

Towards Identifying Networks with Internet Clients Using Public Data

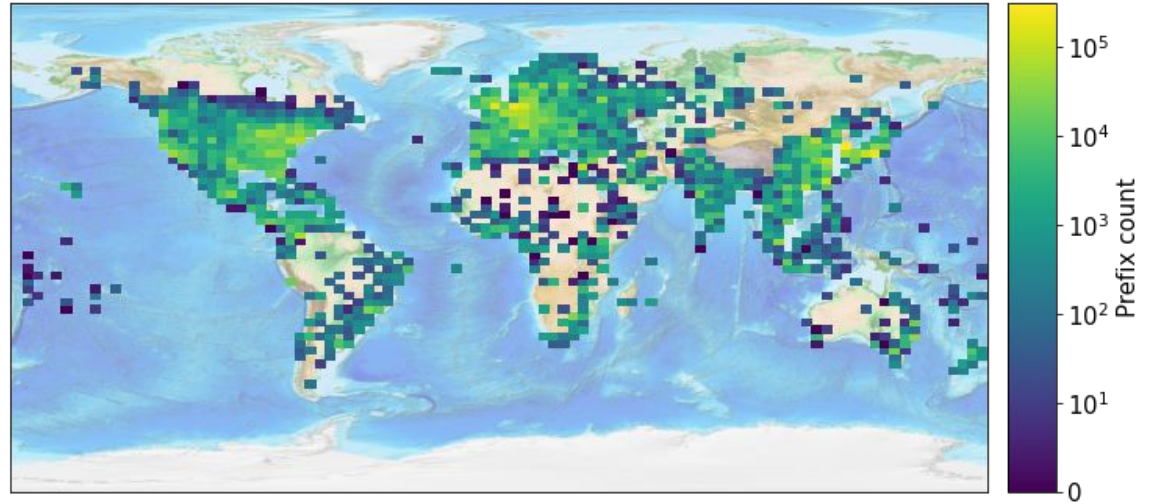
Weifan Jiang^{†*}, Tao Luo^{†*}, Thomas Koch[†], Yunfan Zhang[†],
Ethan Katz-Bassett[†], Matt Calder^{‡†}



*: primary authors

We measured web client activity globally

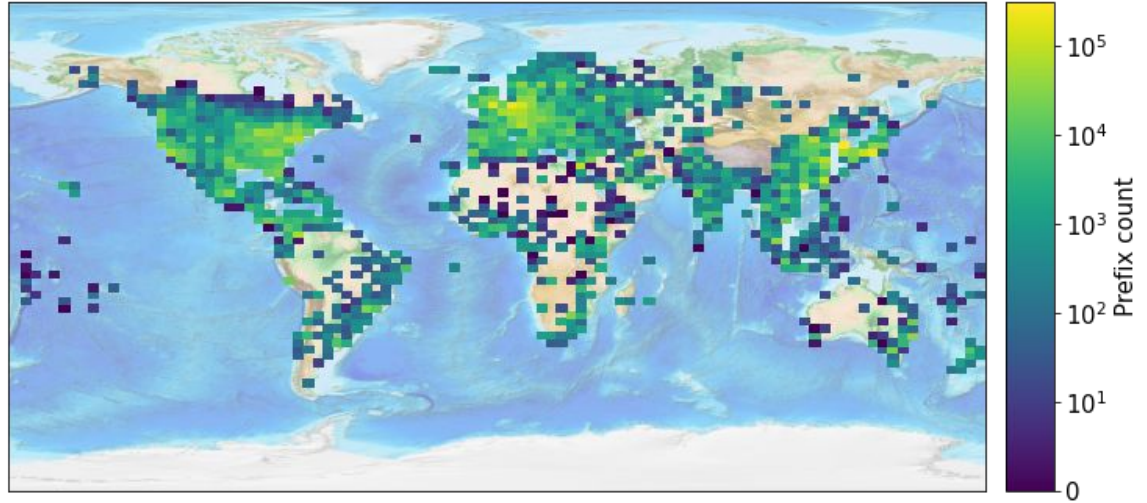
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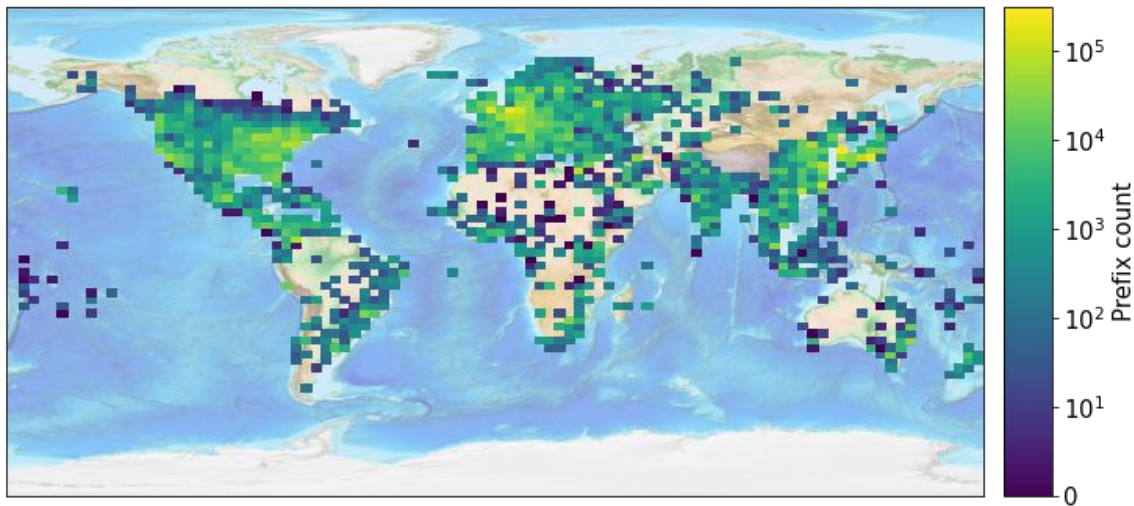
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These prefixes...

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- Good coverage that rivals the set of prefixes with client activities seen by Microsoft



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But we sometimes lack the data/tools to do so :(

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 - only provides coarse AS-level granularities

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1. focus on client-driven activity
2. use replicable methods
3. prefix-level granularity at global scale

Contributions

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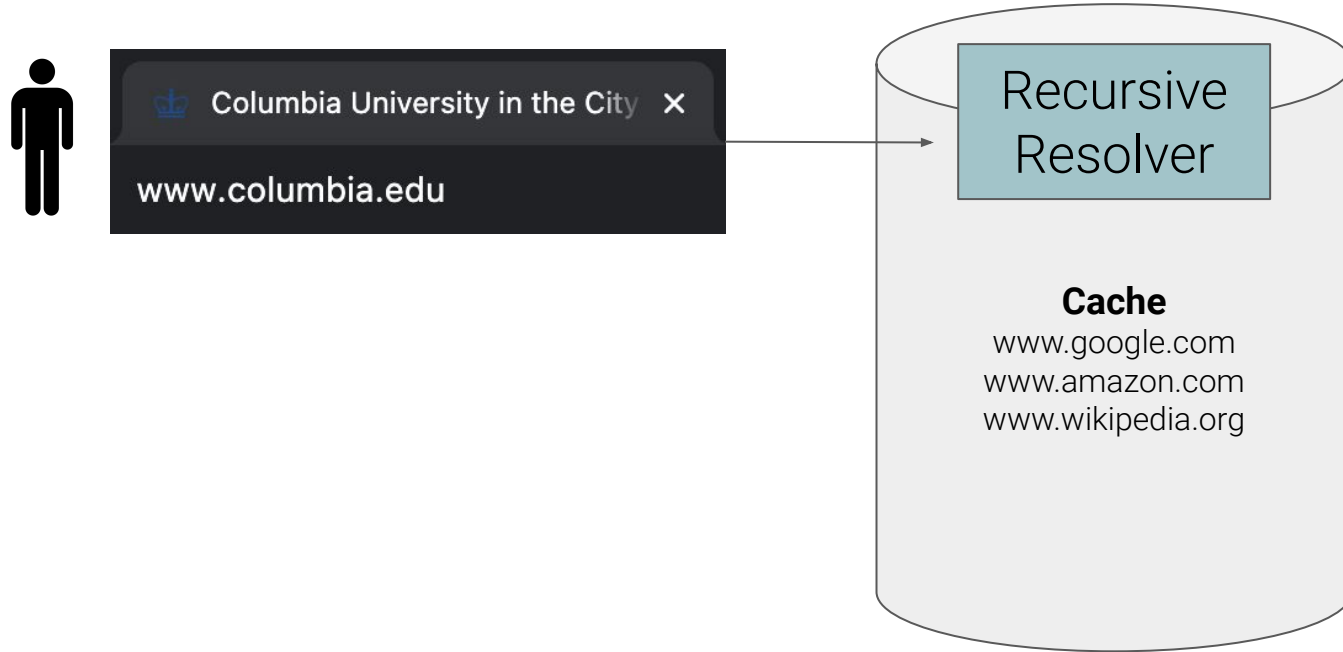
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 - CACHE PROBING (will be covered in this talk)
 - DNS LOGS (please refer to our paper for details)

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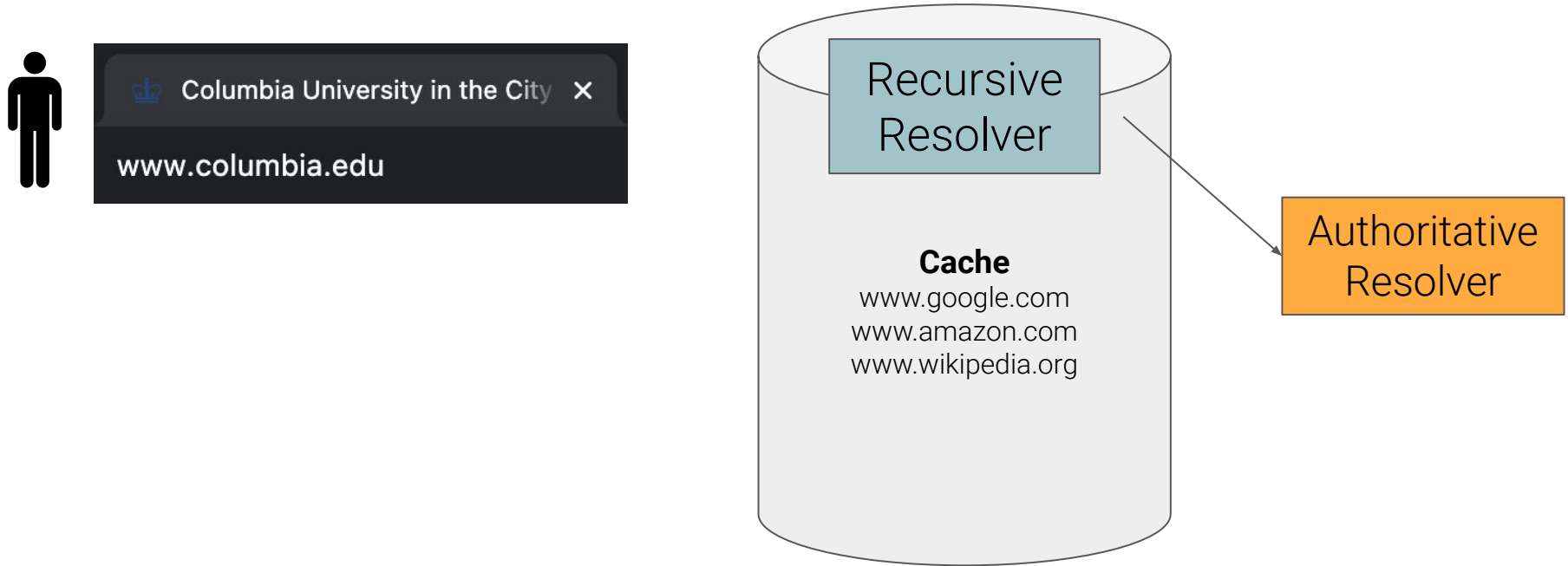
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 - CACHE PROBING (will be covered in this talk)
 - DNS LOGS (please refer to our paper for details)
2. Cross-comparison with the public APNIC dataset and the privileged Microsoft data to show our methodologies achieve good global coverage

Inferring client activity by DNS cache snooping

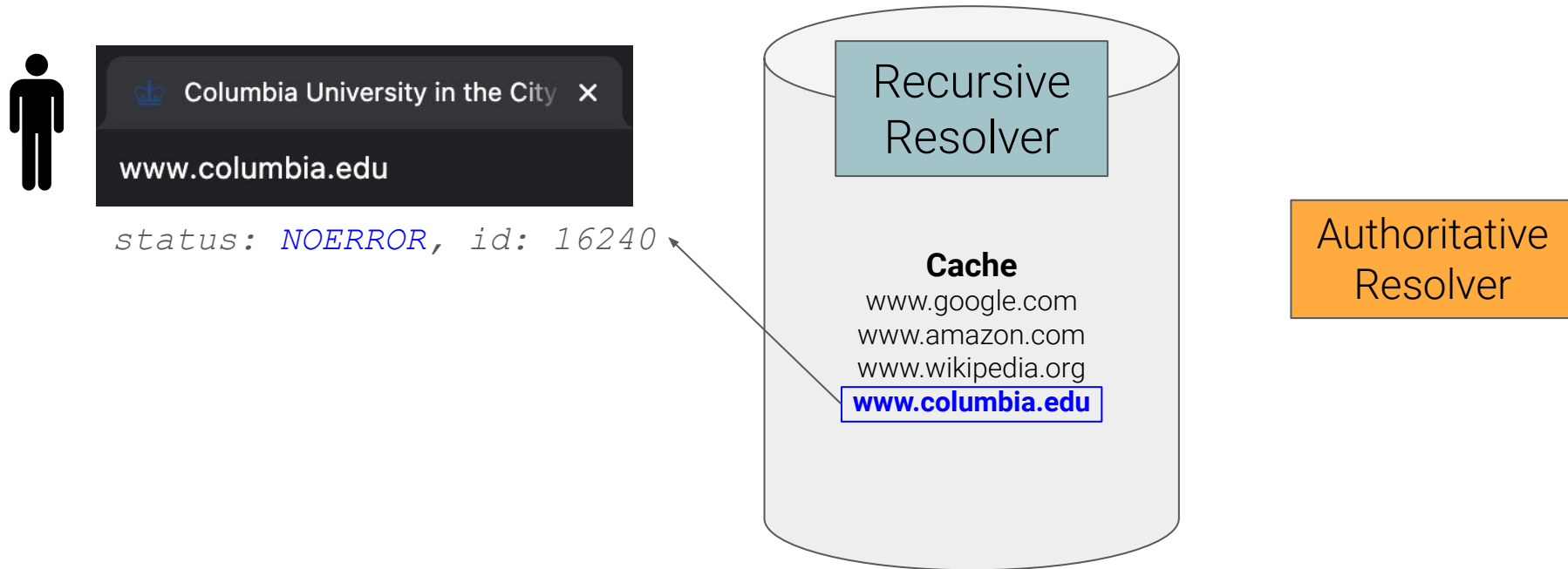
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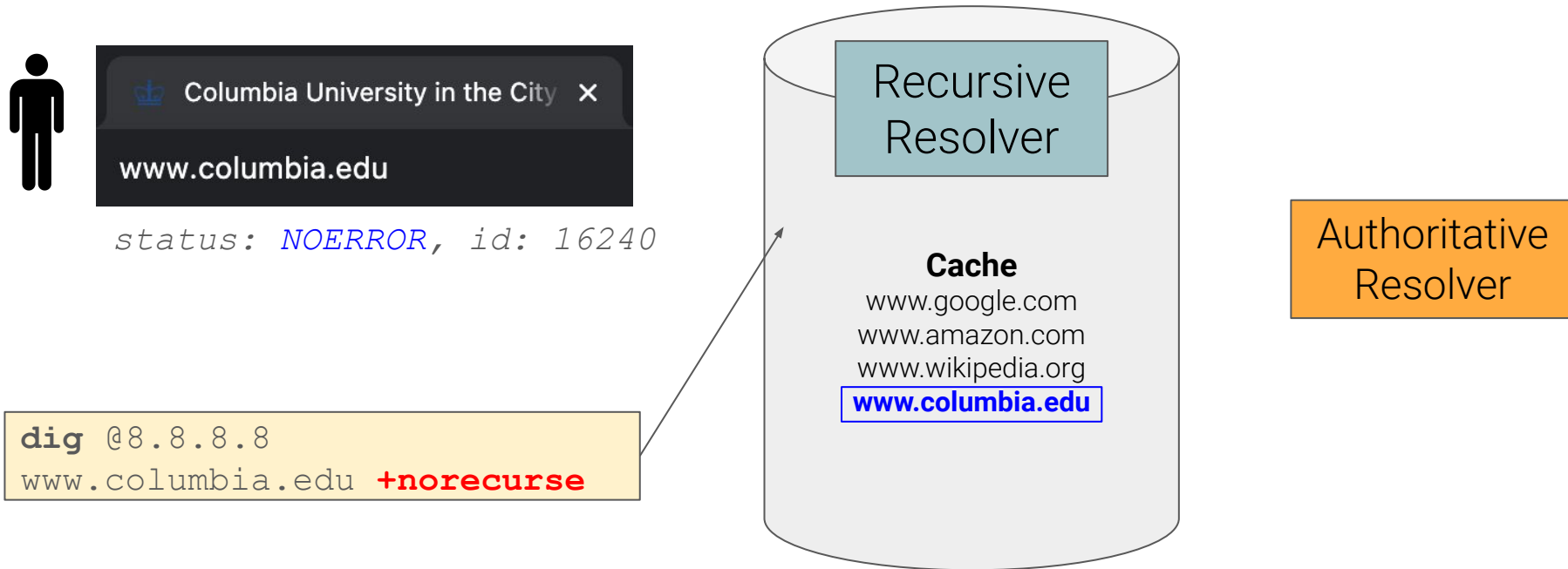
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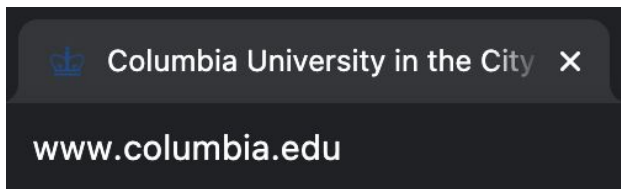
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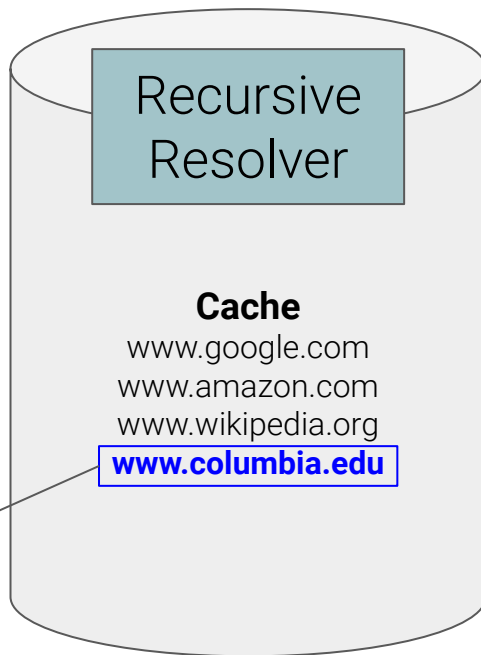


status: NOERROR, id: 16240

```
dig @8.8.8.8
```

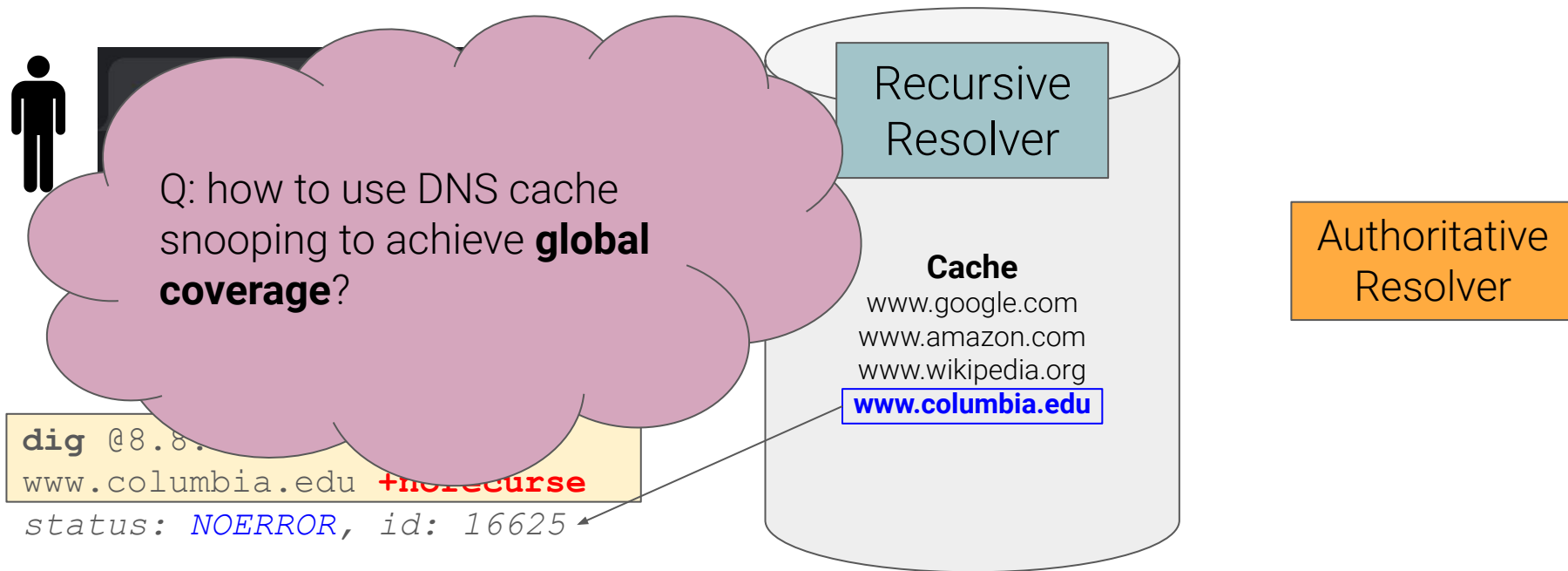
```
www.columbia.edu +norecurse
```

```
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```



Authoritative
Resolver

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Inferring client activity by DNS cache snooping



Q: how to use DNS cache snooping to achieve **global coverage**?

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Recursive
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Idea: send DNS queries to recursive resolvers in ISPs around the world!

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Recursive Resolver

~~Idea: send DNS queries to recursive resolvers in ISPs around the world!~~

Does **NOT** give global coverage!

Recursive Resolver



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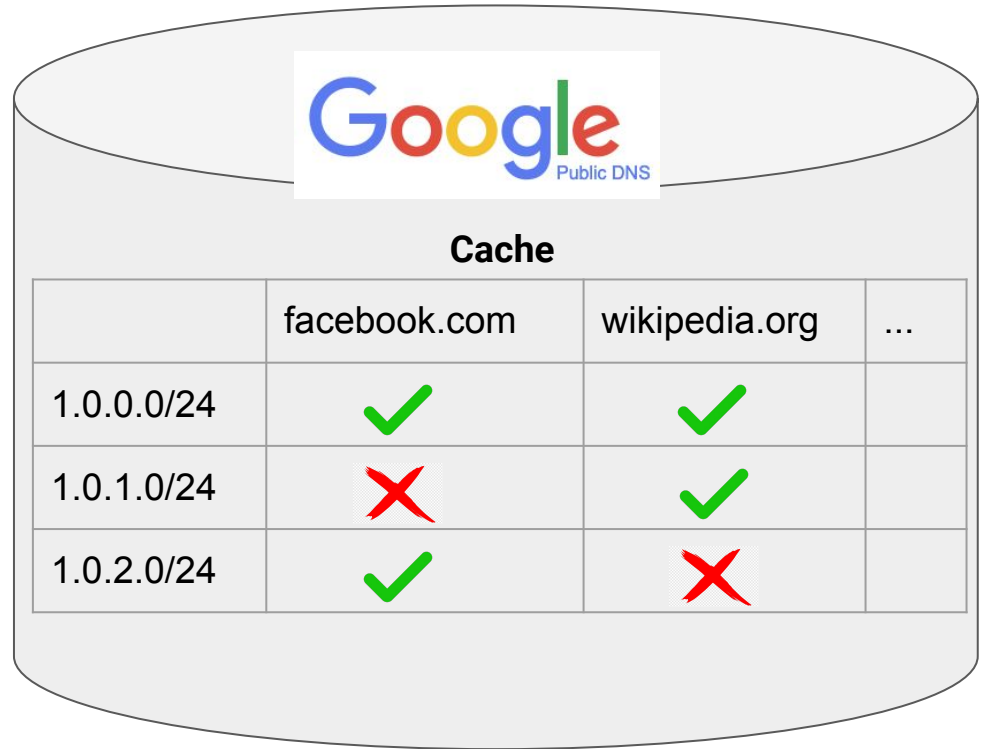
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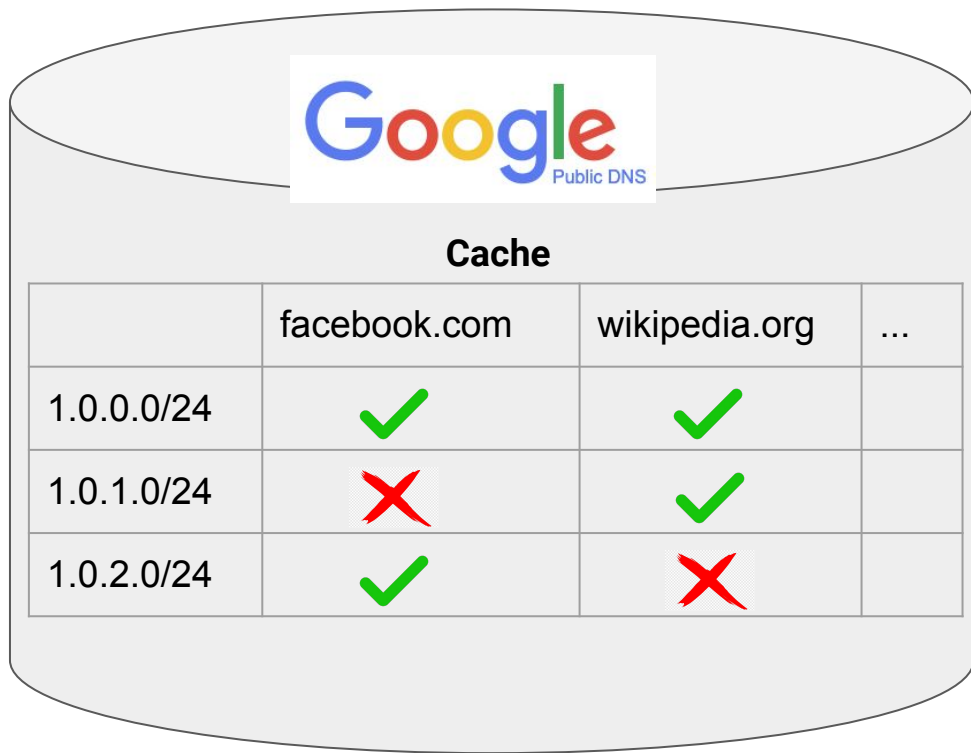
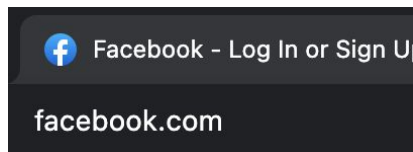
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 - if client specifies an ECS prefix, Google Public DNS will use that instead of client's own prefix
- Google Public DNS maintains separate cache entry per client prefix



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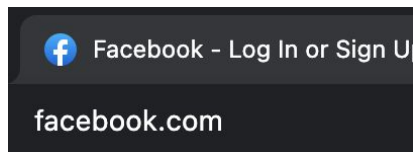
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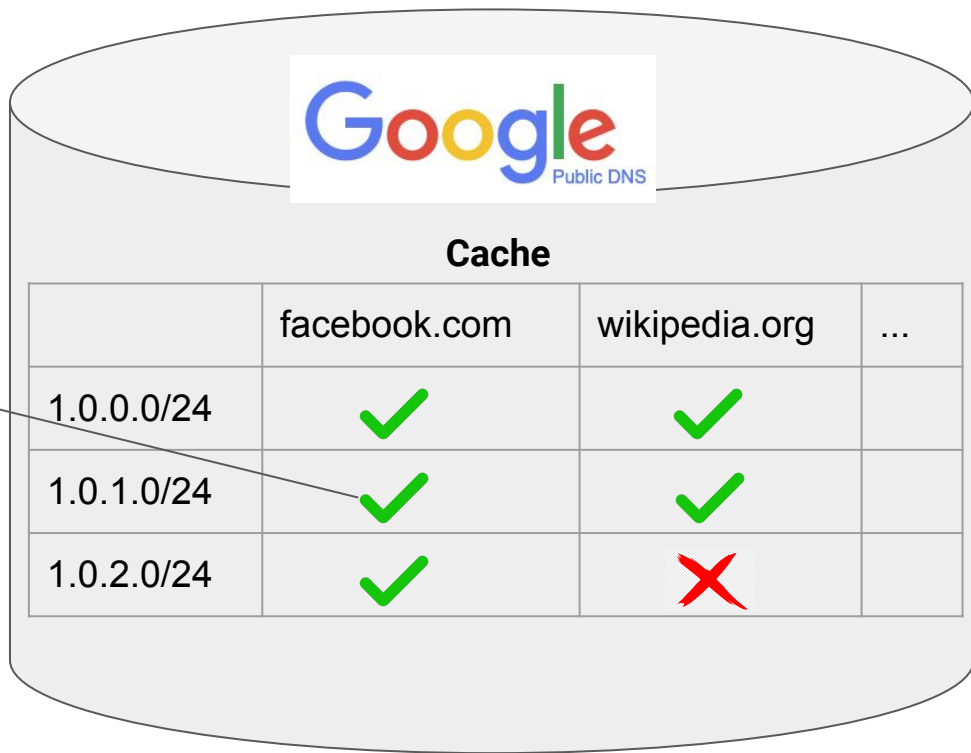
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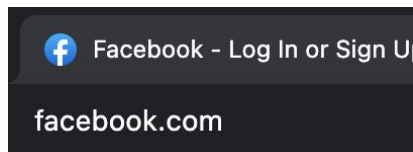
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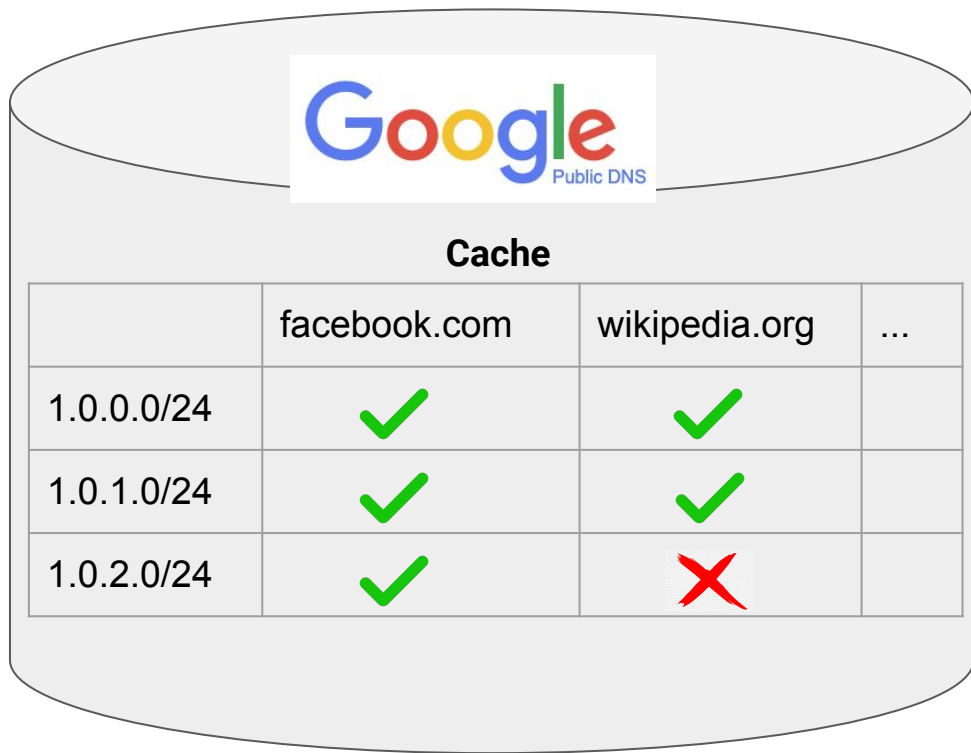


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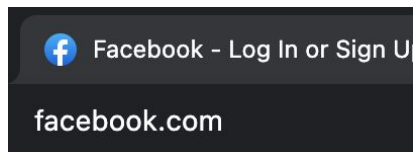
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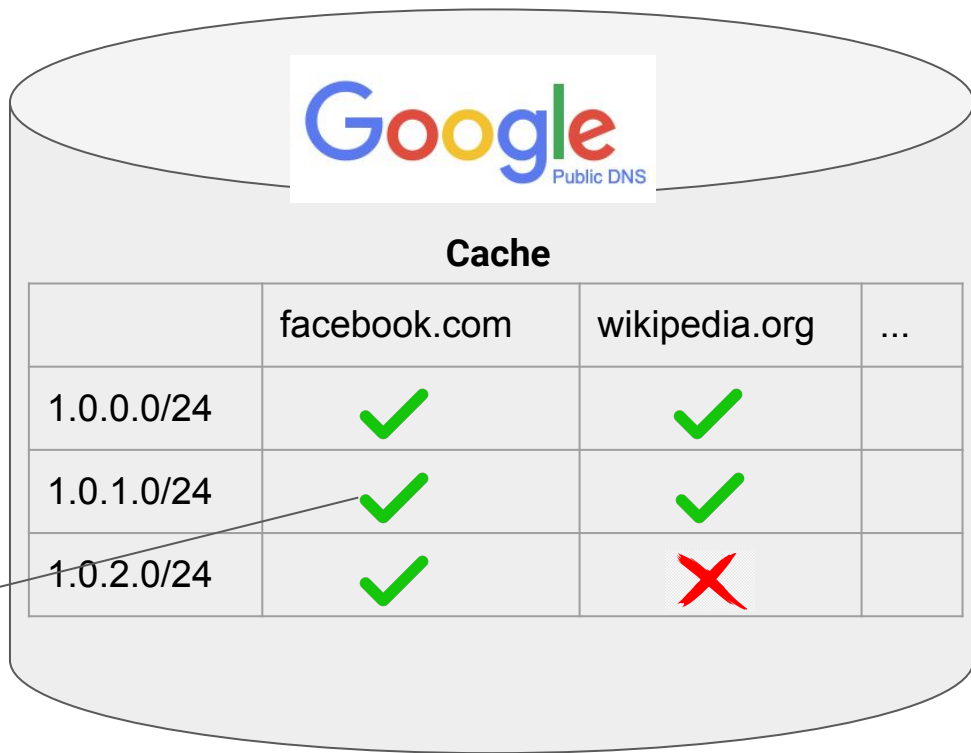
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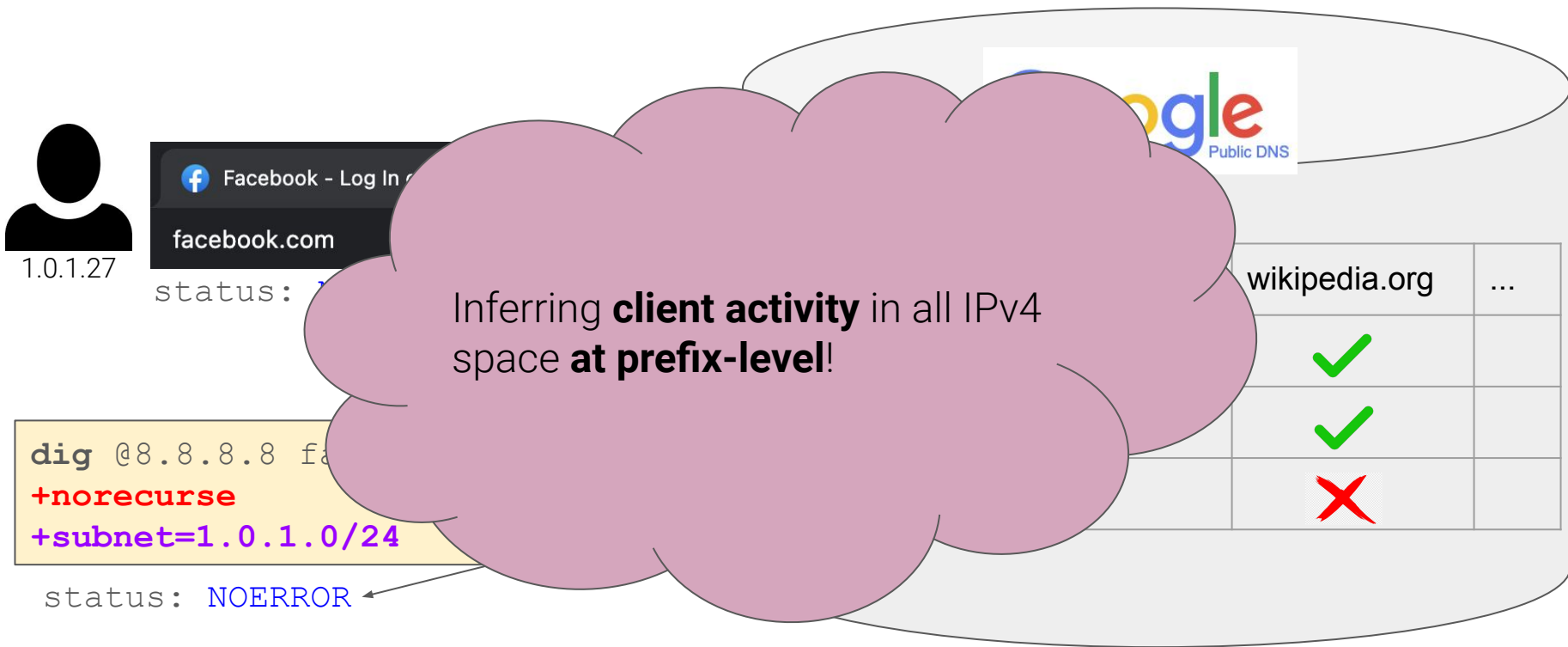
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dig @8.8.8.8 facebook.com  
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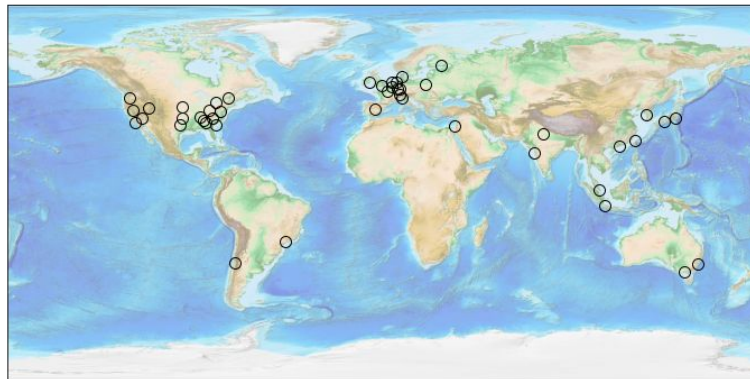


Geo-distributed Measurements

- Our goal: global coverage

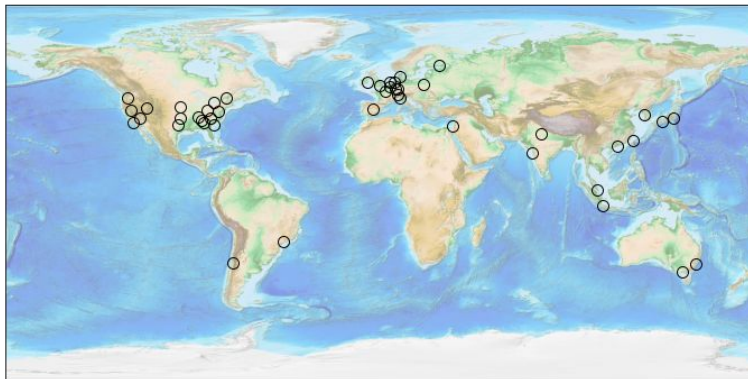
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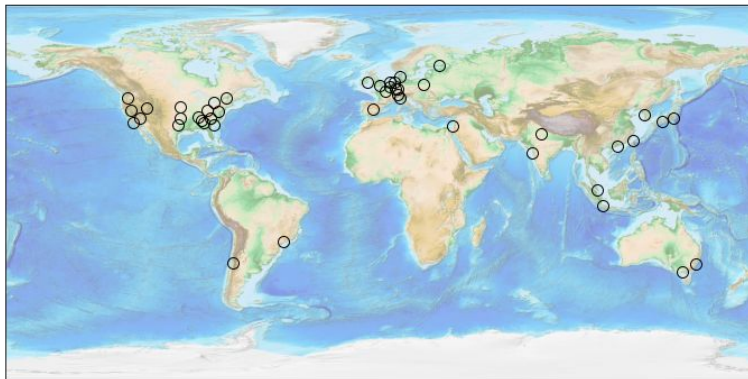
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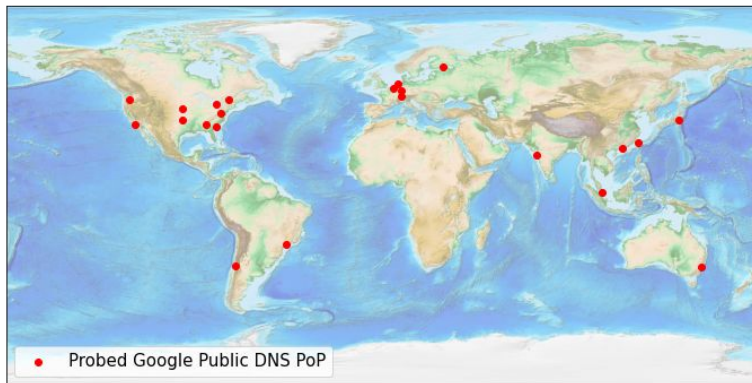
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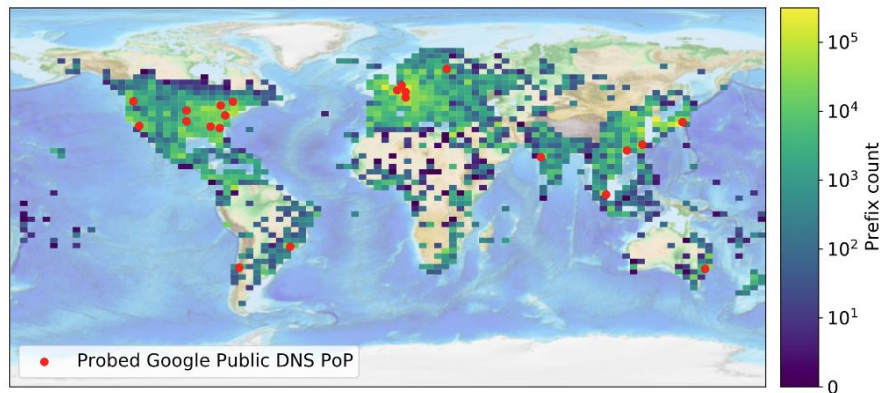
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Global coverage of IPv4 prefixes!

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Our methodologies produce a **good approximation** of privileged dataset

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Towards a traffic map of the Internet ([HotNets'21](#))

Thomas Koch, Weifan Jiang, Tao Luo, Petros Gigis, Yunfan Zhang, Kevin Vermeulen, Emile Aben, Matt Calder, Ethan Katz-Bassett, Lefteris Manassakis, Georgios Smaragdakis, Narseo Vallina-Rodriguez

What Google Public DNS PoPs do we hit?

- `dig @8.8.8.8 o-o.myaddr.l.google.com -t TXT` tells us the IP address of the particular Google Public DNS PoP reached.
- Google publishes the IP range and the closest airport code for each PoP¹.

```
$ dig @8.8.8.8 o-o.myaddr.l.google.com -t TXT
...
;; ANSWER SECTION:
o-o.myaddr.l.google.com. 60      IN      TXT     "74.125.18.67"
```

1: [Google Public DNS: Frequently Asked Questions.](#)

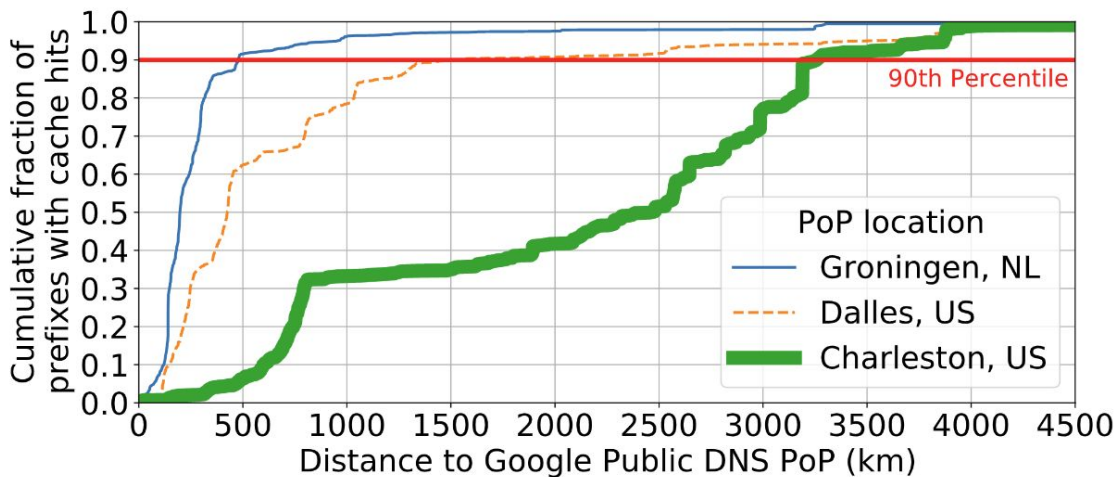
Google Public DNS

Locations of IP address ranges Google

34.64.0.0/24	icn	172.21
34.64.1.0/24	icn	172.21
34.64.2.0/24	icn	172.21
34.101.0.0/24	cgk	172.21
34.101.1.0/24	cgk	172.21
34.101.2.0/24	cgk	172.21
74.125.16.128/26	bom	172.21
74.125.16.192/26	yyz	172.21
74.125.17.128/26	cbf	172.21
74.125.17.192/26	dfw	172.21
74.125.18.0/25	iad	172.21
74.125.18.128/26	iad	172.21

Assigning prefixes to vantage points

- We **randomly selected** 78,637 prefixes and **queried them at all vantage points**.
- For each vantage point, we compute the **geographical radius** that would include **90%** of all cache-hit prefixes in the sample.
- We **use the 90% radius** to assign all 15,527,909 public /24 prefixes to vantage points to reduce probing overhead.
- For prefixes not assigned to any PoPs with above heuristics, we **assign it to the closest 2 PoPs**.



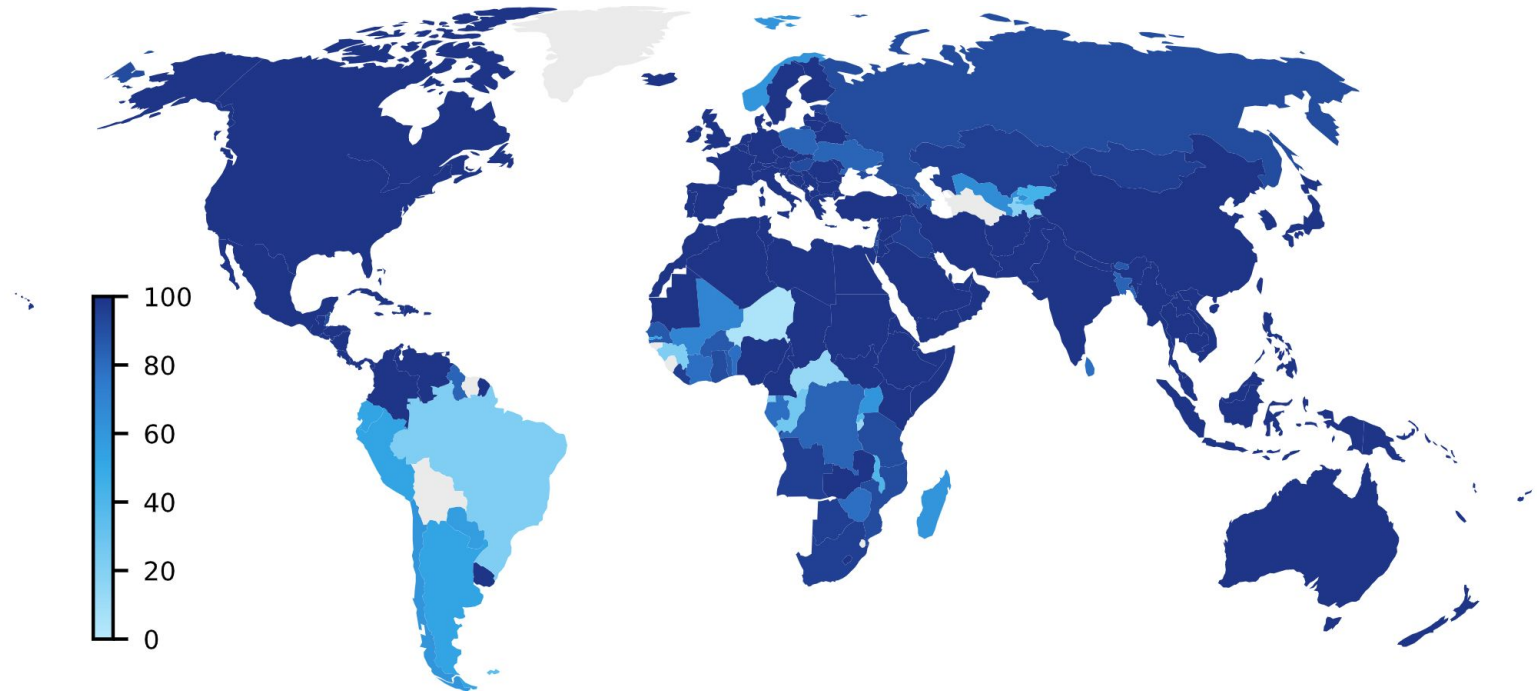
Domain names

Domain	Alexa Topsite Rank - Global (as of Sep 2021)
	1
	2
	7
 WIKIPEDIA	13

Methodology 2: DNS LOGS

- Chromium detects DNS interception by **querying for random strings** of 7-15 lowercase letters:
 - when browser starts, and
 - when device's IP address or DNS configuration changes
- These queries **should not result in cache hit** at recursive resolvers (due to lacking a valid TLD such as ".com"), so the queries **go to a DNS root server**.
- We identify chromium queries using a heuristic that **randomly generated strings should have few collisions** across all roots in one day (based on empirical study, 7 is a good threshold).
- We look for queries matching this pattern **in the DITL traces**. Those queries contain the **IP address of the querier**, which is generally the recursive resolver used by the Chromium client.

Coverage Analysis



Colorscale: percentage of country's APNIC Internet user population seen by CACHE PROBING.

Validate with Microsoft data

What our methodologies saw:

- The ASes found by us are responsible for most of the Microsoft traffic.
- Implication: the ASes missed by us are very small.

	# of ASes seen by MSFT	Volume of traffic to MSFT
CACHE PROBING	55.5%	94.9%
DNS LOGS	59.9%	97.4%
CACHE PROBING U DNS LOGS	77.2%	98.8%

Result analysis

What activity did APNIC miss, but seen by us?

- ASDB¹ (from IMC'21!) tells us what categories an AS belongs to.
- Out of the ASes detected by our methods but missed by APNIC:
 - 10,998 (**39.5%**) are **Internet Service Providers (ISPs)**
 - Outside of ISPs, 4,823 (**17.4%**) are **hosting/cloud providers** → may reflect **non-human web clients**
 - Outside of ISPs, 1,723 (**6.2%**) are **schools** → likely **host human users**

1: Ziv et al. "[ASdb: A System for Classifying Owners of Autonomous Systems](#)".