
15 choices that are used to “nudge” the user into making a decision that is advantageous to
16 the business. For example, making it easier to sign up for a service than cancel it is a dark
17 pattern, as is the use of artificial scarcity (e.g., countdown timers to convey a sense of
18 urgency or “limited time” offers).⁸⁰ Research shows that these techniques are prevalent in
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24 ⁷⁷ Danielle Keats Citron and Pasquale, Frank A., “The Scored Society: Due Process for
Automated Predictions” (2014). Washington Law Review, Vol. 89, 2014, p. 1-, U of Maryland
Legal Studies Research Paper No. 2014-8, Available at SSRN: <https://ssrn.com/abstract=2376209>

25 ⁷⁸ Kashmir Hill, “Facebook Manipulated 689,003 Users' Emotions For Science.” Forbes,
26 June 28, 2014. <https://www.forbes.com/sites/kashmirhill/2014/06/28/facebook-manipulated-689003-users-emotions-for-science/>

27 ⁷⁹ Killian L. McLoughlin et al., “Misinformation exploits outrage to spread online.”
Science 386,991-996(2024). doi:10.1126/science.adl2829.

28 ⁸⁰ Sara Morrison, “Dark patterns, the tricks websites use to make you say yes, explained.”
Vox, April 1, 2021. <https://www.vox.com/recode/22351108/dark-patterns-ui-web-design-privacy>

1 child-directed online services,⁸¹ and that children are likely to be more susceptible to
2 manipulations than adults.⁸²

3 64. I understand that the Plaintiff in this case argues that they are unable to estimate the
4 approximate ages of their users. However, the law does not appear to be proscriptive as to
5 how services used by children should perform age estimation. Many such technologies
6 exist, which all have benefits and drawbacks. For example, France’s data protection
7 agency, CNIL, published a guide to choosing appropriate technologies.⁸³ The report
8 recommends that to balance user privacy with age estimation accuracy, services should
9 not perform age estimation themselves, but instead should use independent third parties
10 who can confidentially make guarantees to the child-directed services without revealing
11 additional personal information.

12 65. The report⁸⁴ also links to a prototype “implementation of an age-verification system that
13 allows accessing restricted websites without sharing other personally identifiable data.”⁸⁵
14 The recommended system is based on “zero-knowledge proofs,” a concept in
15 cryptography that has been well-known for almost 40 years now,⁸⁶ which allows an entity
16 to prove the validity of a statement without revealing additional details about that
17 statement. As the CNIL report explains, this technology could easily be used to prove to

18
19 ⁸¹ J. Radesky, A. Hiniker, C. McLaren, E. Akgun, A. Schaller, H. M. Weeks, S. Campbell,
20 & A. N. Gearhardt (2022). “Prevalence and Characteristics of Manipulative Design in Mobile
21 Applications Used by Children.” *JAMA network open*, 5(6), e2217641.
22 <https://doi.org/10.1001/jamanetworkopen.2022.17641>

21 ⁸²Dale Kunkel, Brian L. Wilcox, Joanne Cantor, Edward Palmer, Susan Linn, and Peter
22 Dowrick. “Report of the APA task force on advertising and children.” *Washington, DC:*
23 *American Psychological Association* 30 (2004): 60.

22 ⁸³ CNIL, “Online age verification: balancing privacy and the protection of minors.”
23 September 22, 2022. <https://www.cnil.fr/en/online-age-verification-balancing-privacy-and-protection-minors>

24 ⁸⁴ *Ibid.*

24 ⁸⁵ CNIL, “Demonstration of a privacy-preserving age verification process.” June 23, 2022.
25 <https://linc.cnil.fr/demonstration-privacy-preserving-age-verification-process>

25 ⁸⁶ S. Goldwasser, S. Micali, and C. Rackoff. 1985. “The knowledge complexity of
26 interactive proof-systems.” In *Proceedings of the seventeenth annual ACM symposium on Theory*
27 *of computing (STOC '85)*. Association for Computing Machinery, New York, NY, USA, 291–
28 304. <https://doi.org/10.1145/22145.22178>; U. Fiege, A. Fiat, and A. Shamir. 1987. “Zero
27 knowledge proofs of identity.” In *Proceedings of the nineteenth annual ACM symposium on*
28 *Theory of computing (STOC '87)*. Association for Computing Machinery, New York, NY, USA,
210–217. <https://doi.org/10.1145/28395.28419>.

1 an online service that a user is above or below the age of 18 without revealing additional
2 personal information about that user.

3 66. I understand that Plaintiff implies that it is not possible to reliably determine Internet
4 users' geographic locations in order to determine which regulations apply. This is
5 incorrect. There are many widely-used methods for identifying where in the world an
6 Internet user is physically located. At the most basic level, public and private databases
7 exist that map IP addresses—again, these are transmitted with every Internet connection—
8 to physical locations. This technology is known as “geoIP” and is used by many Internet
9 services to automatically determine where in the world their users come from. For
10 example, MaxMind provides a free database for this purpose that claims 99.8% accuracy
11 in determining a user's country and 80% accuracy for state/region.⁸⁷ Private databases,
12 such as those maintained by several of NetChoice's members, are likely to be more
13 accurate.

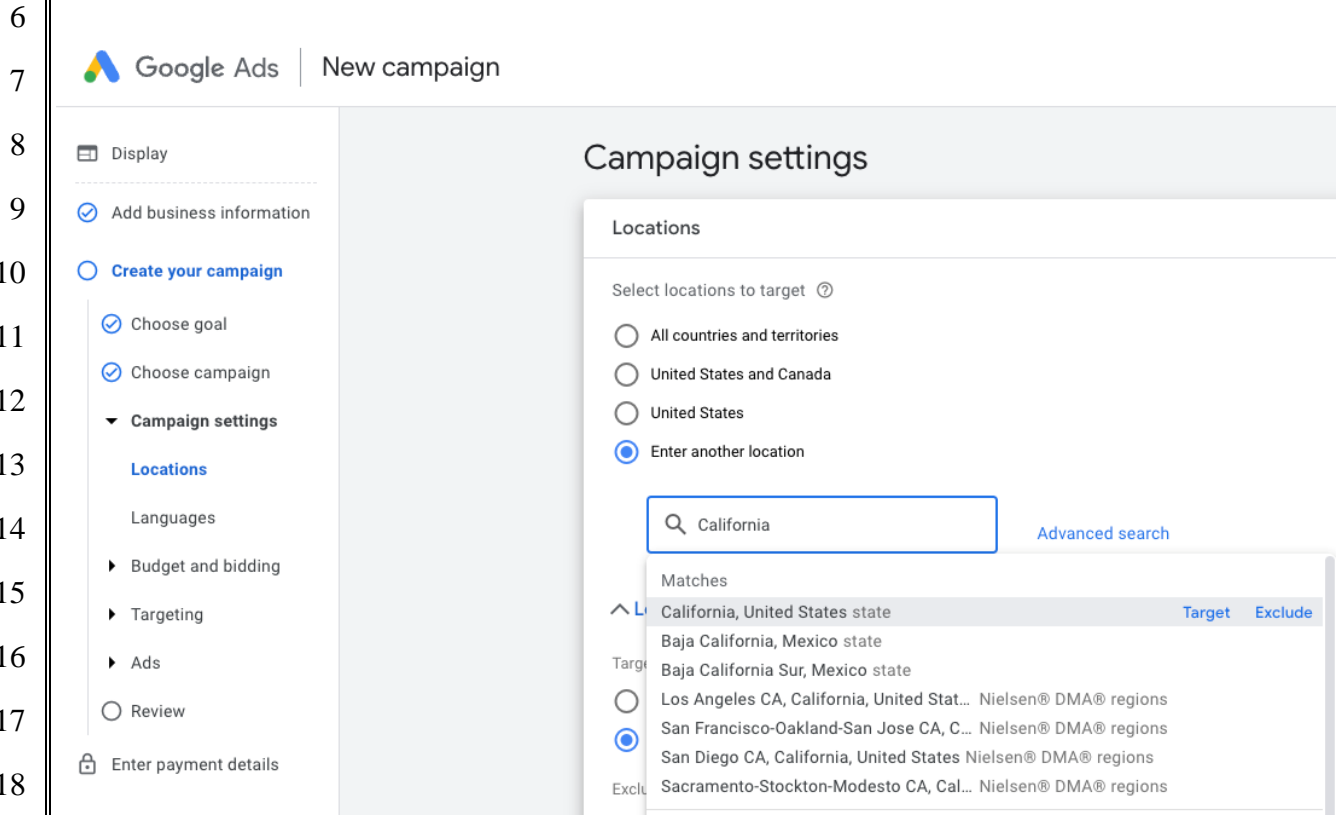
14 67. For example, Meta is already using geoIP data to automatically determine which Internet
15 users should receive protections under CCPA/CPRA. Their documentation explains: “we
16 will determine if a person is in California or not based on certain available signals which
17 may include IP address or advertising ID, when those are available.”⁸⁸ Google similarly
18 automatically detects when users are located in California for the purposes of
19 CCPA/CPRA compliance: “you can select the advertising partners that are eligible to
20 receive bid requests for users Google determines are in California.”⁸⁹

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27 ⁸⁷ <https://support.maxmind.com/hc/en-us/articles/4407630607131-Geolocation-Accuracy>

28 ⁸⁸ <https://www.facebook.com/business/help/1151133471911882>

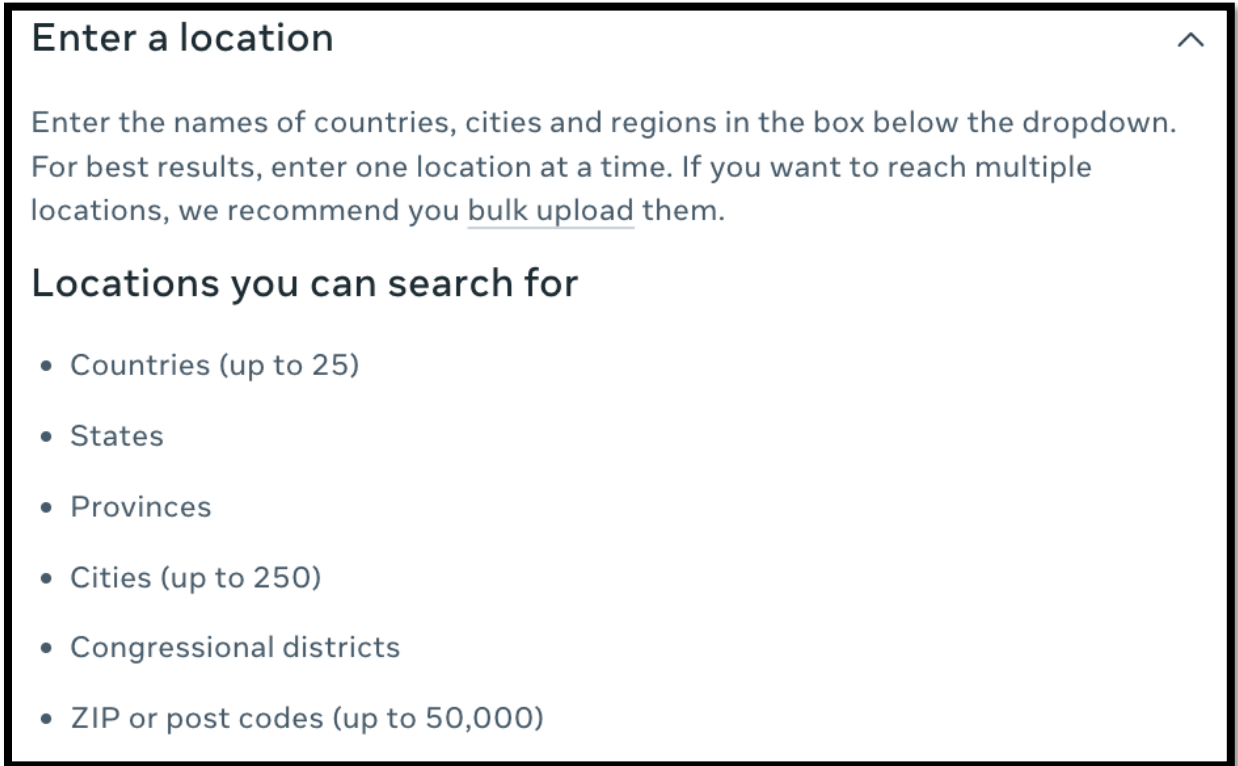
⁸⁹ <https://support.google.com/adsense/answer/9560818?hl=en>

1 68. Both companies named above also allow their customers to specifically target ads to
 2 Internet users located within California. For example, here is a true and correct screenshot
 3 from Google Ads’ targeting configuration interface, <https://ads.google.com/>, accessed on
 4 March 28, 2023, which allows advertisers to show ads to people specifically located
 5 within California:



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1 69. Below is a true and correct screenshot from Meta’s Business Help Center website,
2 <https://www.facebook.com/business/help/365561350785642?id=176276233019487>,
3 accessed on March 28, 2023, describing how their customers can target ads to residents of
4 specific states:



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19 70. Yahoo!, another NetChoice member, also allows their customers to target ads to Internet
20 users in specific states, even using California as an example. Below is a true and correct
21 screenshot from Yahoo!’s Developer Network website,
22 <https://developer.yahoo.com/dsp/docs/lines/targeting-geos.html#target-geographic-areas>,
23 accessed on March 28, 2023:
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Target Named Geographic Locations

Follow the steps below to target named geographic locations, such as countries, states, DMAs, or cities.

1. Select the **Country/State/Region/Sub Region/Metro Area/DMA/City,Zip** radio button.
2. From the Type dropdown, select **Country, State, Region, Sub Region, Metro Area, DMA, City**.

Note

You can start by targeting named locations first and then go back and add zip, postal or prefix codes or vice versa.

3. In the Target text box, type the first few letters of the location you want to target. For example, type all or part of the word "California", then locate it in the dropdown list.

Target

The screenshot shows a search interface with a text input field containing 'Calif'. Below the input is a dropdown menu with the following structure:

- STATE 3
 - California, United States
 - Baja California, Mexico
 - Baja California Sur, Mexico
- SUB REGION 3
 - California, Usulután, El Salvador - Municipality
 - California, Santander Department, Colombia - Municipality
 - California, Parana, Brazil - Municipality
- REGION 3

71. In addition to geoIP lookups using available tools (many of which are already in use by NetChoice's members, in many cases for geolocating users to California for the purpose of determining CCPA/CPRA applicability), other methods exist for geolocating users, such as access to GPS hardware or other device sensors. For example, mobile apps running on the Android platform have access to Google's Geolocation services, which use nearby cellular towers and WiFi networks to determine the user's location, including

1 providing the accuracy radius.⁹⁰ Apple’s iOS platform offers similar functionality, which
2 also make use of nearby cellular networks, WiFi hotspots, and other sensor data.⁹¹

3 72. Similarly, all of the major web browsers support functionality to geolocate their users,⁹²
4 which usually make use of multiple methods, including using WiFi network information,
5 GPS hardware, geoIP databases, and other data sources. Using these methods, the
6 operators of online services have the ability to identify their users with street-level
7 accuracy.

8 73. Thus, the technology to identify California consumers within a reasonable degree of
9 accuracy already exists and is already in use by many of NetChoice’s members.

10 **OPINIONS**

11 74. For the reasons I set out in this declaration, I believe that the AADC takes a reasonable
12 approach to children’s online safety. Based on my research and experience, consumers
13 broadly believe that they are being protected by privacy laws that simply do not exist.
14 Requiring online services to disclose policies in a manner accessible to their users and to
15 enforce those policies would go a long way towards helping consumers make informed
16 decisions about their personal privacy.

17 75. The technologies needed to comply with the AADC’s requirements already exist and are
18 already in widespread use. Behaviors that the AADC prohibits have already been
19 prohibited by major platforms. For example, child-directed Android apps are prohibited
20 from collecting location data or performing behavioral advertising.⁹³

21 76. As demonstrated above, consumers overwhelmingly want the practices this law requires for
22 services that are likely to be accessed by children: limiting privacy-invasive tracking,
23 providing safe defaults, and considering the harm to their users.

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27 ⁹⁰ <https://developers.google.com/maps/documentation/geolocation/overview>

⁹¹ <https://developer.apple.com/documentation/corelocation>

⁹² https://developer.mozilla.org/en-US/docs/Web/API/Geolocation_API

⁹³ <https://support.google.com/googleplay/android-developer/answer/9893335?hl=en>

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77. Finally, I believe that it is reasonable for services likely to be used by children to consider the harm they may have on their users. In fact, I think it's not unreasonable to ask that the offeror of any product or service consider the harm they might be causing to others.

I declare under penalty of perjury that the foregoing is true and correct. Executed on this 6th, day of December, 2024 in Berkeley, California.



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EXHIBIT 1

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education

- 2009 **PhD** in Computation, Organizations, and Society
School of Computer Science Carnegie Mellon University
- 2004 **BS** in Computer Engineering
School of Engineering and Applied Science University of Virginia

experience

- AppCensus, Inc.** San Francisco, CA
 - 2022–Now Chief Scientist / Co-Founder
 - 2019–2022 CTO / Co-Founder
- International Computer Science Institute** Berkeley, California
 - 2016–Now Research Director, Usable Security & Privacy Group
 - 2013–2016 Senior Researcher, Networking and Security Group
- University of California, Berkeley** Berkeley, California
 - 2011–Now Research Scientist, Electrical Engineering and Computer Sciences
- National Institute of Standards and Technology** Gaithersburg, Maryland
 - 2010–2011 Research Scientist, Visualization and Usability Group
- Brown University** Providence, Rhode Island
 - 2009-2010 Postdoctoral Researcher, Computer Science Department
- Microsoft Research** Redmond, Washington
 - 2008 Research Intern, Security and Privacy Group
 - 2008 Research Intern, VIBE Group
- PARC** Palo Alto, California
 - 2006 Research Intern, Computer Science Laboratory

publications*

refereed journal publications

“Protect Me Tomorrow”: Commitment Nudges to Remedy Compromised Passwords
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*Over 13,000 citations and h-index=52: <https://scholar.google.com/citations?hl=en&user=WN9t4n0AAAAJ>

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Feal, Á., Gamba, J., Tapiador, J., Wijesekera, P., Reardon, J., Egelman, S., and Vallina-Rodriguez, N. International Conference on Computers, Privacy and Data Protection (*CPDP '20*), 2020.

A Qualitative Model of Older Adults' Contextual Decision-Making About Information Sharing

Frik, A., Bernd, J., Alomar, N., and Egelman, S. Workshop on the Economics of Information Security (*WEIS '20*), 2020.

Empirical Measurement of Systemic 2FA Usability

Reynolds, J., Samarin, N., Barnes, J., Judd, T., Mason, J., Bailey, M., and Egelman, S. Proceedings of the 29th USENIX Security Symposium (*USENIX Security '20*), 2020.

A Promise Is A Promise: The Effect of Commitment Devices on Computer Security Intentions

Frik, A., Malkin, N., Harbach, M., Peer, E., and Egelman, S. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '19*), 2019.

Privacy and Security Threat Models and Mitigation Strategies of Older Adults

Frik, A., Nurgalieva, L., Bernd, J., Lee, J. S., Schaub, F., and Egelman, S. Proceedings of the 15th Symposium on Usable Privacy and Security (*SOUPS '19*), 2019, Berkeley, CA, USA.

Information Design in An Aged Care Context

Nurgalieva, L., Frik, A., Ceschel, F., Egelman, S., and Marchese, M. Proceedings of the 13th International Conference on Pervasive Computing Technologies for Healthcare (*PervasiveHealth '19*), 2019, New York, NY, USA.

50 Ways to Leak Your Data:

An Exploration of Apps' Circumvention of the Android Permissions System

Reardon, J., Feal, A., Wijesekera, P., On, A. E. B., Vallina-Rodriguez, N., and Egelman, S. Proceedings of the 24th USENIX Security Symposium (*USENIX Security '19*), 2019, Berkeley, CA, USA. **USENIX**

Security Distinguished Paper Award / AEPD Emilio Aced Personal Data Protection Research Award / CNIL-INRIA Privacy Award

- An Experience Sampling Study of User Reactions to Browser Warnings in the Field
Reeder, R. W., Felt, A. P., Consolvo, S., Malkin, N., Thompson, C., and Egelman, S. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '18*), 2018.
- Contextualizing Privacy Decisions for Better Prediction (and Protection)
Wijesekera, P., Reardon, J., Reyes, I., Tsai, L., Chen, J.-W., Good, N., Wagner, D., Beznosov, K., and Egelman, S. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '18*), 2018. **SIGCHI Honorable Mention Award**
- Let's go in for a closer look: Observing passwords in their natural habitat
Pearman, S., Thomas, J., Naeini, P. E., Habib, H., Bauer, L., Christin, N., Cranor, L. F., Egelman, S., and Forget, A. Proc. of the ACM SIGSAC Conference on Computer & Communications Security (*CCS '17*), 2017, New York, NY, USA.
- Turtle Guard: Helping Android Users Apply Contextual Privacy Preferences
Tsai, L., Wijesekera, P., Reardon, J., Reyes, I., Egelman, S., Wagner, D., Good, N., and Chen, J.-W. Proceedings of the 13th Symposium on Usable Privacy and Security (*SOUPS '17*), 2017.
- The Feasibility of Dynamically Granted Permissions:
 Aligning Mobile Privacy with User Preferences
Wijesekera, P., Baokar, A., Tsai, L., Reardon, J., Egelman, S., Wagner, D., and Beznosov, K. Proceedings of the 2017 IEEE Symposium on Security and Privacy (*Oakland '17*), 2017.
- Do or Do Not, There Is No Try: User Engagement May Not Improve Security Outcomes
Forget, A., Pearman, S., Thomas, J., Acquisti, A., Christin, N., Cranor, L. F., Egelman, S., Harbach, M., and Telang, R. Proc. of the 12th Symposium on Usable Privacy and Security (*SOUPS '16*), 2016.
- Behavior Ever Follows Intention? A Validation of the Security Behavior Intentions Scale (SeBIS)
Egelman, S., Harbach, M., and Peer, E. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '16*), 2016. **SIGCHI Honorable Mention Award**
- The Anatomy of Smartphone Unlocking: A Field Study of Android Lock Screens
Harbach, M., Luca, A. D., and Egelman, S. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '16*), 2016. **SIGCHI Honorable Mention Award**
- Keep on Lockin' in the Free World: A Transnational Comparison of Smartphone Locking
Harbach, M., Luca, A. D., Malkin, N., and Egelman, S. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '16*), 2016. **SIGCHI Honorable Mention Award**
- The Teaching Privacy Curriculum
Egelman, S., Bernd, J., Friedland, G., and Garcia, D. Proceedings of the 47th ACM technical symposium on Computer Science Education (*SIGCSE '16*), 2016.
- Android Permissions Remystified: A Field Study on Contextual Integrity
Wijesekera, P., Baokar, A., Hosseini, A., Egelman, S., Wagner, D., and Beznosov, K. 24th USENIX Security Symposium (*USENIX Security 15*), 2015, Washington, D.C.
- Is This Thing On? Communicating Privacy on Ubiquitous Sensing Platforms
Egelman, S., Kannavara, R., and Chow, R. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '15*), 2015, New York, NY, USA.
- Scaling the Security Wall: Developing a Security Behavior Intentions Scale (SeBIS)
Egelman, S., and Peer, E. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '15*), 2015, New York, NY, USA. **SIGCHI Honorable Mention Award**
- Fingerprinting Web Users through Font Metrics
Fifield, D., and Egelman, S. Proceedings of the 19th international conference on Financial Cryptography and Data Security (*FC'15*), 2015.
- Somebody's Watching Me? Assessing the Effectiveness of Webcam Indicator Lights
Portnoff, R., Lee, L., Egelman, S., Mishra, P., Leung, D., and Wagner, D. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (*CHI '15*), 2015, New York, NY, USA.

- Are You Ready to Lock? Understanding User Motivations for Smartphone Locking Behaviors
Egelman, S., Jain, S., Portnoff, R. S., Liao, K., Consolvo, S., and Wagner, D. Proc. of the ACM SIGSAC Conference on Computer & Communications Security (CCS '14), 2014, New York, NY, USA.
- The Effect of Developer-Specified Explanations for Permission Requests on Smartphone User Behavior
Tan, J., Nguyen, K., Theodorides, M., Negron-Arroyo, H., Thompson, C., Egelman, S., and Wagner, D. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '14), 2014, Toronto, Canada.
- The Importance of Being Earnest [in Security Warnings]
Egelman, S., and Schechter, S. Proceedings of the 17th international conference on Financial Cryptography and Data Security (FC'13), 2013, Okinawa, Japan.
- My Profile Is My Password, Verify Me! The Privacy/Convenience Tradeoff of Facebook Connect
Egelman, S. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '13), 2013, Paris, France.
- Does My Password Go up to Eleven? The Impact of Password Meters on Password Selection
Egelman, S., Sotirakopoulos, A., Muslukhov, I., Beznosov, K., and Herley, C. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '13), 2013, Paris, France.
- When It's Better to Ask Forgiveness than Get Permission: Attribution Mechanisms for Smartphone Resources
Thompson, C., Johnson, M., Egelman, S., Wagner, D., and King, J. Proceedings of the Ninth Symposium on Usable Privacy and Security (SOUPS '13), 2013, Newcastle, United Kingdom.
- Android permissions: user attention, comprehension, and behavior
Felt, A. P., Ha, E., Egelman, S., Haney, A., Chin, E., and Wagner, D. Proceedings of the Eighth Symposium on Usable Privacy and Security (SOUPS '12), 2012, Washington, D.C. **SOUPS Best Paper Award (2012) / SOUPS Impact Award (2017)**
- Facebook and privacy: it's complicated
Johnson, M., Egelman, S., and Bellovin, S. M. Proceedings of the Eighth Symposium on Usable Privacy and Security (SOUPS '12), 2012, Washington, D.C.
- It's all about the Benjamins: Incentivizing users to ignore security advice
Christin, N., Egelman, S., Vidas, T., and Grossklags, J. Proceedings of the 15th international conference on Financial Cryptography and Data Security (FC'11), 2011, Gros Islet, St. Lucia.
- Oops, I did it again: mitigating repeated access control errors on facebook
Egelman, S., Oates, A., and Krishnamurthi, S. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '11), 2011, Vancouver, BC, Canada.
- Of passwords and people: measuring the effect of password-composition policies
Komanduri, S., Shay, R., Kelley, P. G., Mazurek, M. L., Bauer, L., Christin, N., Cranor, L. F., and Egelman, S. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '11), 2011, Vancouver, BC, Canada. **SIGCHI Honorable Mention Award**
- Timing is everything?: the effects of timing and placement of online privacy indicators
Egelman, S., Tsai, J., Cranor, L. F., and Acquisti, A. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '09), 2009, Boston, MA, USA.
- It's No Secret: Measuring the Security and Reliability of Authentication via 'Secret' Questions
Schechter, S., Brush, A. J. B., and Egelman, S. Proceedings of the 2009 IEEE Symposium on Security and Privacy (Oakland '09), 2009, Los Alamitos, CA, USA.
- It's not what you know, but who you know: a social approach to last-resort authentication
Schechter, S., Egelman, S., and Reeder, R. W. Proceedings of the SIGCHI Conference on Human Factors in Computing Systems (CHI '09), 2009, Boston, MA, USA.
- Crying wolf: an empirical study of SSL warning effectiveness
Sunshine, J., Egelman, S., Almuhammedi, H., Atri, N., and Cranor, L. F. Proceedings of the 18th USENIX Security Symposium (SSYM'09), 2009, Montreal, Canada.
- Family accounts: a new paradigm for user accounts within the home environment

Egelman, S., Brush, A. J. B., and Inkpen, K. M. Proceedings of the 2008 ACM Conference on Computer Supported Cooperative Work (CSCW '08), 2008, San Diego, CA, USA.

You've Been Warned: An empirical study of the effectiveness of browser phishing warnings
Egelman, S., Cranor, L. F., and Hong, J. CHI '08: Proceeding of The 26th SIGCHI Conference on Human Factors in Computing Systems (CHI '08), 2008, Florence, Italy. **SIGCHI Honorable Mention Award**

Phinding Phish: Evaluating Anti-Phishing Tools

Zhang, Y., Egelman, S., Cranor, L. F., and Hong, J. Proceedings of the 14th Annual Network & Distributed System Security Symposium (NDSS '07), 2007, San Diego, CA.

Power Strips, Prophylactics, and Privacy, Oh My!

Gideon, J., Egelman, S., Cranor, L., and Acquisti, A. Proceedings of the Second Symposium on Usable Privacy and Security (SOUPS '06), 2006, Pittsburgh, PA.

An analysis of P3P-enabled web sites among top-20 search results

Egelman, S., Cranor, L. F., and Chowdhury, A. Proceedings of the 8th international conference on Electronic commerce: The new e-commerce: innovations for conquering current barriers, obstacles and limitations to conducting successful business on the internet (ICEC '06), 2006, Fredericton, New Brunswick, Canada.

refereed workshop publications

Challenges in Inferring Privacy Properties of Smart Devices:

Towards Scalable Multi-Vantage Point Testing Methods

Girish, A., Prakash, V., Egelman, S., Reardon, J., Tapiador, J., Huang, D. Y., Matic, S., and Vallina-Rodriguez, N. Proceedings of the 3rd International CoNEXT Student Workshop (CoNEXT-SW '22), 2022, Rome, Italy.

Identifying and Classifying Third-Party Entities in Natural Language Privacy Policies

Hosseini, M. B., Pradhan, K., Reyes, I., and Egelman, S. Proceedings of the Second Workshop on Privacy in Natural Language Processing (PrivateNLP '20), 2020.

Do You Get What You Pay For? Comparing The Privacy Behaviors of Free vs. Paid Apps

Han, C., Reyes, I., On, A. E. B., Reardon, J., Feal, A., Bamberger, K. A., Egelman, S., and Vallina-Rodriguez, N. The Workshop on Technology and Consumer Protection (ConPro '19), 2019.

Privacy Controls for Always-Listening Devices

Malkin, N., Egelman, S., and Wagner, D. Proceedings of the New Security Paradigms Workshop (NSPW '19), 2019, San Carlos, Costa Rica.

On The Ridiculousness of Notice and Consent: Contradictions in App Privacy Policies

Okoyomon, E., Samarin, N., Wijesekera, P., On, A. E. B., Vallina-Rodriguez, N., Reyes, I., Feal, A., and Egelman, S. The Workshop on Technology and Consumer Protection (ConPro '19), 2019.

Better Late(r) than Never: Increasing Cyber-Security Compliance by Reducing Present Bias

Frik, A., Egelman, S., Harbach, M., Malkin, N., and Peer, E. Workshop on the Economics of Information Security (WEIS '18), 2018.

"What Can't Data Be Used For?" Privacy Expectations about Smart TVs in the U.S.

Malkin, N., Bernd, J., Johnson, M., and Egelman, S. Proceedings of the European Workshop on Usable Security (EuroUSEC '18), 2018.

Personalized Security Messaging: Nudges for Compliance with Browser Warnings

Malkin, N., Mathur, A., Harbach, M., and Egelman, S. Proceedings of the European Workshop on Usable Security (EuroUSEC '17), 2017.

"Is Our Children's Apps Learning?" Automatically Detecting COPPA Violations

Reyes, I., Wijesekera, P., Razaghpanah, A., Reardon, J., Vallina-Rodriguez, N., Egelman, S., and Kreibich, C. The Workshop on Technology and Consumer Protection (ConPro '17), 2017.

Information Disclosure Concerns in The Age of Wearable Computing

Lee, L. N., Lee, J. H., Egelman, S., and Wagner, D. Proceedings of the NDSS Workshop on Usable Security (USEC '16), 2016.

- The Myth of the Average User:
Improving Privacy and Security Systems through Individualization
Egelman, S., and Peer, E. Proceedings of the 2015 Workshop on New Security Paradigms (NSPW '15), 2015, Twente, The Netherlands.
- Teaching Privacy: What Every Student Needs to Know
Friedland, G., Egelman, S., and Garcia, D. Proceedings of the 46th SIGCSE technical symposium on computer science education (Workshop), 2015.
- U-PriSM 2: The Second Usable Privacy and Security for Mobile Devices Workshop
Chiasson, S., Crawford, H., Egelman, S., and Irani, P. Proc. of the 15th International Conference on Human-computer Interaction with Mobile Devices and Services (MobileHCI '13), 2013, Munich, Germany.
- Markets for Zero-day Exploits: Ethics and Implications
Egelman, S., Herley, C., and Oorschot, P. C. van Proceedings of the 2013 Workshop on New Security Paradigms Workshop (NSPW '13), 2013, Banff, Alberta, Canada.
- Choice Architecture and Smartphone Privacy: There's A Price for That
Egelman, S., Felt, A. P., and Wagner, D. The 2012 Workshop on the Economics of Information Security (WEIS '12), 2012, Berlin, Germany.
- How Good Is Good Enough? The sisyphian struggle for optimal privacy settings
Egelman, S., and Johnson, M. Proceedings of the Reconciling Privacy with Social Media Workshop (CSCW '12 Workshop), 2012, Seattle, WA.
- It's Not Stealing if You Need It: A Panel on the Ethics of Performing Research Using Public Data of Illicit Origin
Egelman, S., Bonneau, J., Chiasson, S., Dittrich, D., and Schechter, S. Proceedings of the 16th International Conference on Financial Cryptography and Data Security (FC'12), 2012.
- How to ask for permission
Felt, A. P., Egelman, S., Finifter, M., Akhawe, D., and Wagner, D. Proceedings of the 7th USENIX conference on Hot Topics in Security (HotSec'12), 2012, Bellevue, WA.
- I've got 99 problems, but vibration ain't one: a survey of smartphone users' concerns
Felt, A. P., Egelman, S., and Wagner, D. Proceedings of the second ACM workshop on Security and privacy in smartphones and mobile devices (SPSM '12), 2012, Raleigh, North Carolina, USA.
- Toward Privacy Standards Based on Empirical Studies
Egelman, S., and McCallister, E. The Workshop on Web Tracking and User Privacy (W3C Workshop), 2011, Princeton, NJ.
- Please Continue to Hold: An Empirical Study on User Tolerance of Security Delays
Egelman, S., Molnar, D., Christin, N., Acquisti, A., Herley, C., and Krishnamurthi, S. Workshop on the Economics of Information Security (WEIS '10) (WEIS '10), 2010, Cambridge, MA.
- Tell Me Lies: A Methodology for Scientifically Rigorous Security User Studies
Egelman, S., Tsai, J., and Cranor, L. F. Proceedings of the Workshop on Studying Online Behavior (CHI '10 Workshop), 2010, Atlanta, GA.
- This is Your Data on Drugs: Lessons Computer Security Can Learn from the Drug War
Molnar, D., Egelman, S., and Christin, N. Proceedings of the 2010 Workshop on New Security Paradigms (NSPW '10), 2010, Concord, Massachusetts, USA.
- Security user studies: methodologies and best practices
Egelman, S., King, J., Miller, R. C., Ragouzis, N., and Shehan, E. CHI '07 Extended Abstracts on Human Factors in Computing Systems (CHI EA '07), 2007, San Jose, CA, USA.
- The Effect of Online Privacy Information on Purchasing Behavior: An Experimental Study
Tsai, J., Egelman, S., Cranor, L., and Acquisti, A. Proceedings of the 2007 Workshop on the Economics of Information Security (WEIS '07), 2007, Pittsburgh, PA, USA.
- Studying the Impact of Privacy Information on Online Purchase Decisions
Egelman, S., Tsai, J., Cranor, L. F., and Acquisti, A. Proceedings of the Workshop on Privacy and HCI: Methodologies for Studying Privacy Issues (CHI '06 Workshop), 2006, Montreal, Canada.

book chapters and magazine articles

50 Ways to Leak Your Data: An Exploration of Apps' Circumvention of the Android Permissions System

Reardon, J., Feal, Á., Wijesekera, P., On, A. E. B., Vallina-Rodriguez, N., and Egelman, S. ;*login*: 2019, USENIX Association.

Predicting Privacy and Security Attitudes

Egelman, S., and Peer, E. *Computers and Society*, 2015, ACM.

Crowdsourcing

Egelman, S., Chi, E., and Dow, S. *Ways of Knowing in HCI*, 2013, Springer.

Helping users create better passwords

Ur, B., Kelley, P. G., Komanduri, S., Lee, J., Maass, M., Mazurek, M., Passaro, T., Shay, R., Vidas, T., Bauer, L., Christin, N., Cranor, L. F., Egelman, S., and Lopez, J. ;*login*: 2012, USENIX Association.

Suing Spammers for Fun and Profit

Egelman, S. ;*login*: 2004, USENIX Association.

Installation

Egelman, S. *Peter Norton's Complete Guide to Linux*, 1999, Macmillan Computer Publishing.

User Administration

Egelman, S. *Peter Norton's Complete Guide to Linux*, 1999, Macmillan Computer Publishing.

awards and recognition

2022

CNIL-INRIA Privacy Award

50 Ways to Leak Your Data: An Exploration of Apps' Circumvention of the Android Permissions System, with J. Reardon, A. Feal, P. Wijesekera, A. Elazari Bar On, and N. Vallina-Rodriguez.

Emilio Aced Personal Data Protection Research Award

50 Ways to Leak Your Data: An Exploration of Apps' Circumvention of the Android Permissions System, with J. Reardon, A. Feal, P. Wijesekera, A. Elazari Bar On, and N. Vallina-Rodriguez.

2020

Caspar Bowden Award for Outstanding Research in Privacy Enhancing Technologies

"Won't Somebody Think of the Children?" Examining COPPA Compliance at Scale, with I. Reyes, P. Wijesekera, J. Reardon, A. Elazari, A. Razaghpahanah, and N. Vallina-Rodriguez.

2019

USENIX Security Symposium Distinguished Paper Award

50 Ways to Leak Your Data: An Exploration of Apps' Circumvention of the Android Permissions System, with J. Reardon, A. Feal, P. Wijesekera, A. Elazari Bar On, and N. Vallina-Rodriguez.

2018

SIGCHI Honorable Mention Award (Best Paper Nominee)

Contextualizing Privacy Decisions for Better Prediction (and Protection), with P. Wijesekera, J. Reardon, I. Reyes, L. Tsai, J.-W. Chen, N. Good, D. Wagner, and K. Beznosov.

2017

Symposium on Usable Privacy and Security (SOUPS) Impact Award

Android Permissions: User Attention, Comprehension, and Behavior, with A. P. Felt, E. Ha, A. Haney, E. Chin, and D. Wagner.

Elected ACM Senior Member

Association for Computing Machinery (ACM)

2016

Symposium on Usable Privacy and Security (SOUPS) Distinguished Poster Award

Risk Compensation in Home-User Computer Security Behavior: A Mixed-Methods Exploratory Study, with S. Pearman, A. Kumar, N. Munson, C. Sharma, L. Slyper, L. Bauer, and N. Christin.

- SIGCHI Honorable Mention Award (Best Paper Nominee)**
Behavior Ever Follows Intention? A Validation of the Security Behavior Intentions Scale (SeBIS), with M. Harbach and E. Peer.
- SIGCHI Honorable Mention Award (Best Paper Nominee)**
The Anatomy of Smartphone Unlocking: A Field Study of Android Lock Screens, with M. Harbach and A. De Luca.
- SIGCHI Honorable Mention Award (Best Paper Nominee)**
Keep on Lockin' in the Free World: A Transnational Comparison of Smartphone Locking, with M. Harbach, A. De Luca, and N. Malkin.
- 2015 **SIGCHI Honorable Mention Award (Best Paper Nominee)**
Scaling the Security Wall: Developing a Security Behavior Intentions Scale, with E. Peer.
- 2012 **AIS Best Publication of 2011**
The Effect of Online Privacy Information on Purchasing Behavior: An Experimental Study, with J. Tsai, L. Cranor, and A. Acquisti.

ISR Best Published Paper
The Effect of Online Privacy Information on Purchasing Behavior: An Experimental Study, with J. Tsai, L. Cranor, and A. Acquisti.

SOUPS Best Paper Award
Android Permissions: User Attention, Comprehension, and Behavior, with A. P. Felt, E. Ha, A. Haney, E. Chin, and D. Wagner.
- 2011 **SIGCHI Honorable Mention Award (Best Paper Nominee)**
Of Passwords and People: Measuring the Effect of Password-Composition Policies, with S. Komanduri, R. Shay, P. G. Kelley, M. Mazurek, L. Bauer, N. Christin, and L. F. Cranor.
- 2008 **SIGCHI Honorable Mention Award (Best Paper Nominee)**
You've Been Warned: An Empirical Study on the Effectiveness of Web Browser Phishing Warnings, with L. Cranor and J. Hong.
- 2006 **Tor Graphical User Interface Design Competition**
 Phase 1 Overall Winner, with L. Cranor, J. Hong, P. Kumaraguru, C. Kuo, S. Romanosky, J. Tsai, and K. Vaniea.

Publisher's Clearing House Finalist
 I may already be a winner.

expert testimony and reports

- 2024 Expert witness for the plaintiffs in *Garner v. Amazon.com, Inc., No. 2:21-cv-00750 (W.D. Wash.)*. I provided a report and testimony explaining how in-home “virtual personal assistants” work, as well as explaining the associated privacy concerns based on the relevant research literature.
- 2024 Expert witness for the plaintiffs in *Lopez et al. v. Apple, Inc., No. 4:19-cv-04577-JSW (N.D. Cal.)*. I provided a report explaining how in-home “virtual personal assistants” work, as well as explaining the associated privacy concerns based on the relevant research literature.
- 2024 Expert witness for the plaintiffs in *Martinez et al. v. D2C, LLC d/b/a UNIVISION NOW, No. 1:23-cv-21394-RNS (S.D. Fla.)*. I provided a report and testimony explaining how the Meta Pixel functions and how it was used to transmit consumers' personally-identifiable information in violation of the Video Privacy Protection Act (VPPA).
- 2024 Expert witness for the plaintiffs in *Bloom v. Zuffa LLC, No. 2:22-cv-00412-RFB-BNW (D. Nev.)*. I provided a report and testimony explaining how the Meta Pixel functions and how it was used to transmit consumers' personally-identifiable information in violation of the Video Privacy Protection Act (VPPA).
- 2024 Expert witness for the plaintiffs in *Clark, et. al. v. Yodlee, Inc., No: 3:20-cv-05991-SK (N.D. Cal.)*. I provided a report and testimony explaining basic data protection concepts and consumer privacy expectations.

2024	Independent expert witness appointed by the court in <i>Czarnionka v. The Epoch Times Association, Inc.</i> , No. 1:22-cv-6348 (S.D.N.Y.). I was asked to perform a technical analysis to confirm that the terms of the injunctive relief were being followed.
2023-2024	Expert witness for the plaintiffs in <i>Frasco v. Flo Health, et al.</i> , No. 3:21-cv-00757 (N.D. Cal.). I provided an expert report and testimony based on my forensic analysis of a mobile app's data collection behaviors (i.e., privacy analysis). I was deposed and also provided rebuttal reports of opposing experts.
2023-2024	Expert witness for the California Department of Justice in <i>NetChoice, LLC v. Bonta</i> , No. 5:22-cv-08861. I provided a declaration opposing the motion to dismiss.
2022	Expert witness for the plaintiffs in <i>Hart, et al. v. TWC Product and Technology LLC</i> , No. 4:20-cv-3842-JST. I provided a rebuttal report and testimony about mobile app data collection behaviors.
2022	Expert witness for the District of Columbia Office of the Attorney General in <i>District of Columbia v. Town Sports International LLC</i> . I provided a rebuttal report and testimony on proper surveying methodology.
2021	Expert witness testifying before the U.S. Senate (Committee on Commerce, Science, and Transportation), hearing on "Protecting Kids Online: Internet Privacy and Manipulative Marketing." Testimony available at: https://www.commerce.senate.gov/2021/5/protecting-kids-online-internet-privacy-and-manipulative-marketing
2017-2019	Expert witness for the plaintiffs in <i>Vizio, Inc., Consumer Privacy Litigation</i> , No. 8:16-ml-02693-JLS-KES, assisting with discovery strategy and providing explanations of relevant privacy research on users' willingness to pay for privacy in order to assist in quantifying damages.
2016	Expert witness for the FTC in <i>FTC v. Amazon.com, Inc.</i> , No. C14-1028-JCC, providing testimony on human-computer interaction (HCI) evaluation methods and critiquing opposing expert's report.
2014-2015	Expert witness for the plaintiffs in <i>Doe vs. Twitter, Inc.</i> , No. CGC-10-503630, providing explanations of relevant privacy research on users' willingness to pay for privacy in order to assist in quantifying damages.
2014	Expert witness for the plaintiffs in <i>Levy v. Universal Parking of Florida, LLC</i> No. 13-cv-22122 (S.D. Fla.), providing written testimony on basic human-computer interaction concepts as they relate to smartphone usage.
2013	Expert witness for the plaintiffs in <i>LinkedIn User Privacy Litigation</i> , No. 12-cv-03088-EJD (N.D. Cal.), providing explanations of information security concepts and providing original research on users' privacy expectations in order to demonstrate and quantify damages.
2012	Expert witness for the plaintiffs in <i>Netflix Privacy Litigation</i> , No. 5:11-cv-00379-EJD (N.D. Cal.), providing explanations of relevant privacy research and the economics of information privacy in order to quantify damages.

grants awarded

2023–2026	NSA: Improving Security and Safety of Neural Networks through Robust Training, Noise Augmentation, and Safety Metrics (H98230-23-C-0275) \$750,000 Co-PI (PI: Michael Mahoney, International Computer Science Institute)
2023–2026	NSF: Measuring, Validating and Improving upon App-Based Privacy Nutrition Labels (CNS-2247951/2247952/2247953) \$600,000 Principal Investigator (Collaborative with Adam Aviv, George Washington University; Chris Kanich, University of Illinois at Chicago)

2022–2025	NSF: Developer Implementation of Privacy in Software Systems (CCF-2217771/2217772)	\$750,000
	Principal Investigator (Collaborative with Primal Wijesekera, International Computer Science Institute; Jon Atwell and Julian Nyarko, Stanford University)	
2022–2026	KACST-UCB Center of Excellence for Secure Computing	\$6,460,000
	Senior Personnel (PI: David Wagner, University of California, Berkeley)	
2021–2022	CITRIS: Auditing the Compliance of California Consumer Privacy Regulations at Scale	\$60,000
	Principal Investigator (Collaborative with Zubair Shafiq, University of California, Davis)	
2019	Google: ASPIRE: SDK Traffic Identification at Scale	\$75,000
	Principal Investigator	
2018-2022	NSF: Mobile Dynamic Privacy and Security Analysis at Scale (CNS-1817248)	\$668,475
	Principal Investigator	
2018-2022	NSF: Contextual Integrity: From Theory to Practice (CNS-1801501/1801307/1801316)	\$1,199,462
	Principal Investigator (Collaborative with Helen Nissenbaum, Cornell University; and Norman Sadeh, Carnegie Mellon University)	
2018-2022	NSF: Increasing Users' Cyber-Security Compliance by Reducing Present Bias (CNS-1817249)	\$558,018
	Principal Investigator	
2018-2023	NSA: The Science of Privacy: Implications for Data Usage (H98230-18-D-0006)	\$3,236,424
	Principal Investigator (Co-PI: Michael Tschantz, International Computer Science Institute)	
2018-2019	DHS: Scaling Contextual Privacy to MDM Environments (FA8750-18-2-0096)	\$480,000
	Principal Investigator	
2018-2019	Rose Foundation: AppCensus: Mobile App Privacy Analysis at Scale	\$40,000
	Principal Investigator (Co-PI: Irwin Reyes, International Computer Science Institute)	
2018	Cisco: Access Controls for an IoT World	\$99,304
	Principal Investigator	
2018	CLTC: Privacy Analysis at Scale	\$50,000
	Principal Investigator	
2018	CLTC: Secure Internet of Things for Senior Users	\$60,590
	Co-PI (PI: Alisa Frik, International Computer Science Institute)	
2017	Mozilla: Towards Usable IoT Access Controls in the Home	\$46,000
	Principal Investigator	
2017	Data Transparency Lab (DTL) / AT&T: Transparency via Automated Dynamic Analysis at Scale	\$55,865
	Principal Investigator	
2017	CLTC: Secure & Usable Backup Authentication	\$48,400
	Co-PI (PI: David Wagner, University of California, Berkeley)	
2016 - 2017	NSF: Teaching Security in CSP (CNS-1636590)	\$200,000
	Co-PI (PI: Julia Bernd, ICSI)	
2016 - 2017	DHS: A Platform for Contextual Mobile Privacy (FA8750-16-C-0140)	\$664,378
	Principal Investigator	
2016 - 2018	CLTC: The Security Behavior Observatory	\$195,962
	Principal Investigator	
2016	CLTC: Using Individual Differences to Tailor Security Mitigations	\$100,000
	Principal Investigator	
2015 - 2018	NSF/BSF: Using Individual Differences to Personalize Security Mitigations (CNS-1528070/BSF-2014626)	\$724,732
	Principal Investigator (Collaborative with Eyal Peer, Bar-Ilan University)	

2015 - 2019	NSF: Security and Privacy for Wearable and Continuous Sensing Platforms (CNS-1514211/1514457/1513584)	\$1,200,000
	Principal Investigator (Collaborative with David Wagner, University of California, Berkeley; and Franziska Roesner, University of Washington)	
2014 - 2016	NSF: Teachers' Resources for Online Privacy Education (DGE-1419319)	\$300,000
	Co-PI (PI: Gerald Friedland, ICSI)	
2014 - 2017	NSA: User Security Behavior	\$200,000
	Subcontract (PIs: Lorrie Cranor, Rahul Telang, Alessandro Acquisti, and Nicholas Christin; Carnegie Mellon University)	
2014	Google: Improving Security Warnings by Examining User Intent	\$71,500
	Principal Investigator	
2013 - 2015	NSF: Designing Individualized Privacy and Security Systems (CNS-1343433/1343451)	\$132,620
	Principal Investigator (Collaborative with Eyal Peer, Carnegie Mellon University)	
2013 - 2016	NSF: A Choice Architecture for Mobile Privacy and Security (CNS-1318680)	\$500,000
	Co-PI (PI: David Wagner, University of California, Berkeley)	
2010	Google: Designing Usable Certificate Dialogs in Chrome	\$60,000
	Principal Investigator	

patents awarded

2023	Automatic identification of applications that circumvent permissions and/or obfuscate data flows (US Patent 11,689,551)
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professional activities

program committees

2024	IEEE Security & Privacy; Workshop on Economics and Information Security (WEIS); Contextual Integrity (CI) Symposium
2023	Privacy Enhancing Technologies Symposium (PETS); IEEE Security & Privacy; Workshop on Economics and Information Security (WEIS)
2022	Contextual Integrity (CI) Symposium
2021	Workshop on Economics and Information Security (WEIS)
2020	ACM CCS; Workshop on Economics and Information Security (WEIS); Symposium on Usable Privacy and Security (SOUPS); USENIX Security
2019	Privacy Enhancing Technologies Symposium (PETS); Workshop on Economics and Information Security (WEIS); Symposium on Usable Privacy and Security (SOUPS)
2018	ACM SIGCHI (Human Factors in Computing Systems); Privacy Enhancing Technologies Symposium (PETS); Workshop on Economics and Information Security (WEIS); ACM Conference on Computer and Communications Security (CCS); Symposium on Usable Privacy and Security (SOUPS); IEEE Security & Privacy ("Oakland")
2017	ACM SIGCHI (Human Factors in Computing Systems); USENIX Security; Privacy Enhancing Technologies Symposium (PETS); New Security Paradigms Workshop (NSPW), Co-Chair ; Workshop on Economics and Information Security (WEIS); ACM Conference on Computer and Communications Security (CCS); Symposium on Usable Privacy and Security (SOUPS)

2016	Workshop on the Economics of Information Security (WEIS), Chair ; New Security Paradigms Workshop (NSPW), Co-Chair ; ACM SIGCHI (Human Factors in Computing Systems); USENIX Security; Symposium on Usable Privacy and Security (SOUPS); ACM WWW; Financial Cryptography and Data Security; Privacy Enhancing Technologies Symposium (PETS)
2015	Symposium on Usable Privacy and Security (SOUPS); USENIX Security; ACM SIGCHI (Human Factors in Computing Systems); Privacy Enhancing Technologies Symposium (PETS); Workshop on the Economics of Information Security (WEIS); ACM WWW; Financial Cryptography and Data Security
2014	ACM SIGCHI (Human Factors in Computing Systems); Financial Cryptography and Data Security; ACM WWW; Privacy Enhancing Technologies Symposium (PETS)
2013	ACM SIGCHI (Human Factors in Computing Systems); Symposium on Usable Privacy and Security (SOUPS); New Security Paradigms Workshop (NSPW); Anti-Phishing Working Group eCrime Researchers Summit
2012	Symposium on Usable Privacy and Security (SOUPS); New Security Paradigms Workshop (NSPW)
2011	Symposium On Usable Privacy and Security (SOUPS); New Security Paradigms Workshop (NSPW); Computers, Freedom, and Privacy (CFP) Conference (poster session co-chair); Software and Usable Security Aligned for Good Engineering (SAUSAGE) Workshop, Co-Chair
2010	Symposium On Usable Privacy and Security (SOUPS)
2008	Conference on Information and Knowledge Management (CIKM)
2007	ACM SIGCHI Workshop - Security User Studies: Methodologies and Best Practices; Anti-Phishing Working Group eCrime Researchers Summit (poster session co-chair)
2006	Computers, Freedom, and Privacy (CFP) Conference

standards committees

2007-2008	W3C Web Security Context (WSC) Working Group
2004-2006	W3C Platform for Privacy Preferences (P3P) 1.1 Working Group

leadership roles

2024-Now	Advisory Board Member, Electronic Privacy Information Center (EPIC)
2012-Now	Director, Berkeley Laboratory for Usable and Experimental Security (BLUES)
2021-2023	Member, ICSI Scientific Leadership Council
2006-2008	Legislative Concerns Chair / Board of Directors, National Association of Graduate and Professional Students (NAGPS)
2006-2008	Vice President for External Affairs, Carnegie Mellon Graduate Student Assembly

teaching

Fall 2019	Usable Privacy and Security	University of California, Berkeley
	Designed and taught a course as part of the School of Information's Masters in Cybersecurity program. Duties included course design and development, grading assignment and exams, supervising class projects, and holding office hours.	
Spring 2017, Spring 2018	Human Factors in Computer Security and Privacy	Brown University
	Instructor for a module on "user interfaces for security" as part of the Executive Masters in Cybersecurity program. Duties included course design and development, grading assignments and exams, supervising thesis projects, and holding office hours.	

- Fall 2007 **Information Security & Privacy (46-861)** Carnegie Mellon University
Teaching assistant duties included developing course materials (topics for lectures, assignments, and exams), grading assignments and exams, holding office hours, and mentoring students about semester-long projects.
- Spring 2006 **Computers and Society (15-290)** Carnegie Mellon University
Teaching assistant duties included giving guest lectures, creating assignments and exams, grading assignments and exams, holding office hours, and mentoring students about semester-long projects.
- Fall 2003 **Information Security (CS 451)** University of Virginia
Teaching assistant duties included giving guest lectures, creating assignments and exams, grading assignments and exams, and holding office hours.
- Fall 2003 **Intellectual Property (TCC 200)** University of Virginia
Teaching assistant duties included grading assignments and holding office hours.
- Spring 2003,
Spring 2004 **Advanced Software Development Methods (CS 340)** University of Virginia
Teaching assistant duties included grading and holding office hours.
- Fall 2002 **Engineering Software (CS 201J)** University of Virginia
Teaching assistant duties included grading assignments and holding office hours.