Who Filters the Filters:

Understanding the Growth, Usefulness and Efficiency of Crowdsourced Ad Blocking

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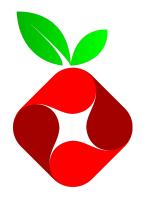
The talk in a slide...

Many web and privacy tools use crowdsource lists















How these lists are maintained is poorly understood
 Who decides what goes in? What comes out? What exceptions exist? etc...

- Web measurement of EasyList
 - Most popular list
 - Mostly "dead weight", 90.16% of rules unused
 - 10k website measurement over 2+ months → practical optimizations
 - How do advertisers & trackers respond?

Overview

Context and Background

What, why and how of EasyList

Methodology

Web scale measurement over two months

Measurement Results

Whats used and unused, rule lifecycle, how do trackers respond, etc?

Applications

Mobile and extension optimizations

Discussion and Conclusion

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Context and Background

- EasyList is the most popular list
- Targets ads and tracking from advertisers
- Text format, RegEx-like format
- EasyList is a large project, 15 years of contributions
- Targets English and "global" sites
- Many different rules, acting on different layers

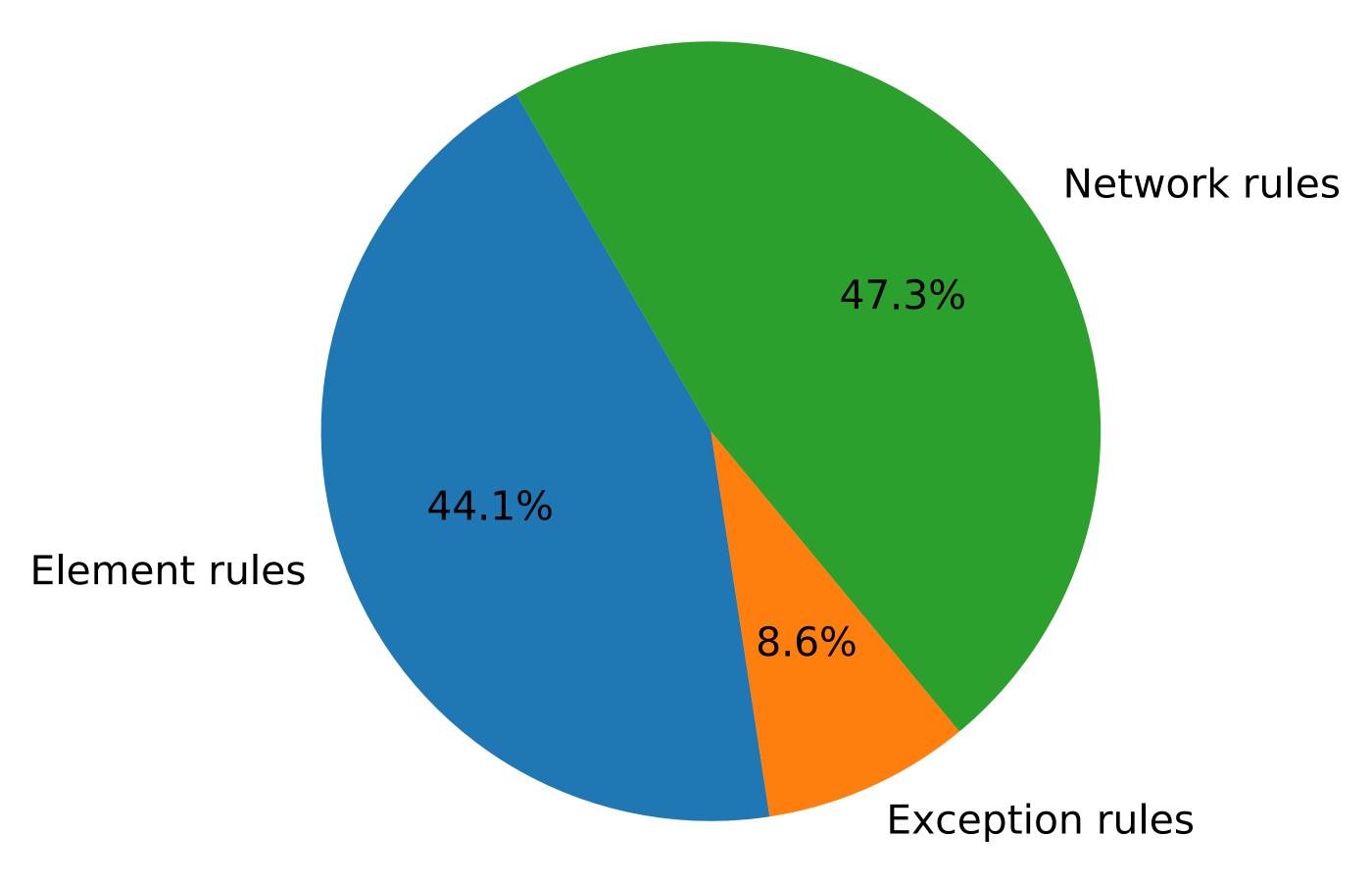
EasyList Types of Rules

Network rules||example.org/ad

Element rulessite.com###iframe

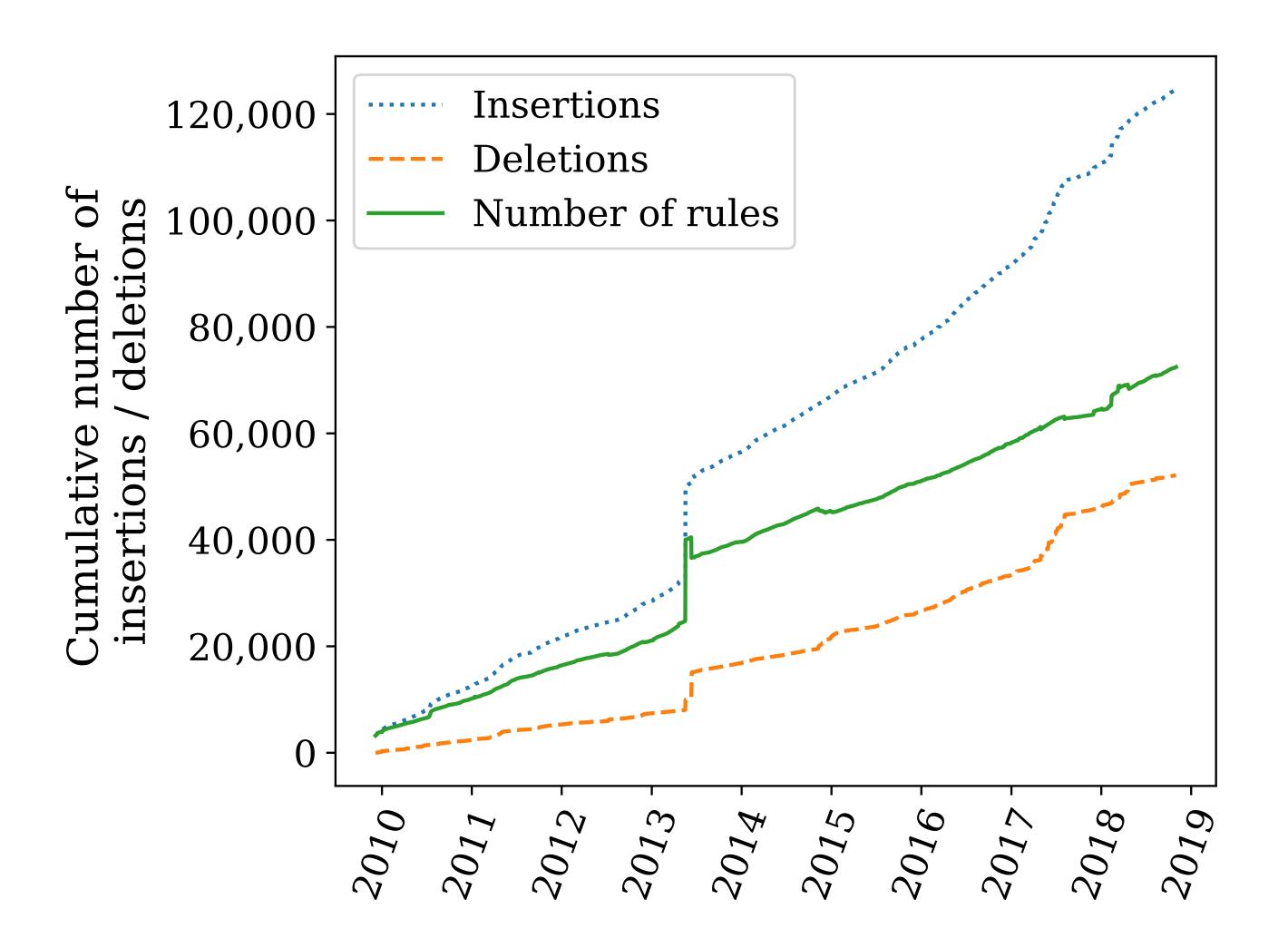
Exception rules
 @@ | example.org/advice

Filters||example.org^script



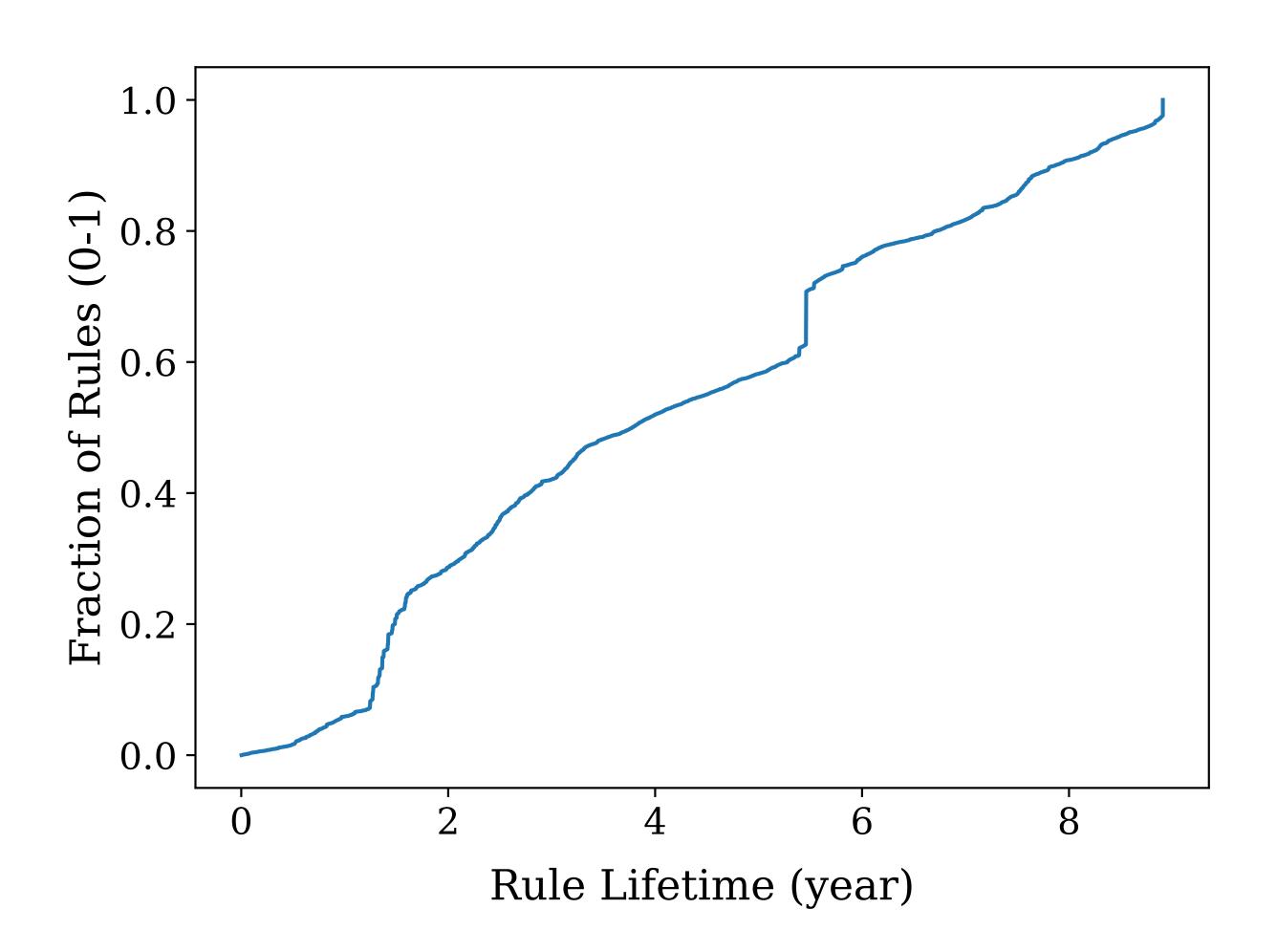
EasyList Over Time

- 2005: Started by Rick Petnel
- 2009: Moves to GitHub
- 2013: Merges with "Fanboy's list"
- 2019: Reaches 72,469 rules
- 2020 (May): Shrinks to ~69k



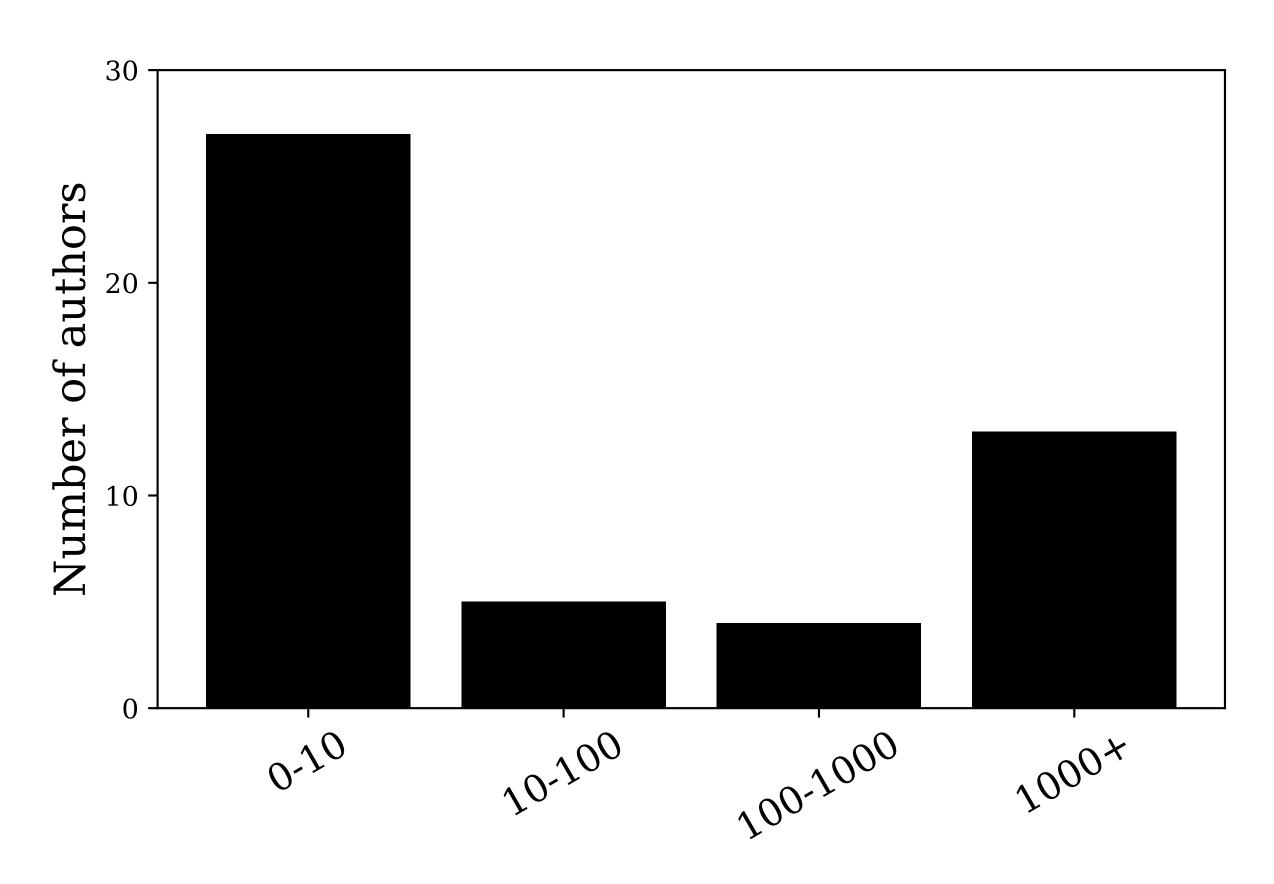
Rule "Life Cycle"

- Measurement of how long a rule stays in the List
- Measured using git commit history
- ~50% rules remain for > ~4 years



Who Contributes To EasyList

- From forum and GitHub
- Five main contributors
 76.87% of commits
- Many small contributors
 65.3% of contributors made <=
 100 commits



Number of commits

Also in the Paper...

- How commit history was tracked across project structure changes
- How often commits are made
- How other tools use EasyList
- Tooling details

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- Practical Applications
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- Discussion and Conclusion

Measurement Goals

Broad Goal: Understand how EasyList and the web interact

Sub Goals:

- How is "rule usefulness" distributed?
- Relationship between rule age and rule usefulness?
- How to advertisers respond to being listed?

Methodology

Instrument a browser:

Record all network requests when visiting a page

Representative automated crawl

Both popular and unpopular websites

Apply EasyList to crawl data:

Determine what would be blocked if that day's EasyList was applied

Browser Instrumentation

• Stock Chromium:

Current stable version of Chromium at time of measurement

Puppeteer automation:

Record all URLs fetched, along with response type, hash and body size

Passive instrumentation:

No changes to page loading or resource requesting

No measurement of page contents:

Omitted measurements of element hiding rules

Representative Automated Crawl

Web domain selection:

"Popular": Alexa 5k

"Unpopular": Random selection from Alexa 5,000-1m

Page selection:

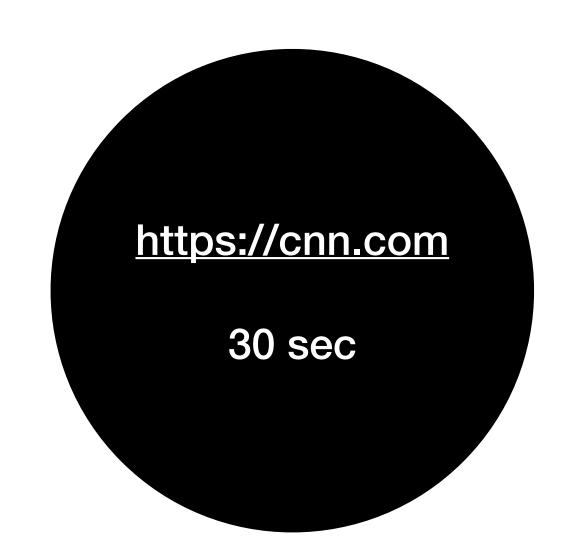
Measured landing page, and three same-eTLD+1 links

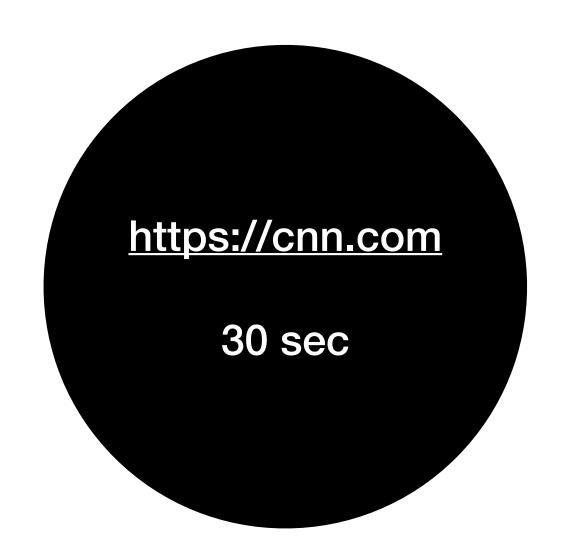
Measurement times:

- Every day for 74 days
- Measured each page for 30 seconds

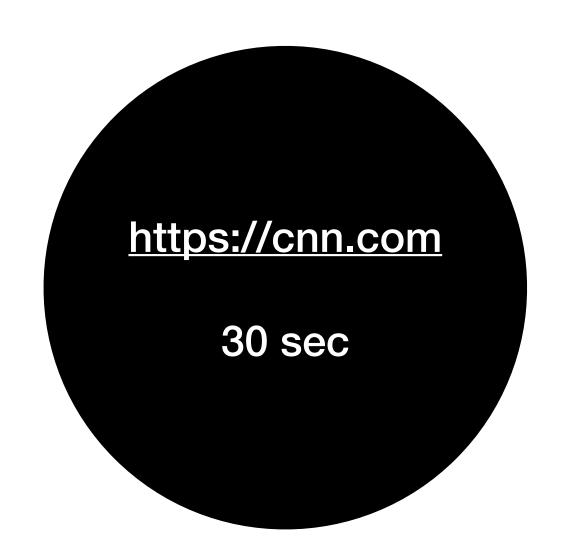
Passive measurement:

No changes to page loading or resource requesting

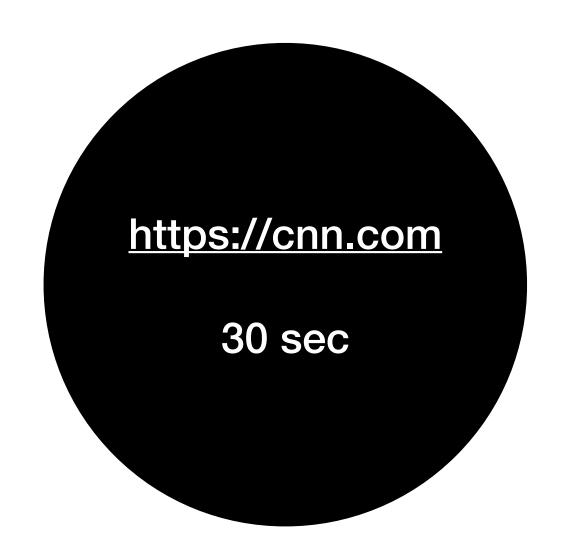


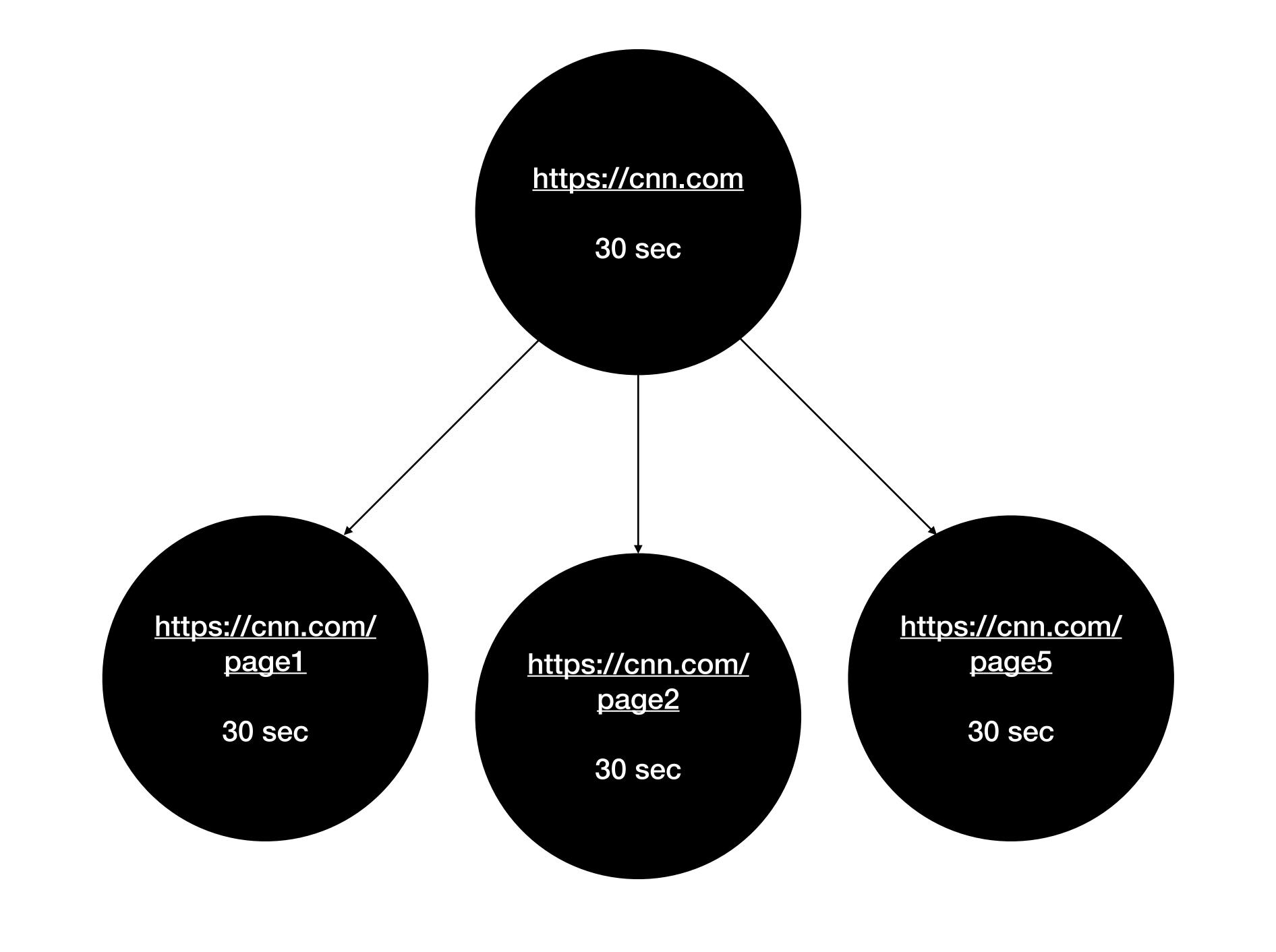


```
<a href="https://advertiser.com">
<a href="https://cnn.com/page1">
<a href="https://othersite.org">
<a href="https://cnn.com/page3">
<a href="https://neat.advertiser.com">
<a href="https://cnn.com/page2">
<a href="https://cnn.com/page2">
<a href="https://youtube.com">
<a href="https://cnn.com/page5">
<a href="https://cnn.co
```



```
<a href="https://advertiser.com">
<a href="https://cnn.com/page1">
<a href="https://othersite.org">
<a href="https://cnn.com/page3">
<a href="https://neat.advertiser.com">
<a href="https://neat.advertiser.com">
<a href="https://cnn.com/page2">
<a href="https://youtube.com">
<a href="https://youtube.com">
<a href="https://cnn.com/page5">
<a href="https://cn
```





Fetched Page	Requested Sub-Resources	EasyList Rule

|--|

https://cnn.com

Requested Sub-Resources

EasyList Rule

Fetched Page

https://cnn.com

Requested Sub-Resources

- https://cnn.com/header.png
- https://cnn.com/ad/img.png
- https://doubleclick.com/iframe
- https://cnn.com/js/script.js

•

EasyList Rule

Fetched Page

https://cnn.com

Requested Sub-Resources

- https://cnn.com/header.png
- https://cnn.com/ad/img.png
- https://doubleclick.com/iframe
- https://cnn.com/js/script.js
- •

EasyList Rule

- •
- */ad/*
- ||doubleclick.com^
- ...

On Omitting Element Rules

Noted network and exception rules
 Did not include element (i.e., cosmetic) rules

Reasoning

- Measurement focus is on privacy and performance
- Highly variable and dependent on user interaction
- Many EasyList consuming tools also omit them (e.g., Privoxy, PiHole)

Summary

- Instrumented automated Chromium
- Visited 10k sites (5k popular, 5k unpopular)
- Recorded:
 - Domains visited
 - Subpages visited
 - Resource requests and responses
 - Matching EasyList network rules

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Study period:

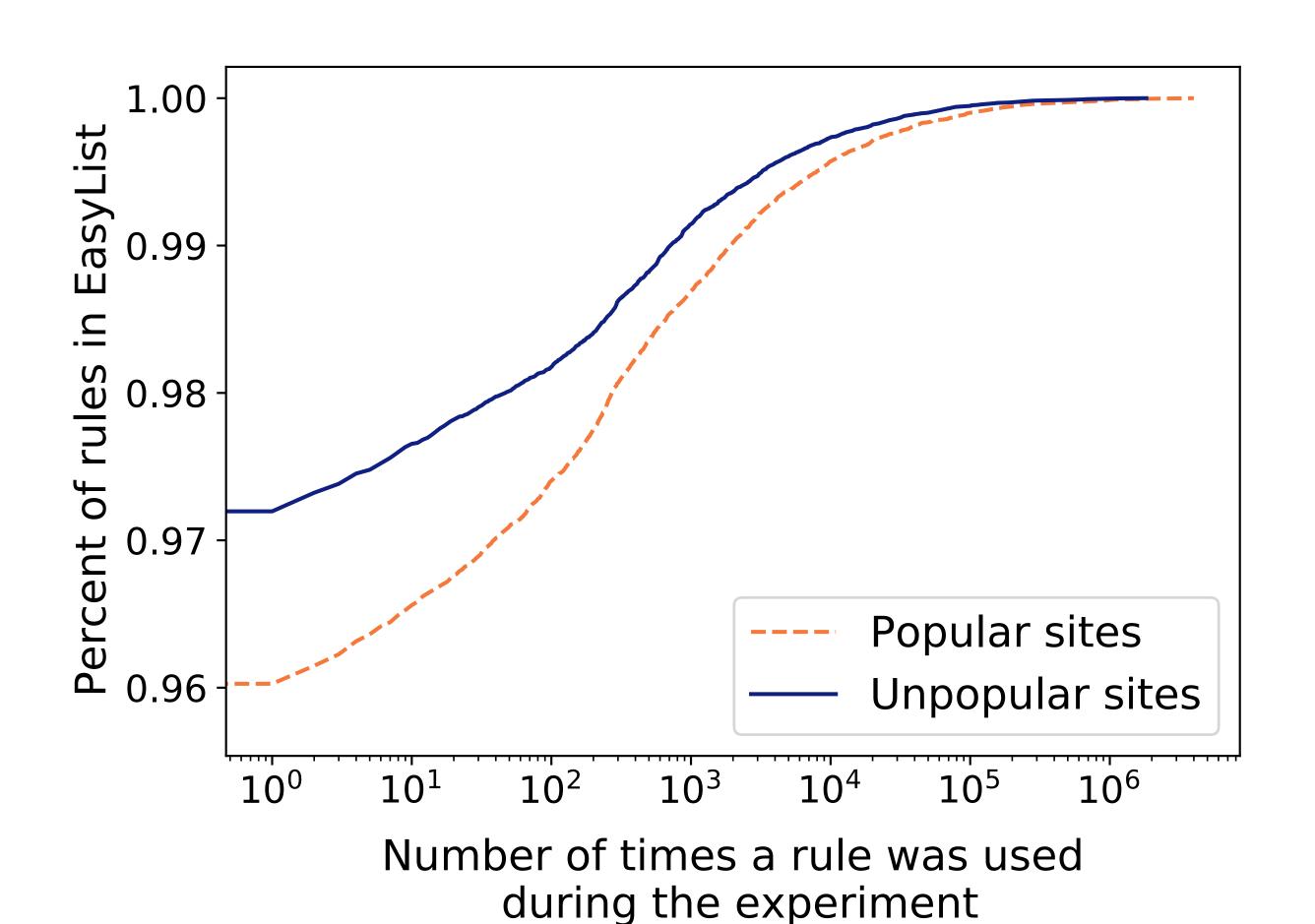
July 24th → October 5th, 2018

- Unresponsive domains:
 400 domains never replied
- 3.74 pages per domain:
 Difference b/c single
 page apps, CF CAPTCHA, etc.

Measurement	Counts
# days	74
# domains	10,000
# non-responsive domains	400
Avg # pages per day	29,776
Avg # pages per domain per day	3.74
Total # pages measured	3,260,479

Proportion of EasyList Rules Used

- Measurement
 % of rules used at least once during the entire experiment
- Most rules were not used
 90.16% never applied
 5.39% used >= 100 times
- Domain popularity not sig



Relationship of Rule Age and Usefulness

Measurement:

Are newer rules more useful?

Answer:

Mixed, but mostly no

- New and old rules are used at least once equally
- Most blocking is done by old rules

	Added during experiment	Added before experiment
Absolute #	2,002	37,826
% used at least once	9.45%	9.84%
Use frequency (of those used at least once)	0.65 per day	6.14 per day

- Same resource, multiple URLs, only some blocked
- Non-blocked URLs occurred after relevant rule
- Compare URLs to observe why not blocked

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- Non-blocked URLs occurred after relevant rule
- Compare URLs to observe why not blocked

<u>To</u>	<u>T</u> 1	<u>T</u> 2
<pre>a.com/ad-script.js b.com/ad-script.js c.com/ad-script.js</pre>		

- Same resource, multiple URLs, only some blocked
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<u>To</u>	<u>T</u> 1	<u>T</u> 2
<pre>a.com/ad-script.js b.com/ad-script.js c.com/ad-script.js</pre>	New Rule: /ad-script.js	

- Same resource, multiple URLs, only some blocked
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- Compare URLs to observe why not blocked

```
\frac{T_0}{a.com/ad-script.js} \\ a.com/ad-script.js \\ b.com/ad-script.js \\ c.com/ad-script.js \\ d-script.js \\ ad-script.js
```

Methodology:

- Same resource, multiple URLs, only some blocked
- Non-blocked URLs occurred after relevant rule
- Compare URLs to observe why not blocked

```
T<sub>0</sub>

a.com/ad-script.js
b.com/ad-script.js
c.com/ad-script.js
```

```
<u>T</u>1
```

```
New Rule:
/ad-script.js
```

<u>T</u>2

```
a.com/ad-script.js
b.com/ad-script.js
c.com/sneaky.js
```

```
a.com/ad-script.js
b.com/ad-script.js
c.com/ad-script.js
```

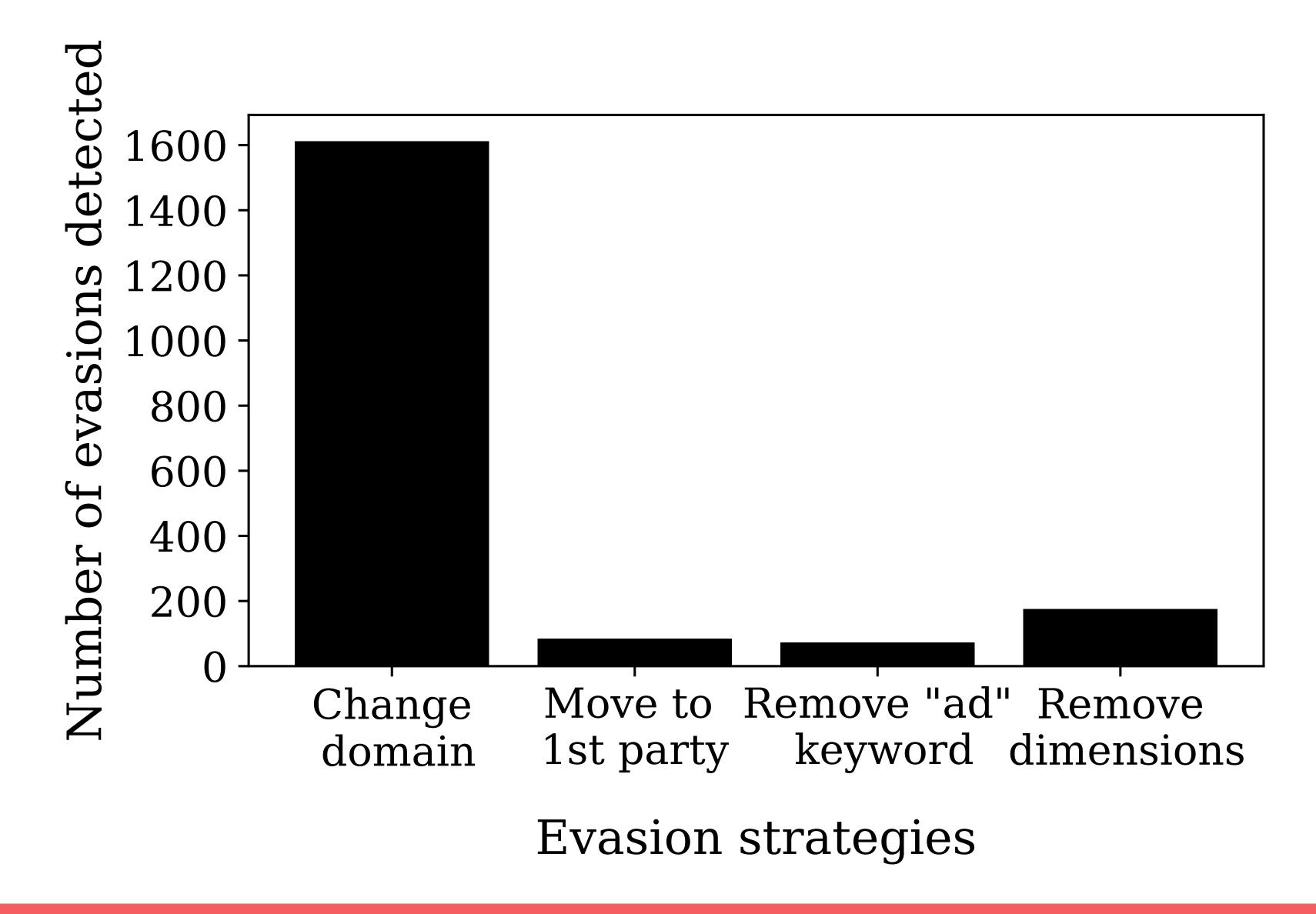
V.S.

c.com/sneaky.js

Changing domain:
 tracker.com/script.js → benign.com/script.js

- Move to 1st party: google-analytics.com/ga.js → cnn.com/ga.js
- Remove "ad" keyword:
 example.org/ads/shoes.png → example.org/images/shoes.png
- Remove dimensions:
 example.org/shoes-320x240.png → example.org/shoes-standard.png

Advertiser Reactions



Also in the Paper...

- How quickly advertisers respond to new rules?
 Most don't...
- Statistical correlation between rule age and use frequency
 Significant positive correlation
- Specific examples of filter list evasions We name names...

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Applications

- Mobile content blocking
 Fitting filter lists in mobile devices, performantly
- Improving performance of extensions
 Left for the paper

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Two related problems

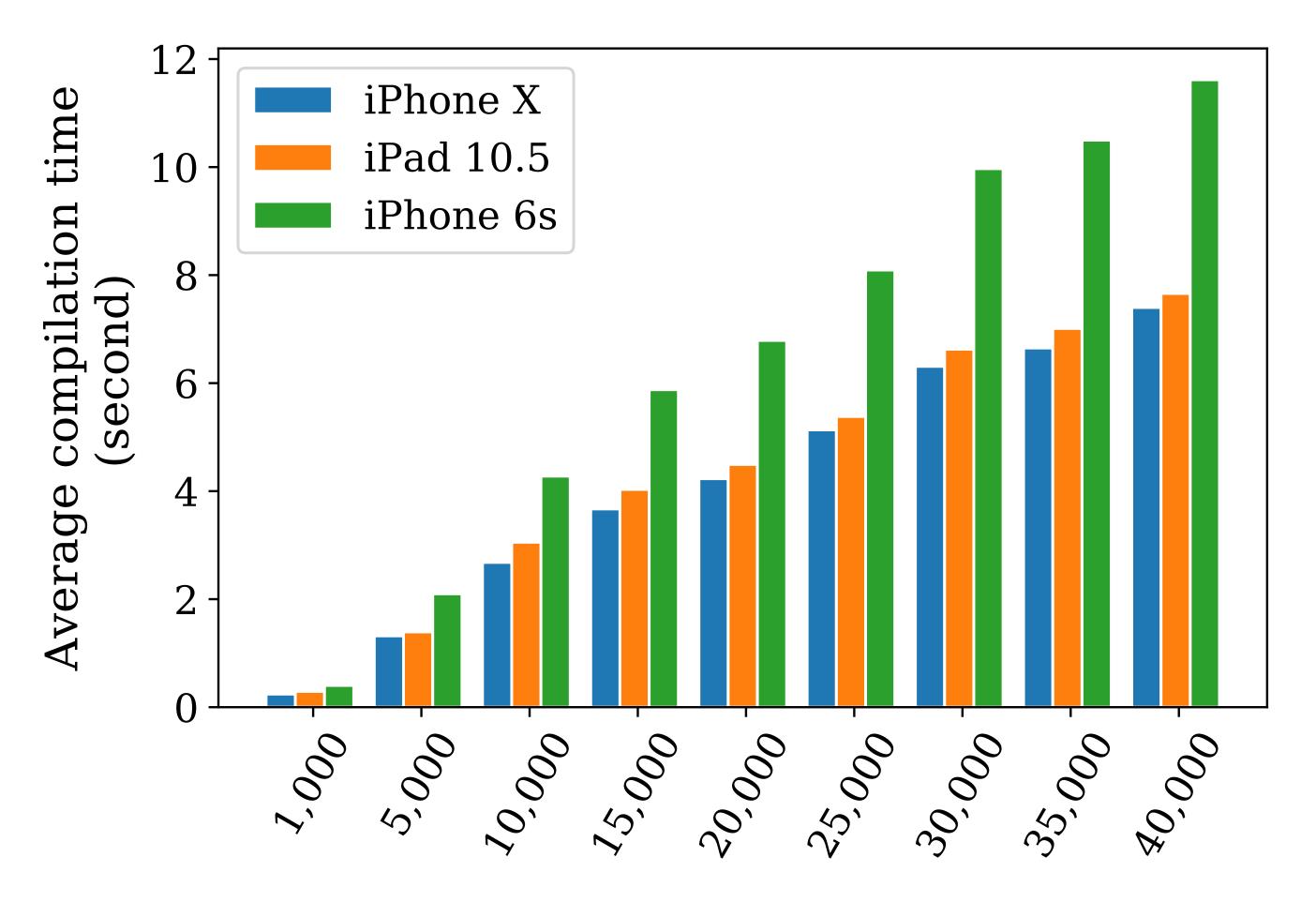
- iOS limits to 50k rules
- Compiling rules is slow on first load

Its not only EasyList...

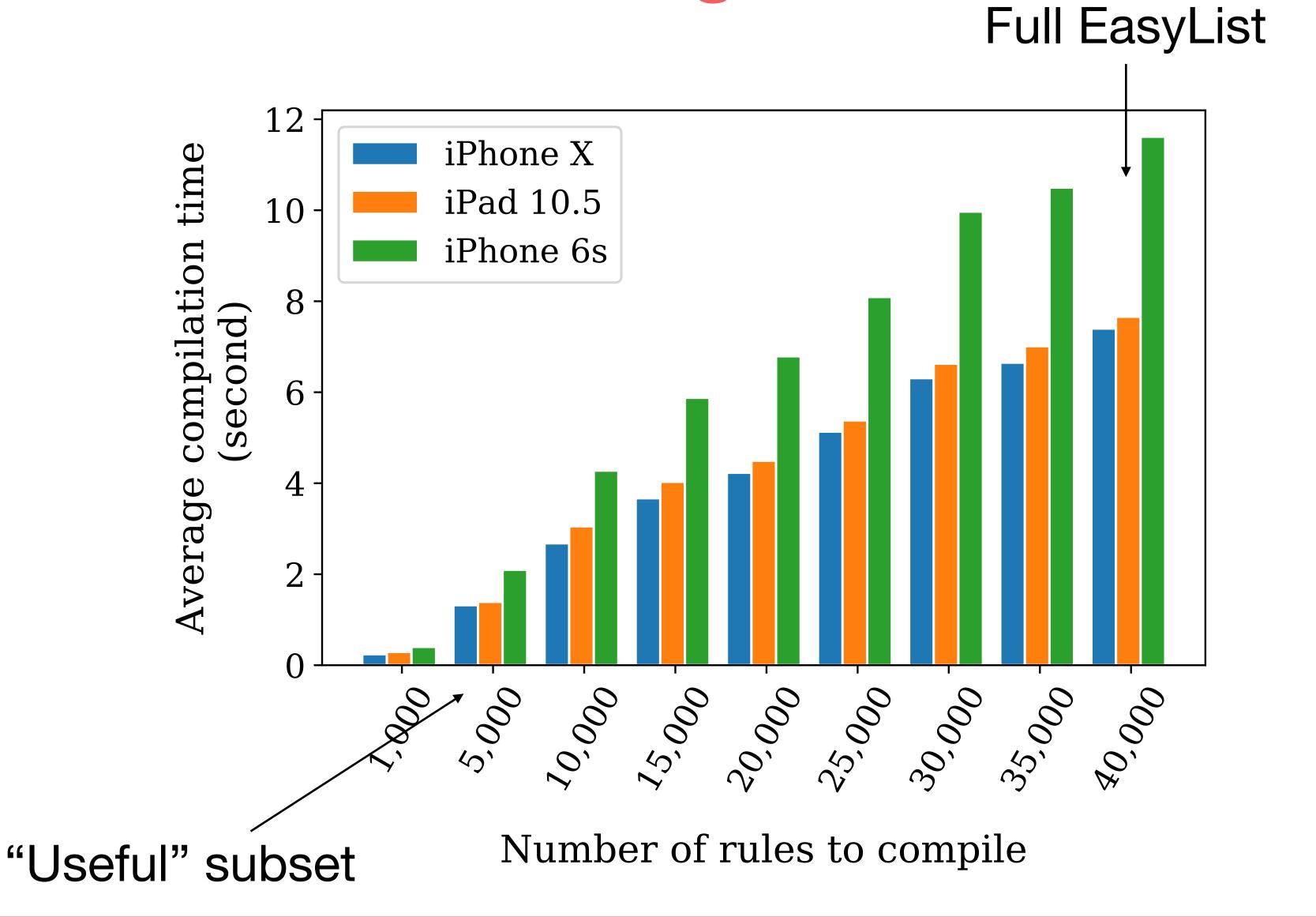
- EasyPrivacy
- Regional lists

Solution

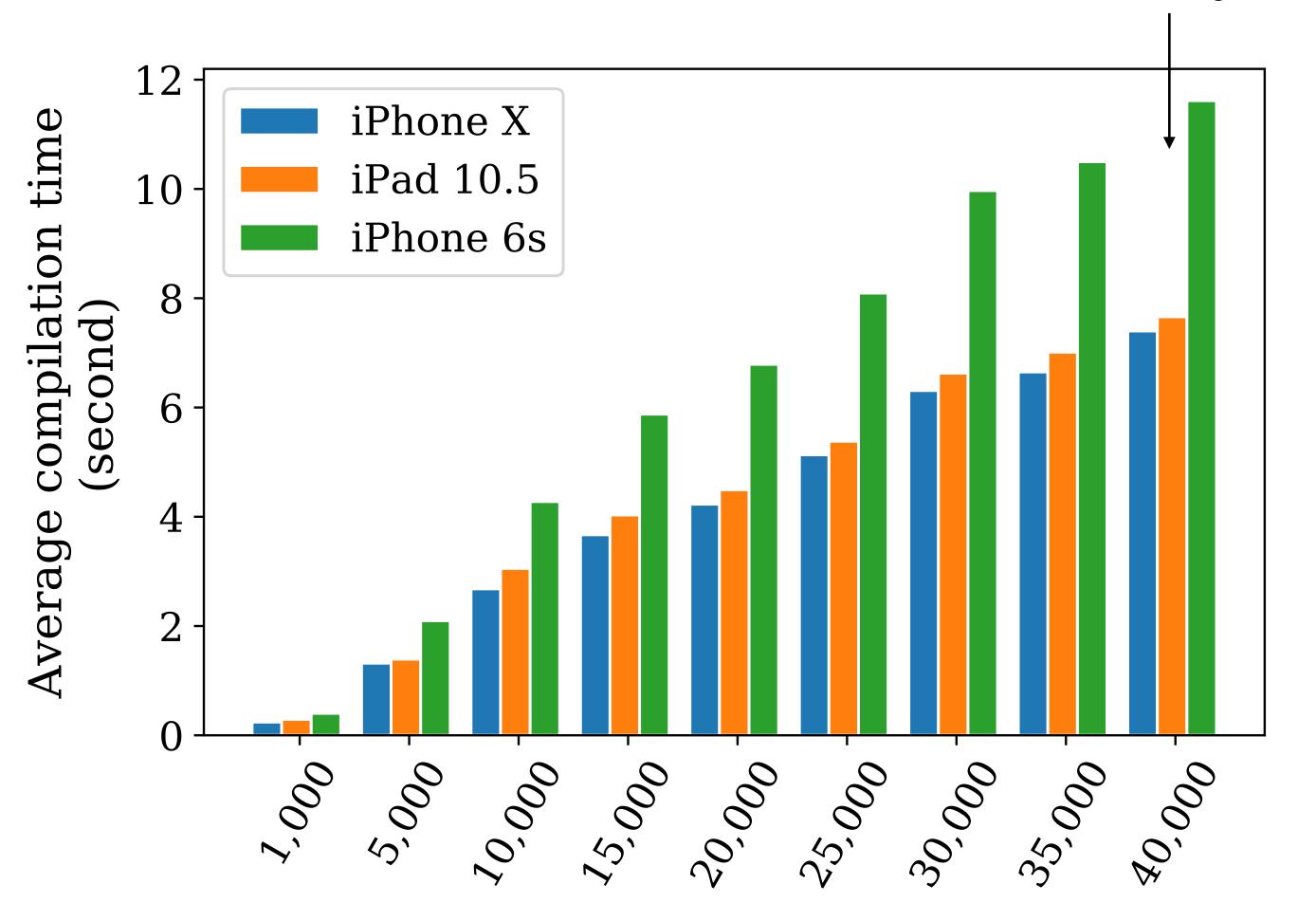
- Use crawl data to identify likely useful rules
- Only load those rules on iOS
- "Slim List"



Number of rules to compile



Full EasyList



Number of rules to compile

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Limitations And Future Work

Web site selection generalizability

We assume interactivity isn't vital We assume "shallow" pages are similar to "deep" pages

Web region and language generalizability

We assume measuring from US IP generalizes
We assume good division between English / global EasyList and regional lists

Varying resource blocking importance

We assume all blocking is equally useful We assume vital, security level protections are dealt with through other means

Summary

- First measurement of how EasyList affects the web
- Broadly used, maintained by five people
- >90% of EasyList provides little benefit
- Quantified taxonomy of filter list evasion
- Measurement allows for use on mobile

Summary and Thank You!

- First measurement of how EasyList affects the web
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