



# Data flows

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thanks: Panagiotis Papadopoulos



Introduction



Web search



Game Theory



Auctions



Data flows



Privacy



Text Ads



Display Ads



Recommender systems



Behavioral targeting



Emerging areas



Final Presentations

Google search: "Car Sales"



\$ 1.72

how does BMW know  
how much Jane Doe is  
worth?

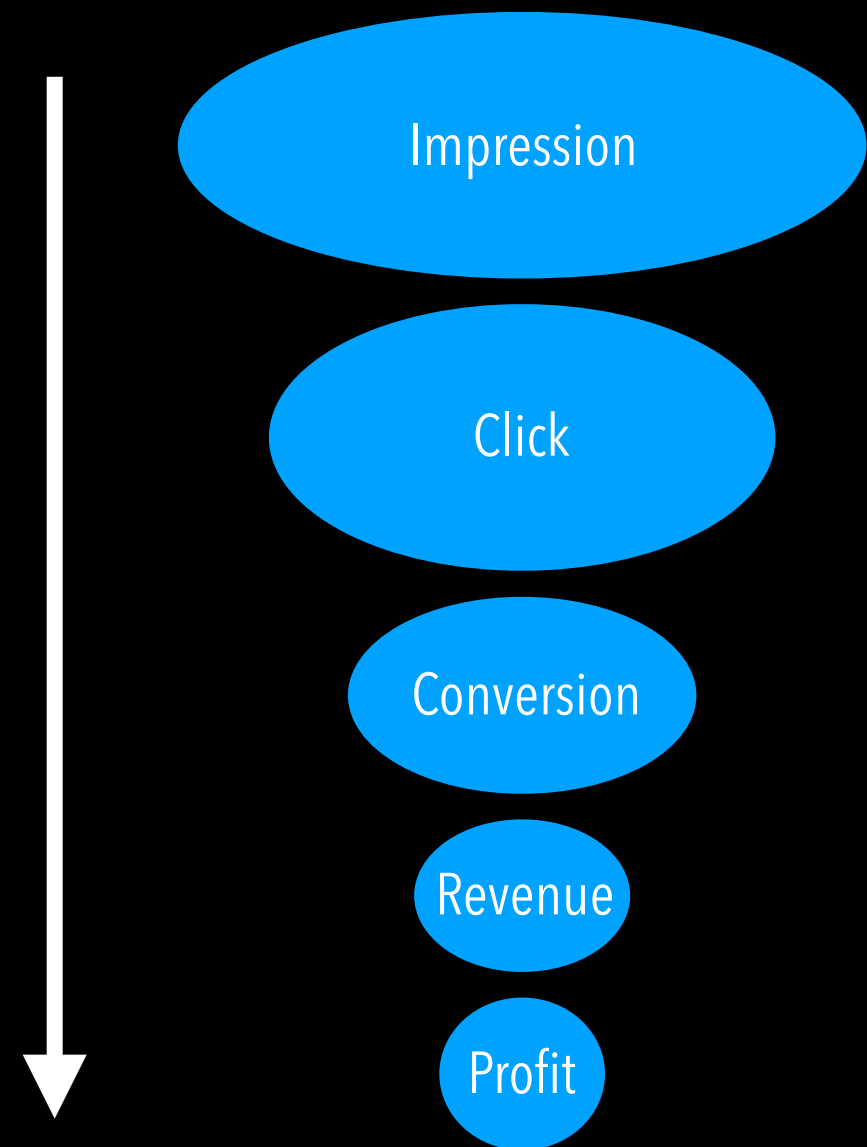


\$ 1.72

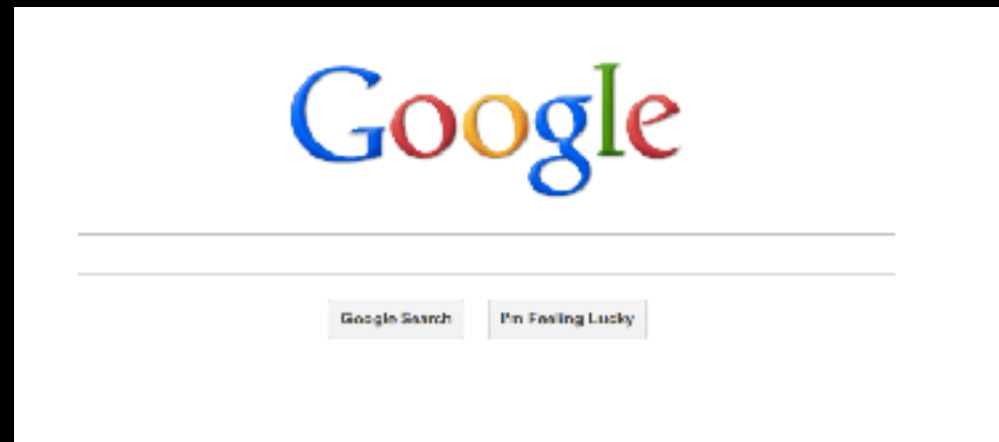
to bid **correctly** on an  
ad auction, advertisers  
need to build profiles

else, they might bid an amount that causes them to lose the auction

one way is to  
model a  
funnel



This is pretty coarse grained



but web-search isn't the  
only place for ads!

display ads within web-sites, mobile apps etc.







how do advertisers  
build profiles about us?

enter the cookie



## 1<sup>st</sup> party:

cookies were invented to  
maintain state of the  
connection on the client;  
often used to maintain  
login credentials at client

# origins

### 1<sup>st</sup> party:

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login credentials at client

# use

### 3<sup>rd</sup> party:

cookies were invented to  
track users across websites

### Single Origin Policy:

origin = **protocol**://**host**:**port**

Network access, Read/write DOM,  
Storage (cookies)

all **three** have to match for pages to  
exchange data

simple in principle, but lots of corner  
cases; browser implementation  
dependent

how does an advertiser  
still know anything about  
you?

Trackers, data brokers (e.g., Axiom) and data management platforms (e.g., Cambridge Analytica, Turn) collect and process user data to form user profiles



# enter the tracker

User profiles may contain information not only from online but **also from the offline world**:

phone number, city/state, email address, SSN, bankruptcy/education information, employment details, information on marriage/divorce, property records, etc.

Profiles are sold in data markets to advertisers for targeted advertising.

to be useful, advertisers need attribution of collected data

# universal ID



gender  
birthdate  
browsing history  
interests  
sexual preferences

"ade87e60-5336-4dd9-9a2a- 763e85516f6d-tuct150ff6a"



# identifying users



data broker id's the user as "userABC"

the advertiser may know that same user as "user123"

how do they figure  
out that "userABC"  
and "user123" are  
the **same** person?

a mechanism to bypass the single-origin policy

allows web companies to share cookies, and match the different IDs they assign for the same user.

# cookie synchronization

157 of top 200 websites (i.e. 78%) have 3rd parties which synchronize cookies with at least one other 3<sup>rd</sup> party

they can reconstruct 62-73% of a user's browsing history\*

Steven Englehardt and Arvind Narayanan. [Online Tracking: A 1-million-site Measurement and Analysis](#). (ACM CCS '16).

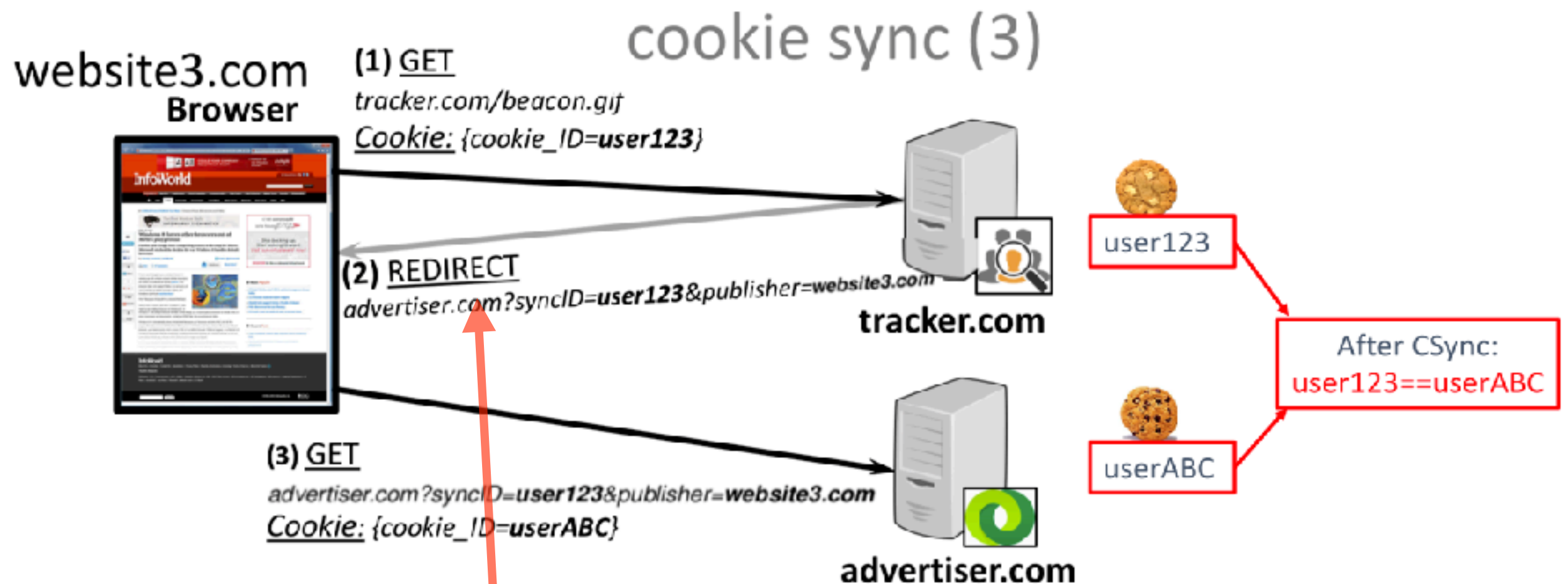
# cookie synchronization



# cookie synchronization



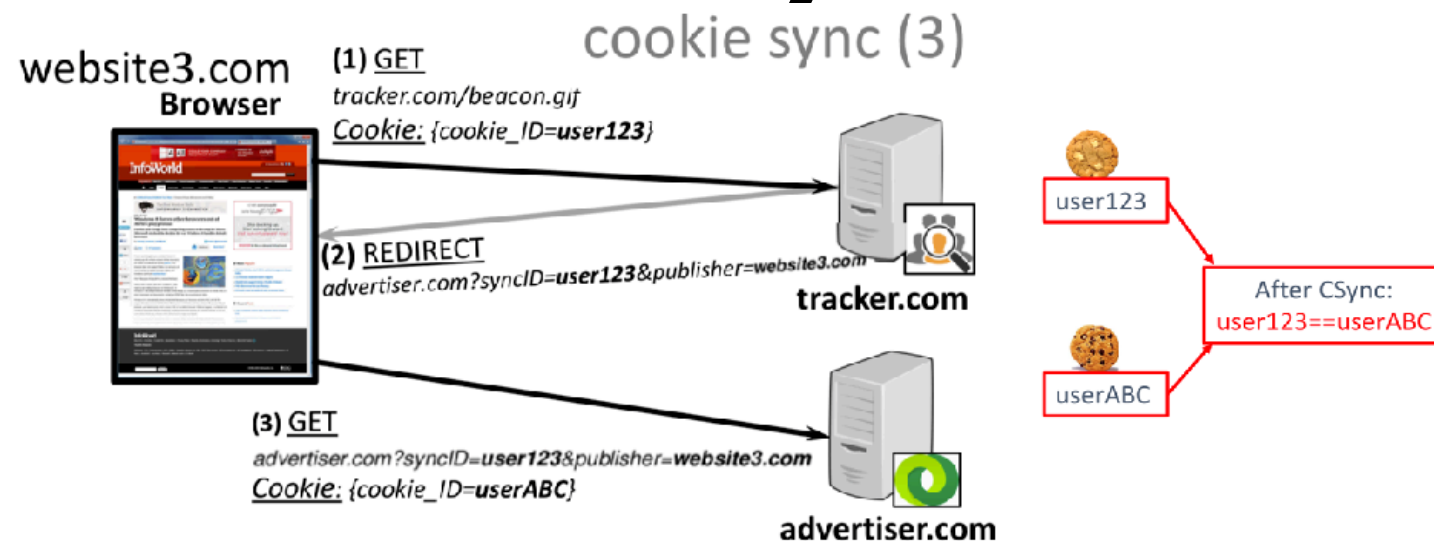
# cookie synchronization



Notice the **redirect**—why is this needed?



# cookie synchronization



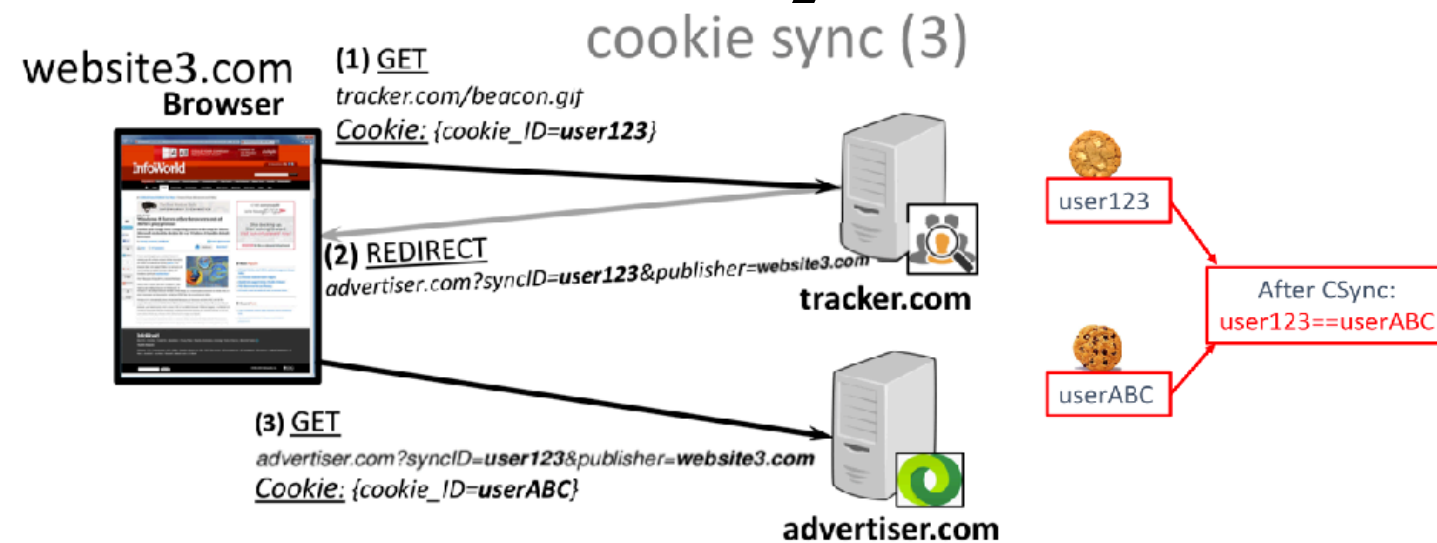
## URLs of Cookie Synchronization HTTP Requests

1. `a.atemda.com/id/csync?s=L2zaWQvMS9lkLzMxOUwOTUw`
2. `bidtheater.com/UserMatch.ashx?bidderid=23&bidderuid=L2zaWQvMS9lkLzMxOUwOTUw&expiration=1426598931`
3. `d.turn.com/r/id/L2zaWQvMS9lkLzMxOUwOTUw/mpid/`

Example real-world 3<sup>rd</sup> party synchronizations

privacy implications

# cookie synchronization



advertiser.com has learnt:

the user has visited website3.com

that the person it knew as user123 is identified as userABC on tracker.com

Server-to-server data merges result in slow loss of anonymity



# why can't we delete cookies?



coupled with **evercookie**, or user fingerprinting, CSync allows re-identification of users even after they delete their cookies

<https://github.com/samyk/evercookie>

# Cookie synchronization in the wild

P. Papadopoulos, N. Kourtellis, and E. Markatos. **Cookie synchronization: Everything you always wanted to know but were afraid to ask**. In The World Wide Web Conference, WWW '19, pages 1432–1442, New York, NY, USA, 2019. ACM.

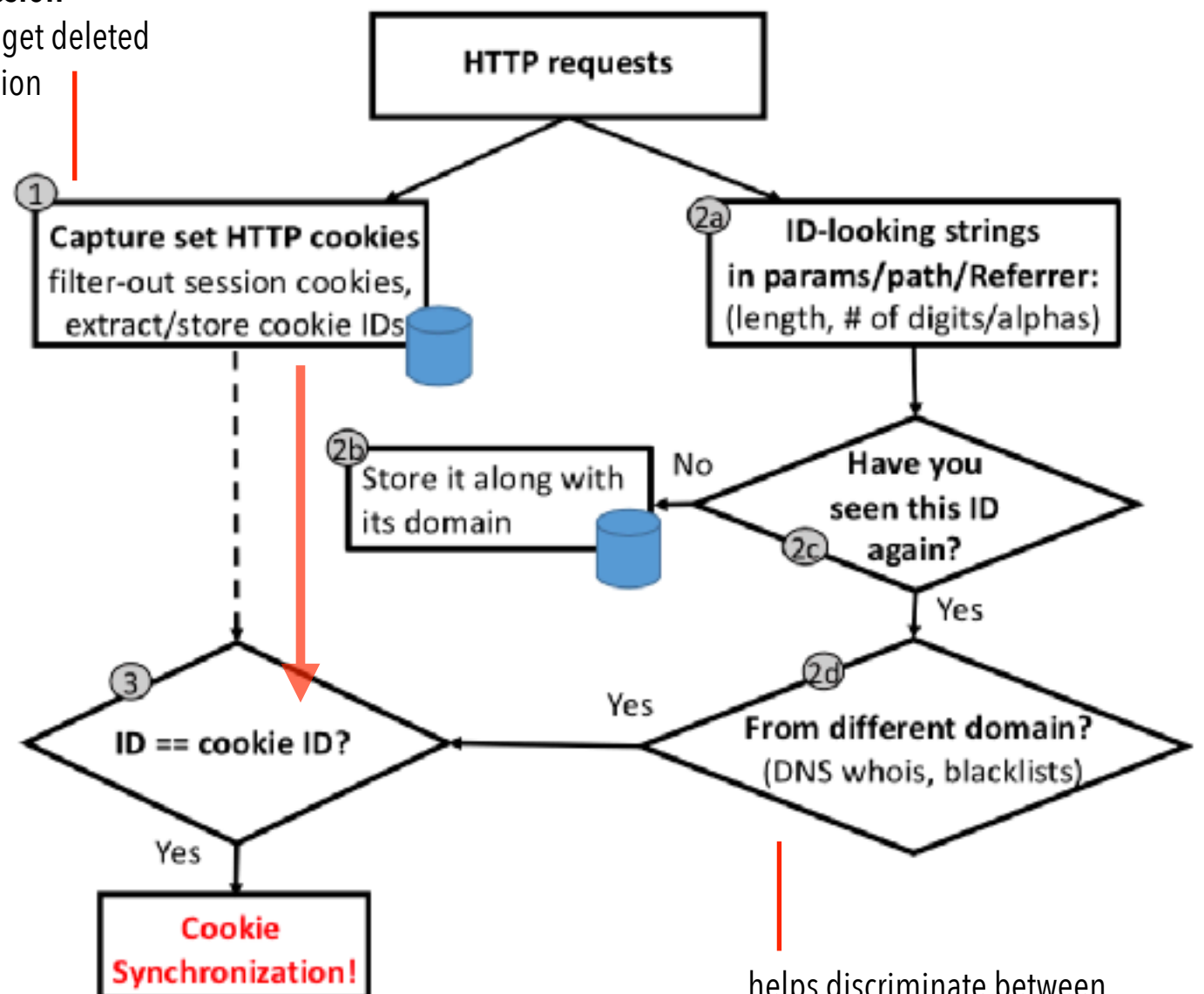
# looking for cookie synchronization

179M HTTP requests  
from mobile devices of  
**850** volunteering users  
across 2016

web traffic redirection  
through a set of  
proxies

use heuristics to detect  
CSync.

filter out session  
cookies that get deleted  
after the session



helps discriminate between  
intentional ID leaking and legitimate  
cases of internal ID-sharing, thus  
avoiding false positives.



# encrypted cookie **synchronization**

the previous method  
relies on IDs being  
synced in plaintext

However, major web companies such as DoubleClick have started **encrypting the cookie ID** in an attempt to protect the actual cookie from being revealed to unwanted parties that may snoop the user's traffic (plugins or even ISPs).

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Under the traditional  
plaintext case of cookie  
ID syncing, the same  
source company can  
sync independently with  
multiple 3rd-parties for  
the same user cookie ID.

But, nothing prevents  
these 3<sup>rd</sup> party  
companies from **syncing**  
**IDs with each other**,  
and determine that they  
have information about  
the same user!

gradual loss of anonymity

with hashing or encryption, in principle, cookie syncing becomes hard and thus may go undetected.



To build this mechanism, they employ machine learning methods (e.g. a **decision tree**), which they train on the ground truth datasets created with the previous, heuristic-based technique.

# detecting encrypted CSync

They analyze various features extracted from the web traffic due to CSync, and train a machine learning classifier to automatically classify a new HTTP connection as being a CSync event or not.

They make the assumption that the various features used to characterize, and eventually detect, CSync with plaintext IDs, are equally used, and have the same distributions and variability as in the CSync with encrypted IDs.

A reasonable assumption, since the companies employing encrypted IDs are not expected to change the rest of their mechanism which delivers these IDs and triggers CSync with their partners; these companies only want to obfuscate the IDs to avoid further, and unwanted, CSync.

# detecting encrypted CSync



## features

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**EntityName**: {domain of recipient company}

**TypeOfEntity** {Content, Social, Advertising, Analytics, Other} **ParamName**: {aid, u, guidm, subuid, tuid, etc.}

**WhereFound**: {parameter in URL, parameter in Referrer, in the URL path}

**StatusCode**: {200, 201, 202, 204, etc.}

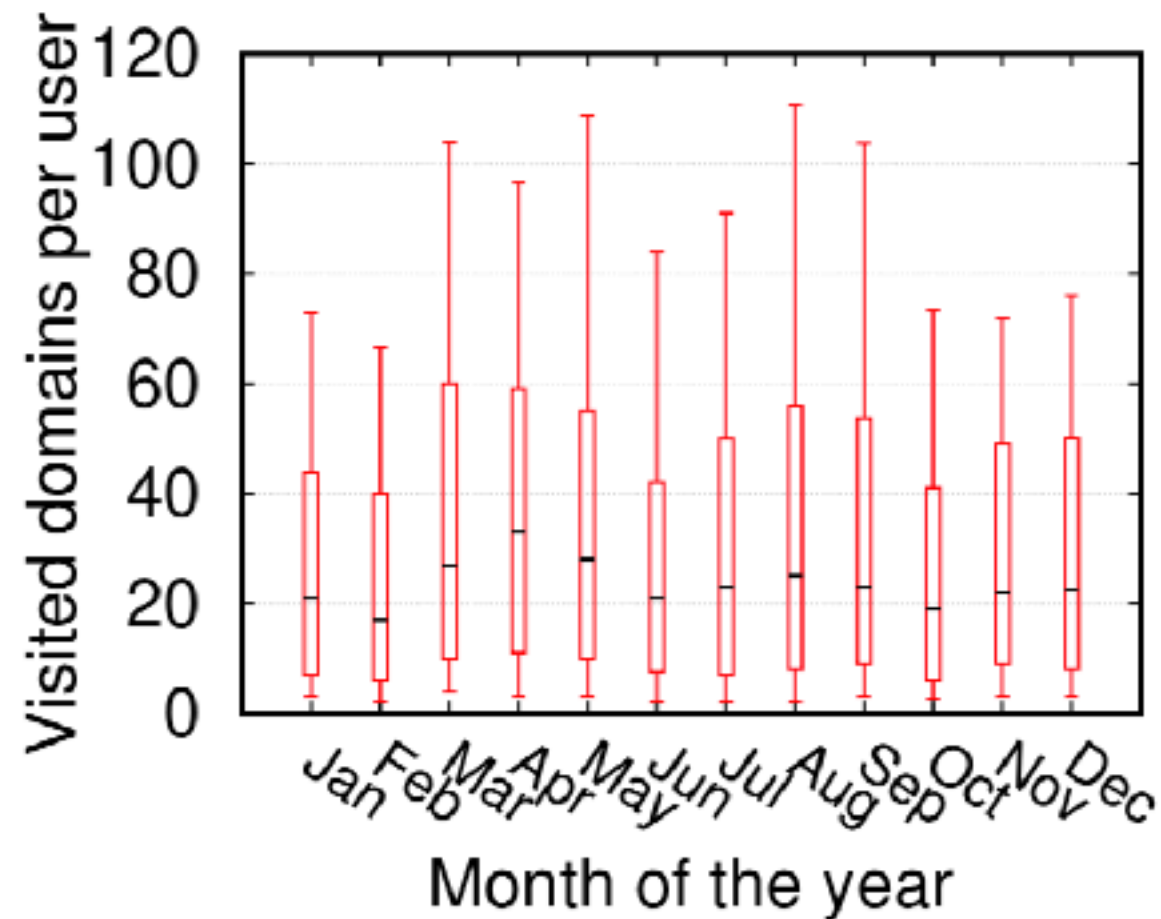
**Browser**: {Firefox, Chrome, Internet Explorer, etc.}

**NoOfParams**: {0, 1, 2, ..., etc.}

# dataset characteristics

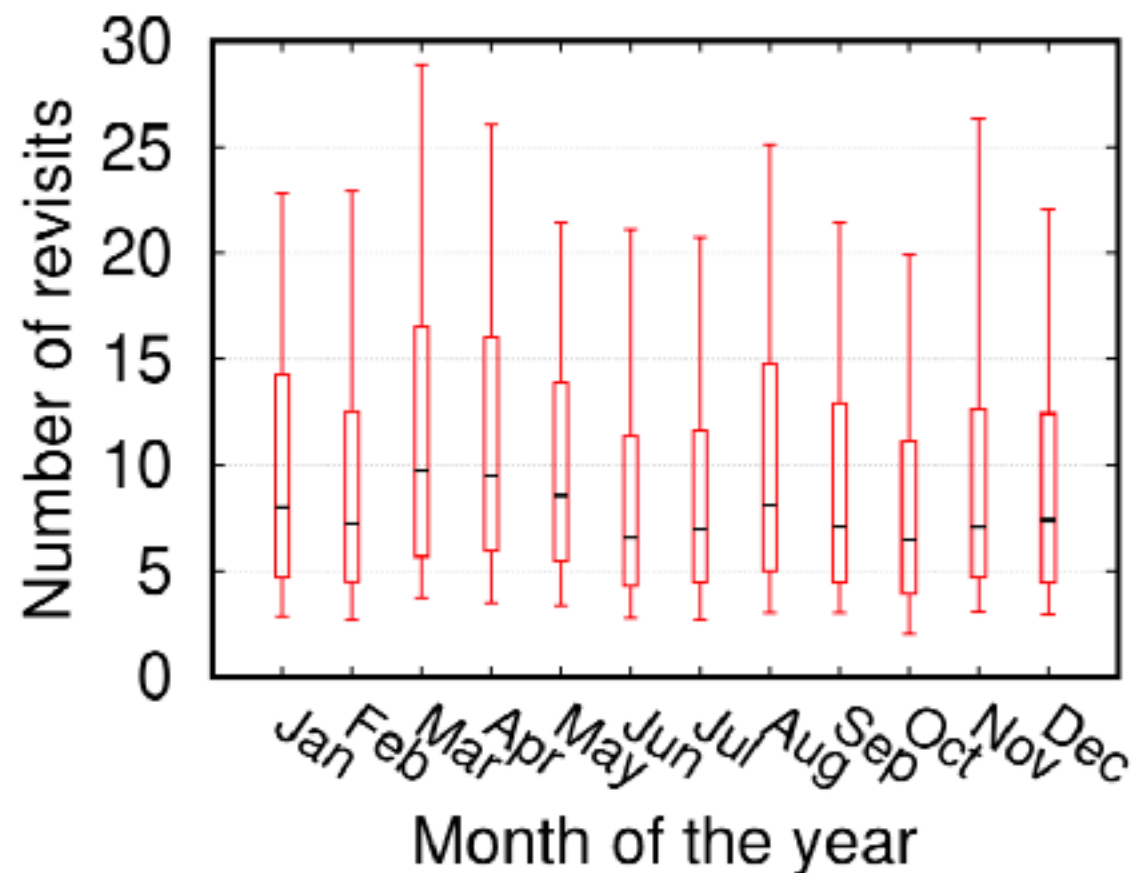
Description	#	Description	#
Total mobile users	850	Unique shared IDs ( $S$ )	68215
Requests captured	179M	Unique userIDs synced ( $C \cap S$ )	22329
Unique Cookies ( $C$ )	8.97M	CSync requests	263635
ID sharing requests	412805		

# dataset characteristics



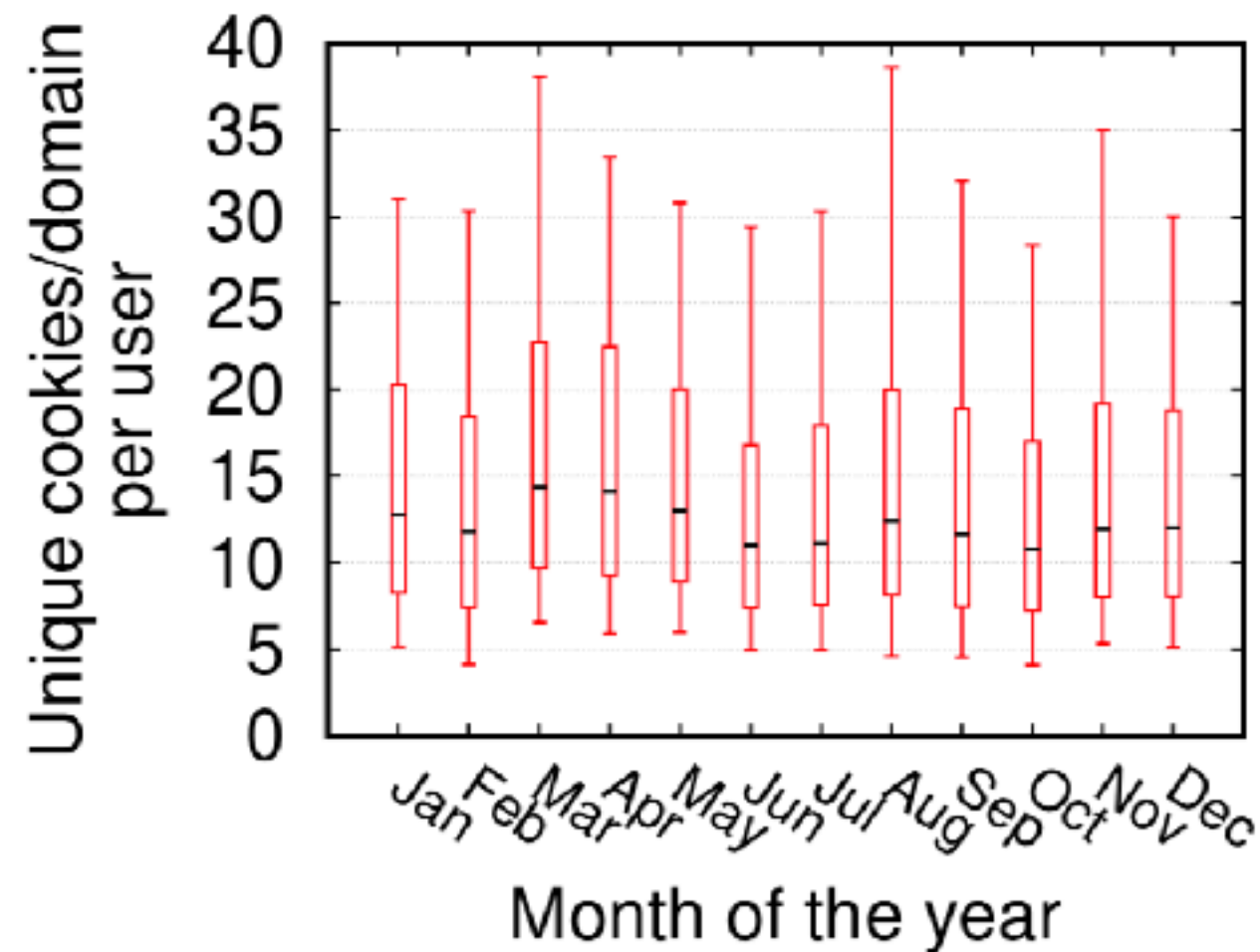
**Figure 4: Distribution of number of unique domains visited per user, per month. The median user in our dataset visits 20 - 30 different domains per month.**

# dataset characteristics



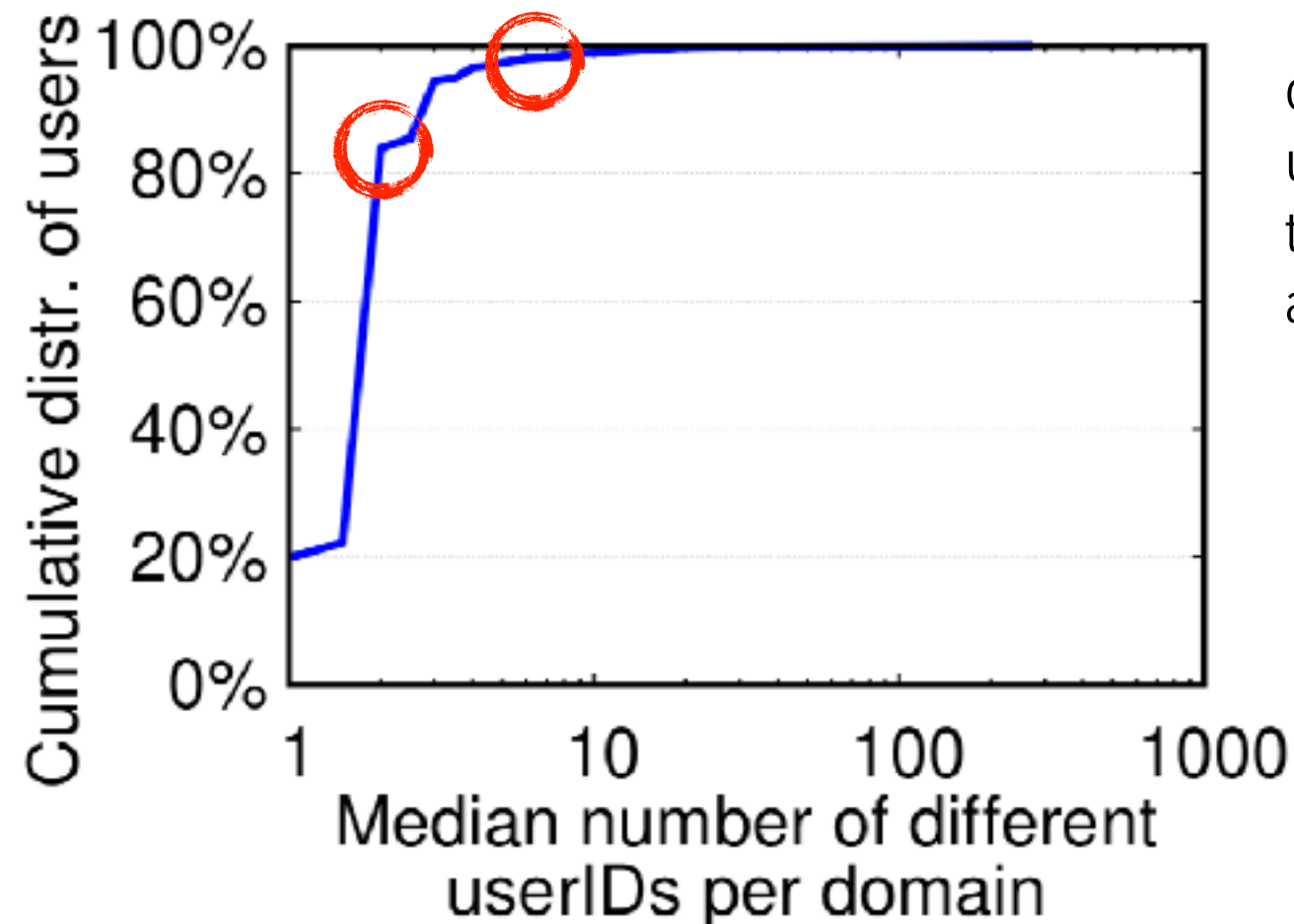
**Figure 5: Distribution of number of times a user revisits the same domain per month. The median user revisits a domain around 7-10 times per month.**

# dataset characteristics



**Figure 6: Number of (first and 3rd-party) cookies per domain per user. We see that the median user receives 12.25 cookies, on average, per visited website.**

# dataset characteristics



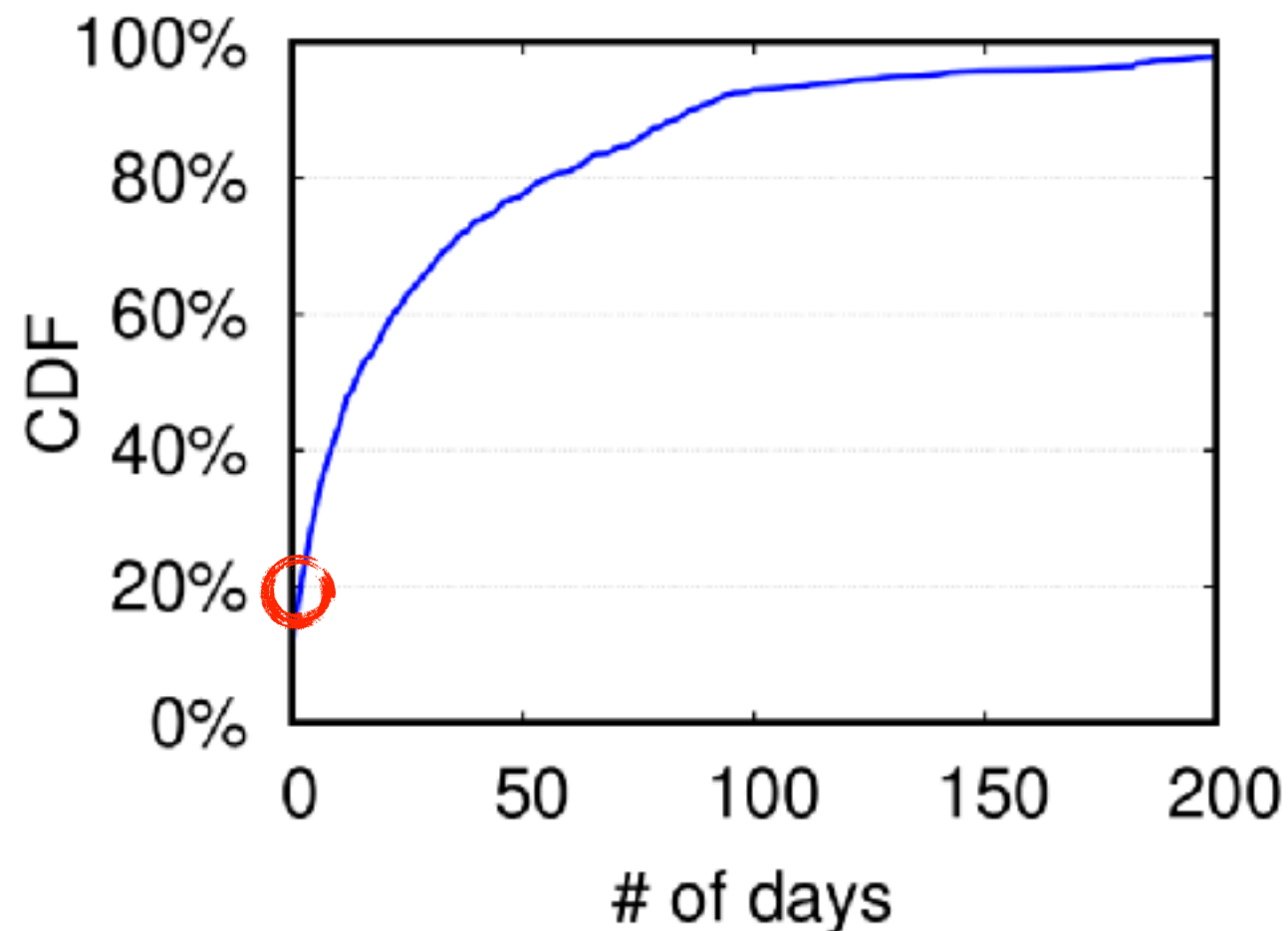
only **1.13%** of the users receive more than 9.5 IDs on average

**Figure 7: Unique userIDs set per domain, across the year. 80% of users are known to a single domain with only ~2 aliases, on average.**

this implies that most users **don't** erase their cookies regularly

# # affected users

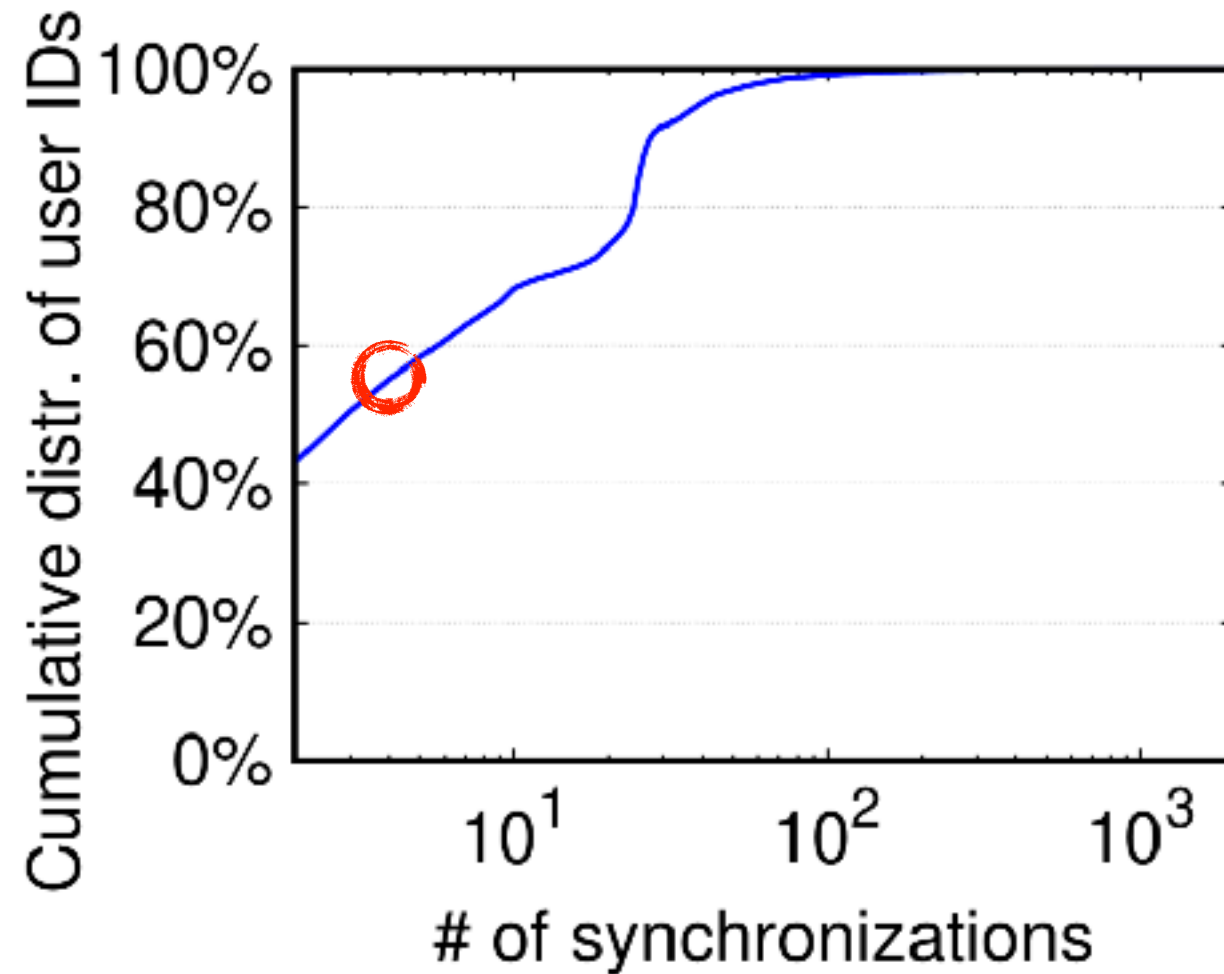
**97%** of users with regular activity the web (>10 HTTP reqs/day) affected



**Figure 8: Distribution of time taken for first CSync to appear per user. 20% of users get their first userID synced in 1 day or less.**



# synchronizations per ID



the average user receives around **1** synchronization per **68** requests.

a **median** user gets up to **6.5** userIDs synced, and **3%** of users has up to **100** userIDs synced.

**Figure 11: Distribution of synchronizations per userID. The median userID gets synced with 3.5 different domains.**

# who initiates?

**Table 4: Breakdown of the CSync triggering factors.**

	<b>Initiator</b>	<b>Portion</b>
(i)	Publisher syncs its userID	2.692%
(ii)	Embedded 3rd-party triggers syncing of its own set userID	49.668%
(iii)	3rd-party uses sync request to share its own set userID	45.697%
(iv)	3rd-party uses sync request to share with other domains the publisher's set userID	0.2658%



# Cookie ID re-use

Cases of domains setting cookies using userIDs previously used by **other** domains.

Example:

baidu.com sets cookie **baiduid = {idA}**

Later, different domains set **their own**  
cookies by using **baiduid = {idA}**

# cookies to summarize



## ID Summary stored in cookie by adap.tv

```
"key=valueclickinc:value=708b532c-5128-4b00-a4f2-  
2b1fac03de81:expiresat=wed apr 01 15:03:42 pdt  
2015,key=mediamathinc:value=60c05435-9357-4b00-  
8135-273a46820ef2:expiresat=thu mar 19 01:09:47 pst  
2015,key=turn:value=2684830505759170345:expiresat=fri  
06 16:43:34 pst 2015,key=rocketfuelinc:value=639511  
149771413484:expiresat=sun mar 29 15:43:36 pst 2015"
```



It includes (previously synced) userIDs and expiration dates as set by 4 different domains.

how can this be possible, when the connection is HTTPS?

# User IDs spill out of TLS!

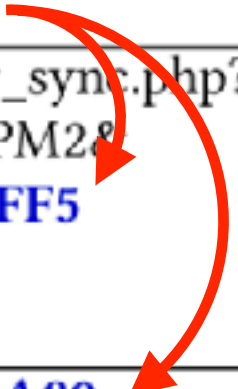
It is caused by CSync events that sync a userID from a TLS cookie with non-TLS 3<sup>rd</sup> parties.

mixing HTTP (un-encrypted) and HTTPS (**encrypted**) is a bad idea

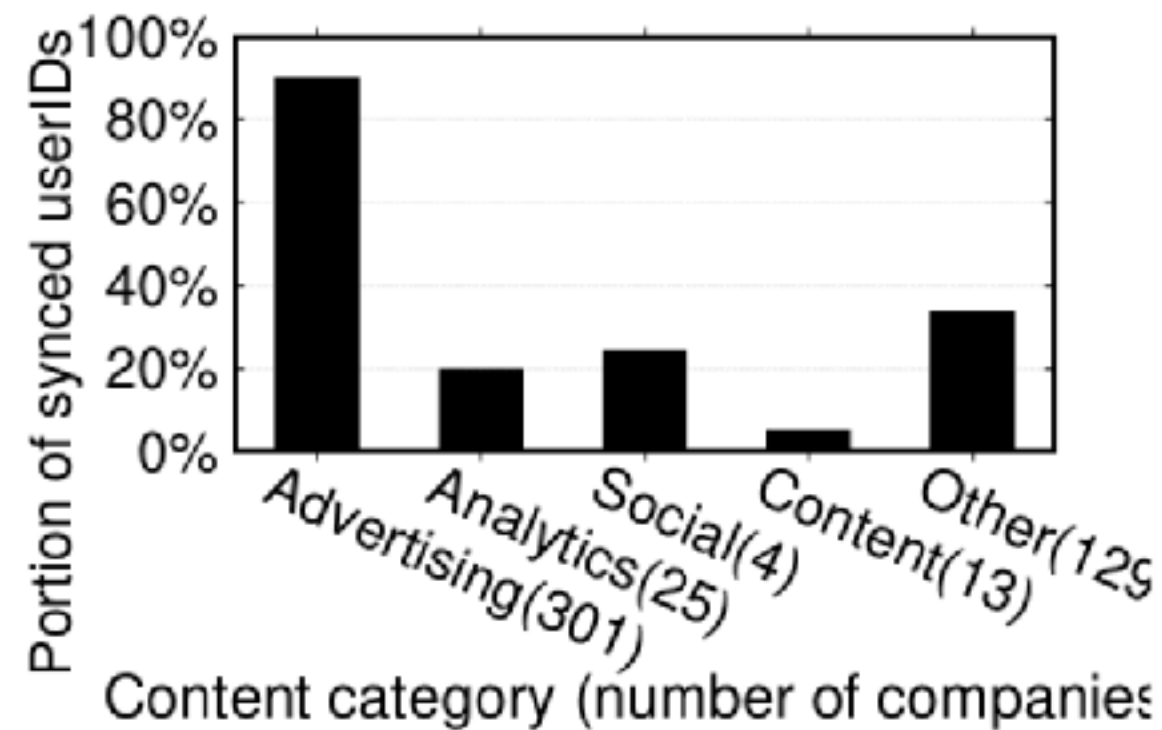
A snooping ISP can eavesdrop

**Table 6: Example of ID-spill from SSL in our dataset.**

Role	Domain
Visited website:	<u>https://</u> financialexpress.com
Cookie setter:	<u>https://</u> tapad.com
SetCookie:	<b>D0821FA0-8A80-4D9E-BC85-C40EAC4E4FF5</b>
Cookie syncer:	<u>http://</u> delivery.swid.switchadhub.com/adserver/user_sync.php? SWID=cf43265166a9ccf5f6fd0472f23776fa&sKey=PM28 sVal= <b>D0821FA0-8A80-4D9E-BC85-C40EAC4E4FF5</b> <u>referrer:</u> <b>financialexpress.com</b> <u>Get-cookie:</u> {cf43265166a9ccf5f6fd0472f23776fa}
Cookie syncer:	<u>http://</u> tags.bluekai.com/site/3096?id= <b>D0821FA0-8A80-4D9E-BC85-C40EAC4E4FF5</b> <u>referrer:</u> <b>financialexpress.com</b> <u>Get-cookie:</u> {c57b29d1-f8e2-11e7-ac1b-0242ac110005}



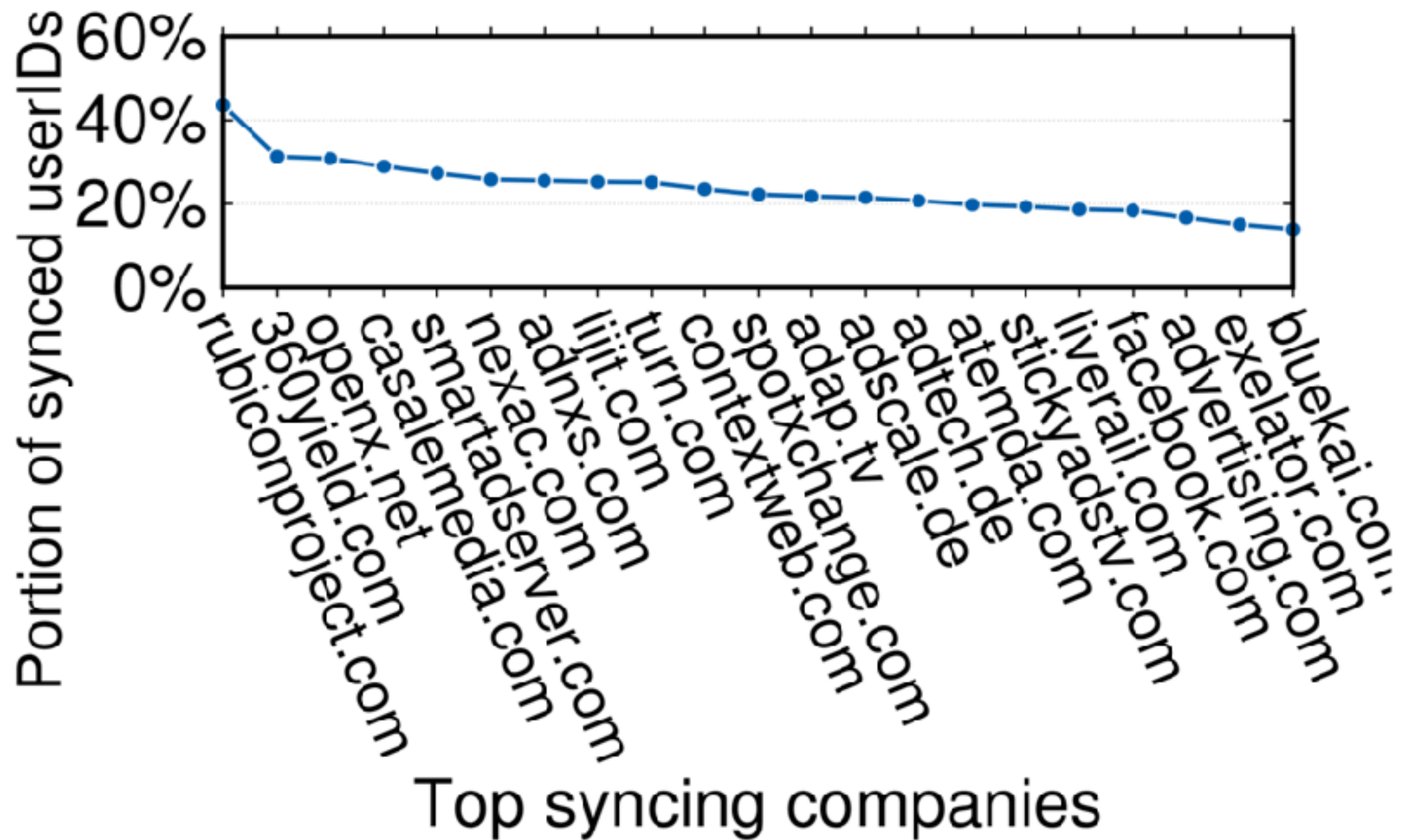
# Top syncing categories



**Figure 14: Portion of synced userIDs learned per content category. As expected, ad-related companies learned the vast majority (90%) of the total synced userIDs in our dataset.**



# Top syncing companies



# Leak of sensitive information

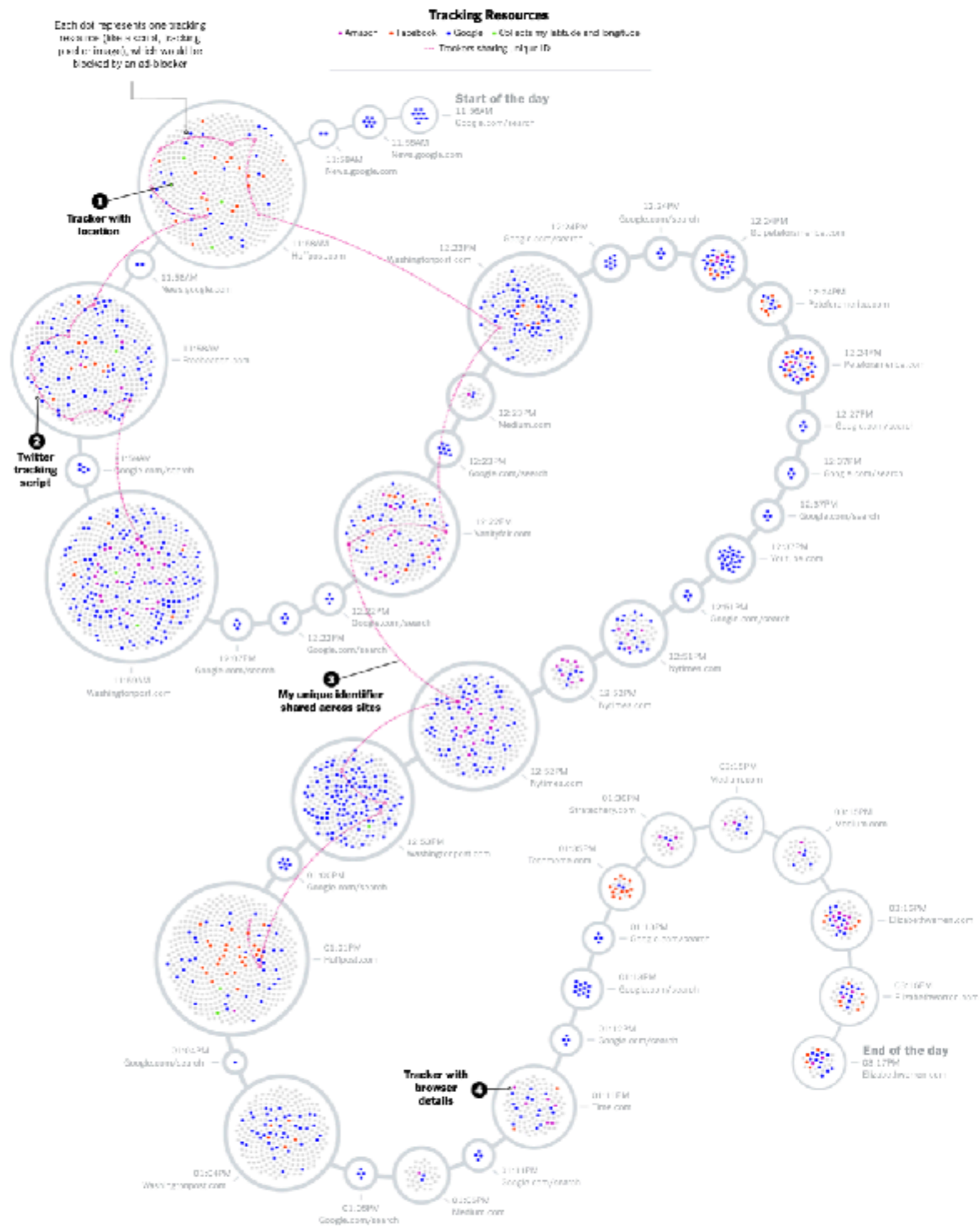
- 13 syncs leaking the user's city level location
- 2 syncs leaking the user's registered phone number
- 10 syncs leaking the user's gender
- 9 syncs leaking the exact user's age
- 3 syncs leaking the user's full birth date
- 2 syncs leaking the user's first and last name
- 16 syncs leaking the user's email address
- 4 syncs leaking user login credentials: username/password

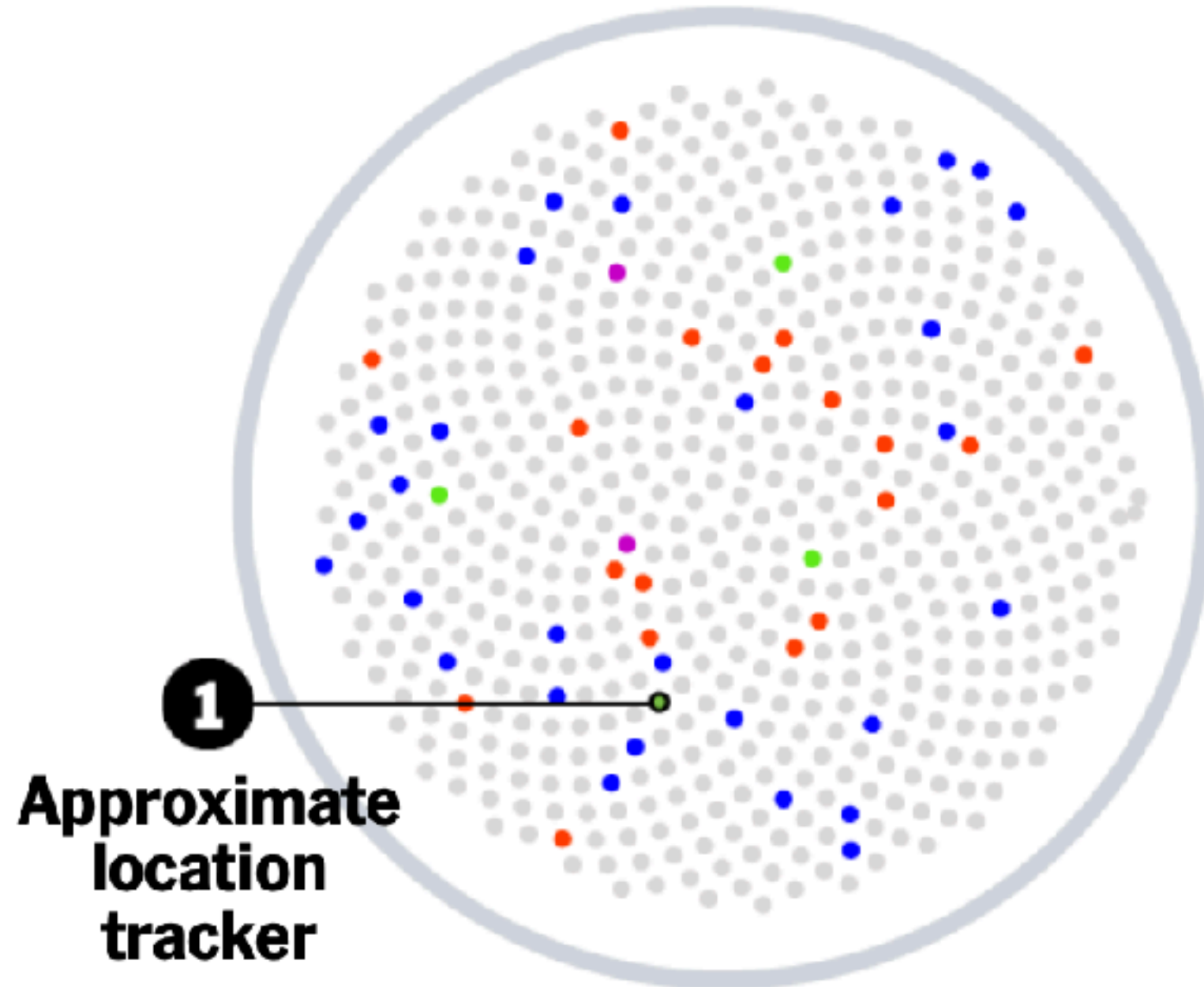


# An example

A New York Times Columnist uses OpenWPM

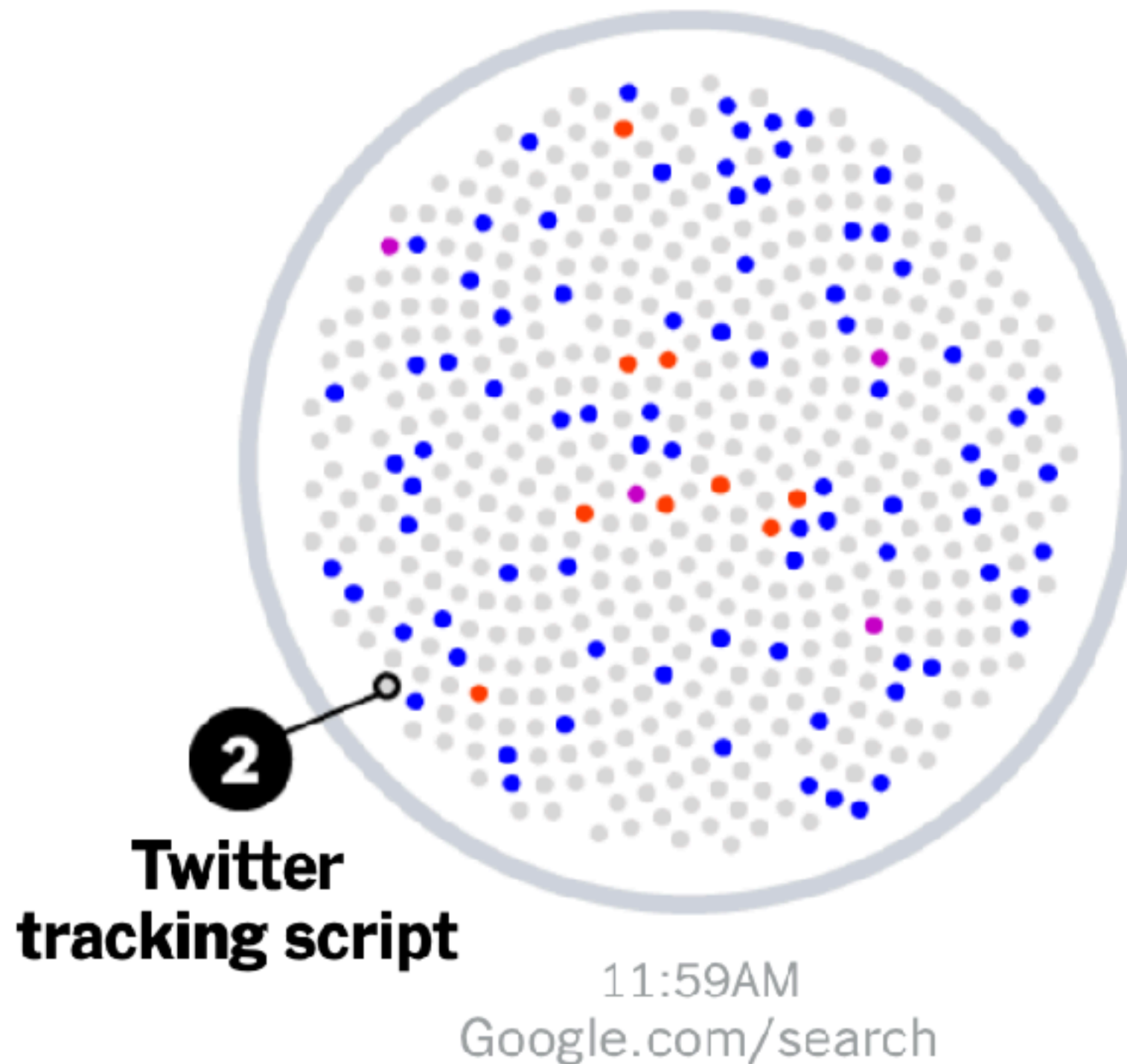
<https://www.nytimes.com/interactive/2019/08/23/opinion/data-internet-privacy-tracking.html>





11:58AM  
Huffpost.com

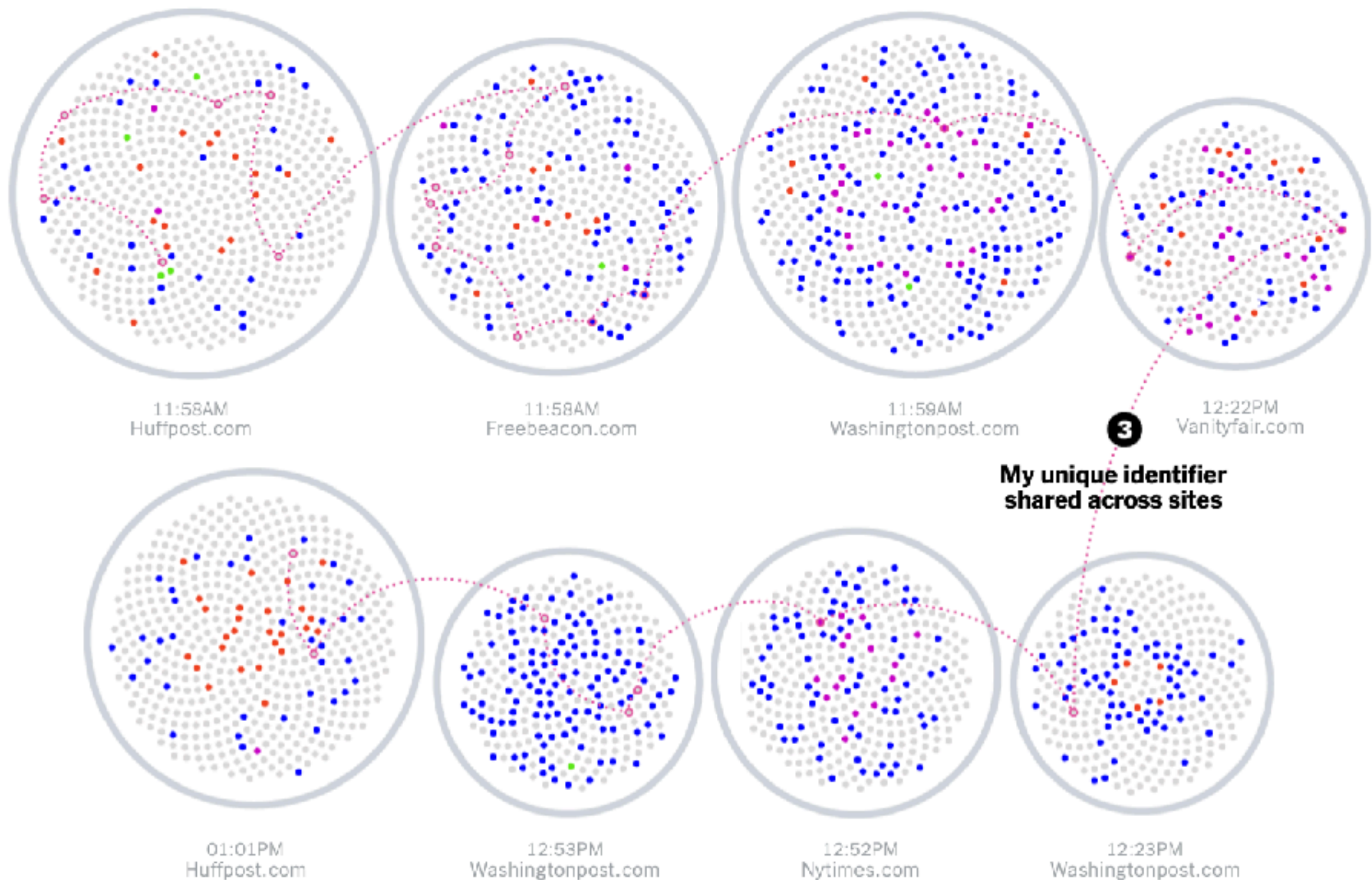
# Where I live



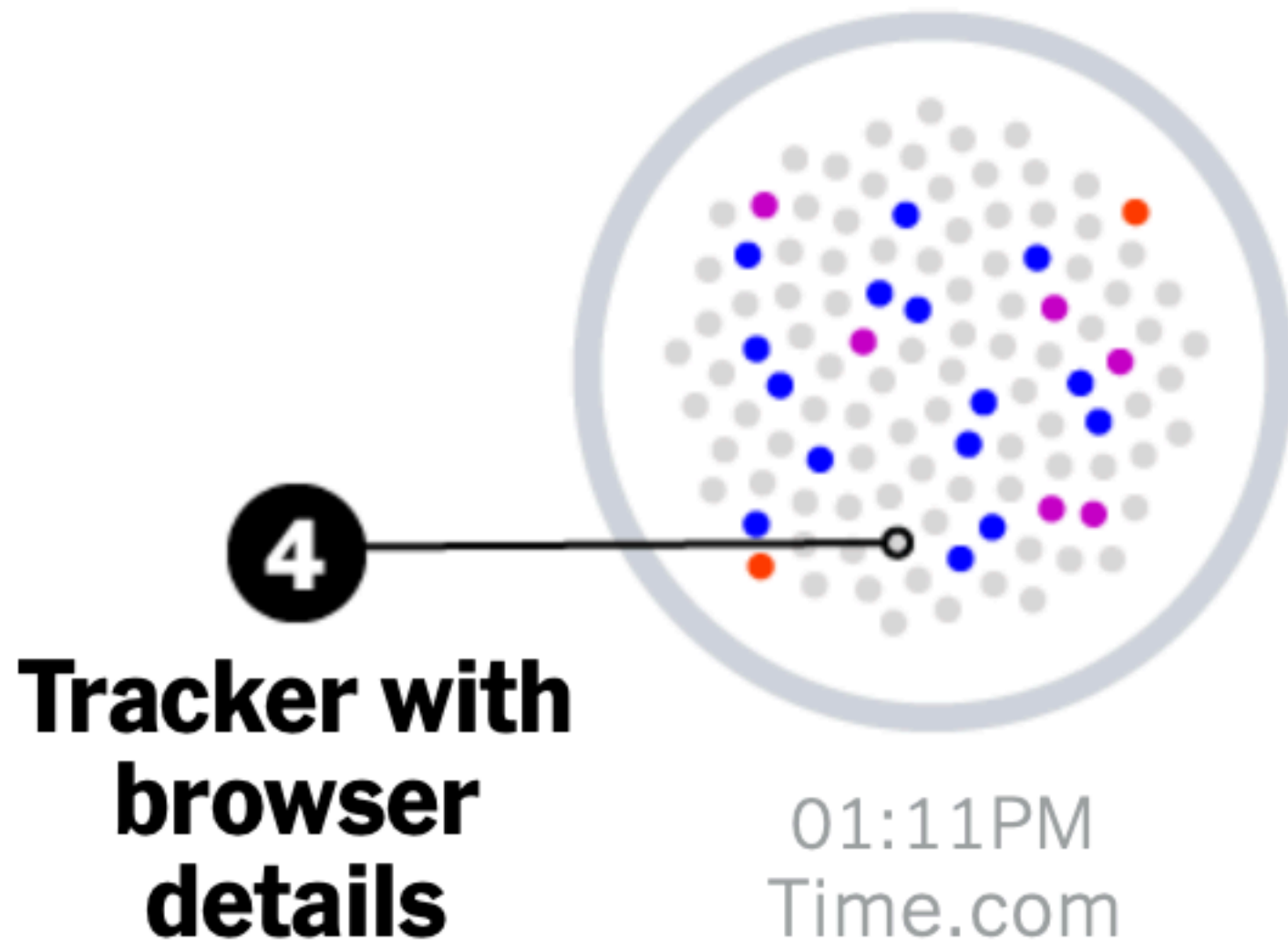
“Tracking scripts like this one for Twitter allow websites to add useful features like share buttons. But the scripts often double as trackers meant to record site visits and build profiles about users. In this case, Twitter can use the information about this page to suggest new followers or sell more targeted advertising on its platform.”

# Widgets or trackers?





**My unique identifier: 5535203407606041218**



“Even when companies don’t have an ID to track me, they can use signals from my computer to guess who I am across sites. That’s partly why trackers like this one received more information about my computer than you could imagine being useful, like my precise screen size. Other trackers received my screen resolution, browser information, operating system details, and more.”

# Fingerprinting

# Counter-measures?

Ad blockers!

97% of users are exposed to CSync at least once. The median user is synced at least once within the first week of browsing.

The average user receives around 1 synchronization per 68 HTTP requests, and gets up to 6.5 of their userIDs synced. The number of domains that learn about the median user after CSyncs grows by a factor of 6.75.

The median userID gets leaked to 3.5 domains, on average.

# Summary:

Cookie synchronization is the basis of much of the data collected by Ad companies

Ad-related domains participate in more than 75% of all CSync through the year, learning as much as 90% of all synced userIDs.

Sensitive information (e.g., gender, birth dates) is sometimes passed to the syncing domain along with the userID.