De-anonymizing Tor Onion Services with Flow Correlation Attacks

Diogo Barradas

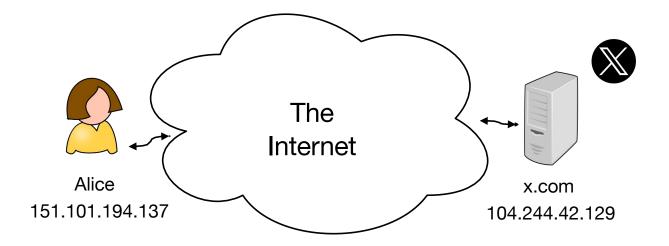
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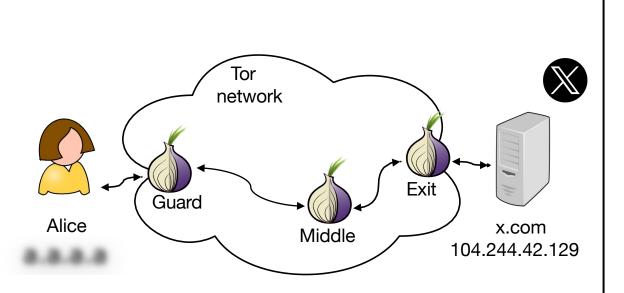
The need for online anonymity

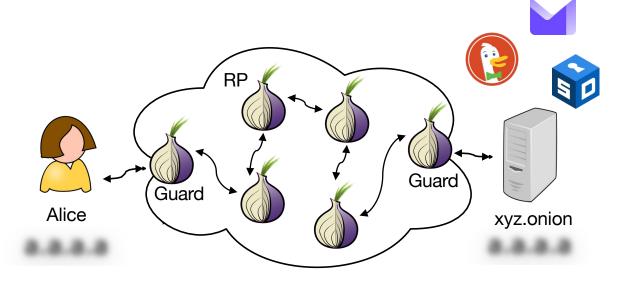
Internet users are at odds with pervasive tracking and online surveillance



Onion routing aims to provide online anonymity by sending users' network traffic through multiple relays

Preserving online anonymity with Tor





Client-side anonymity

Circuits to the "clearweb"

Client- and server-side anonymity

Circuits to onion services

Can we deanonymize Tor traffic? Should we?





+ Protect activists, whistleblowers, etc.



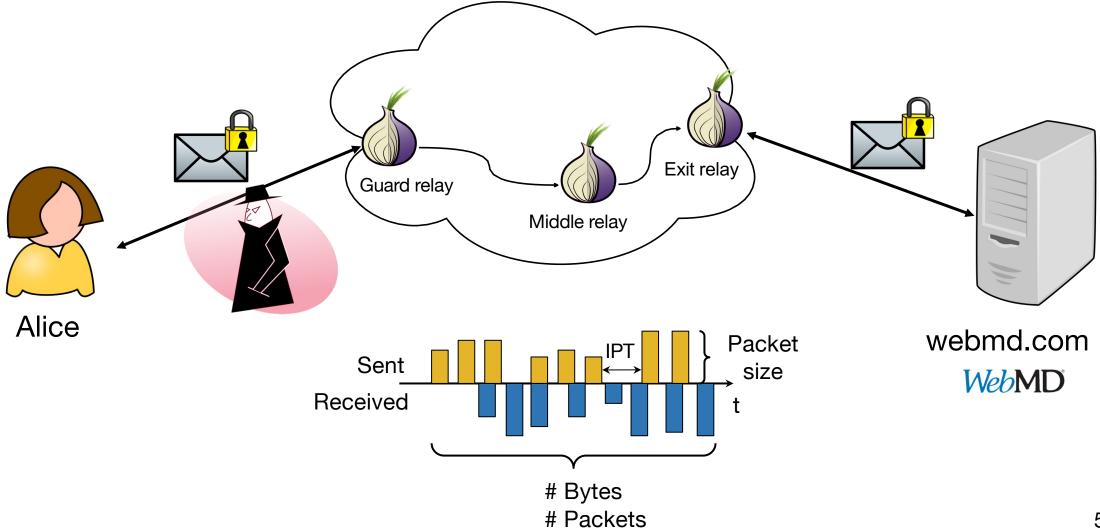
+ Catch (cyber)criminals





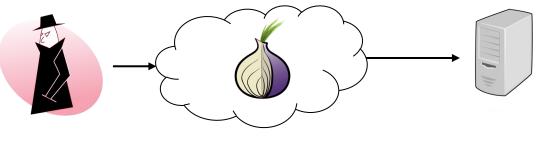


Encrypted Tor connections leak metadata

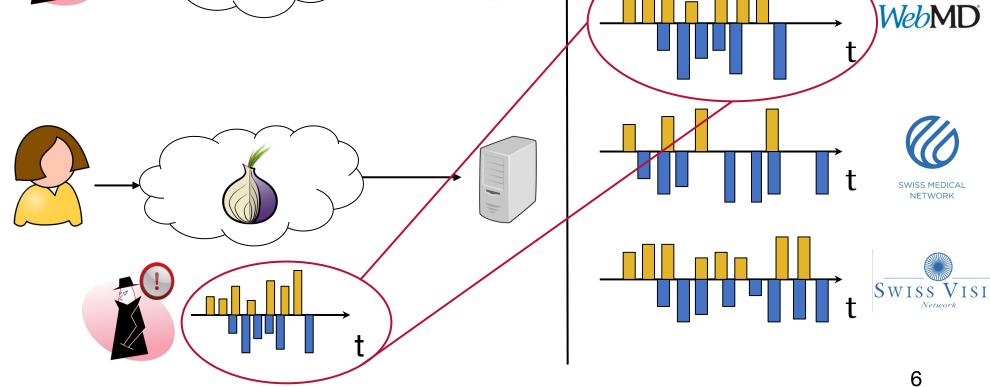


Deanonymizing clients' accesses via Tor (I)

Step 1: Build fingerprint database

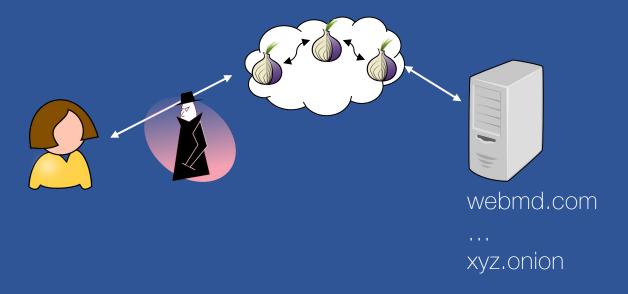


Step 2: Match Alice's traffic



Fingerprints Database

How can one deanonymize Tor traffic?

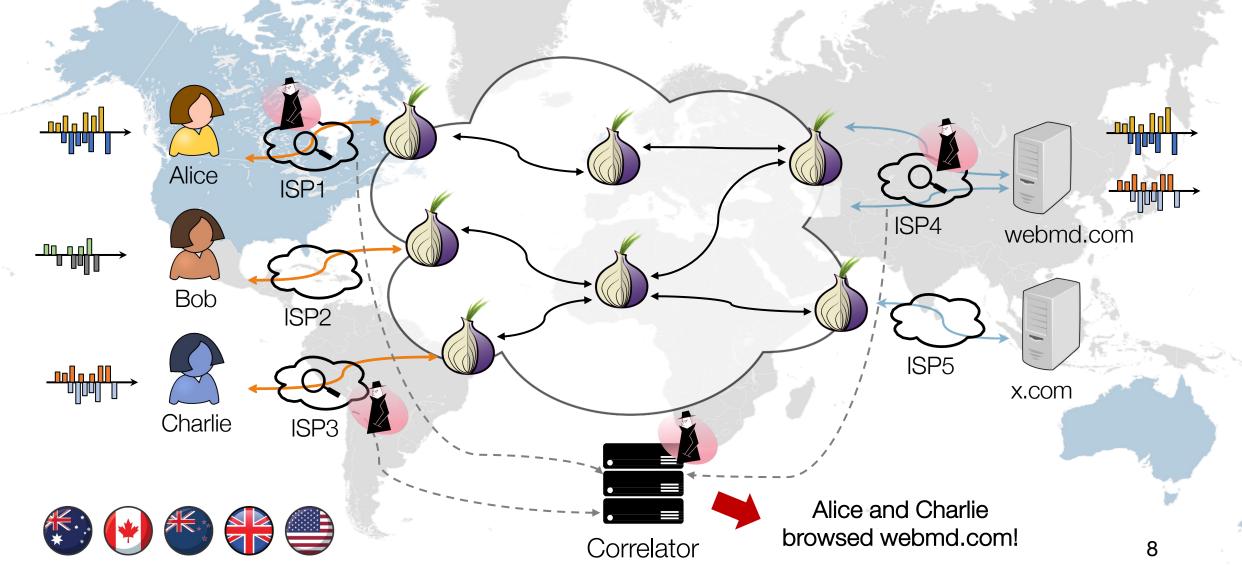


Is there another way?

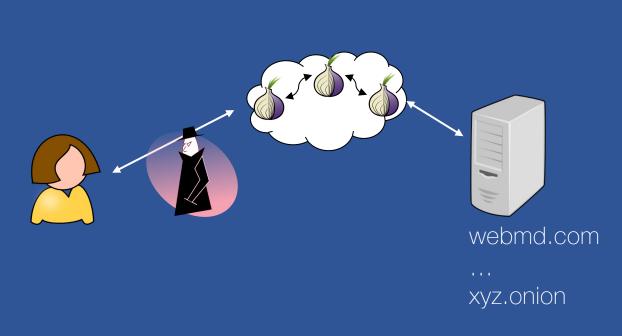
Website Fingerprinting (local adversary)

- + Can tell the website/.onion a client connects to
 - Requires a pre-built database
 - Cannot find a .onion's IP address

Deanonymizing clients' accesses via Tor (II)

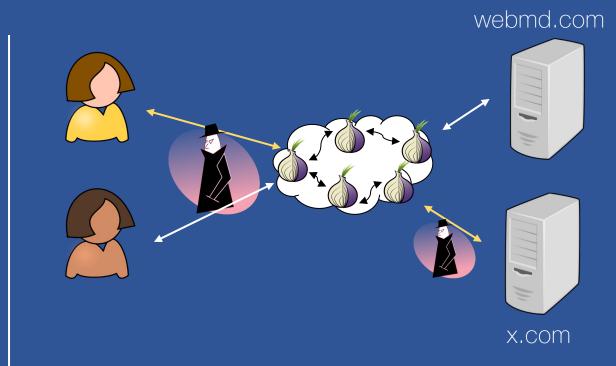


How can one deanonymize Tor traffic?



Website Fingerprinting (local adversary)

- + Can tell the website/.onion a client connects to
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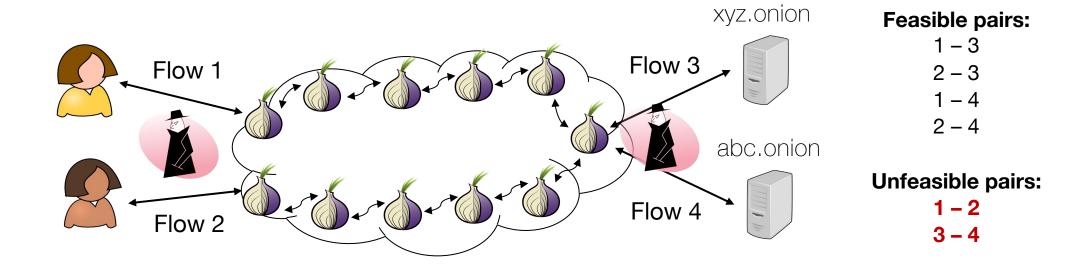


Flow Correlation (global adversary)

- + Can tell what website a client connects to
 - + No need for a database
 - What about onion services?

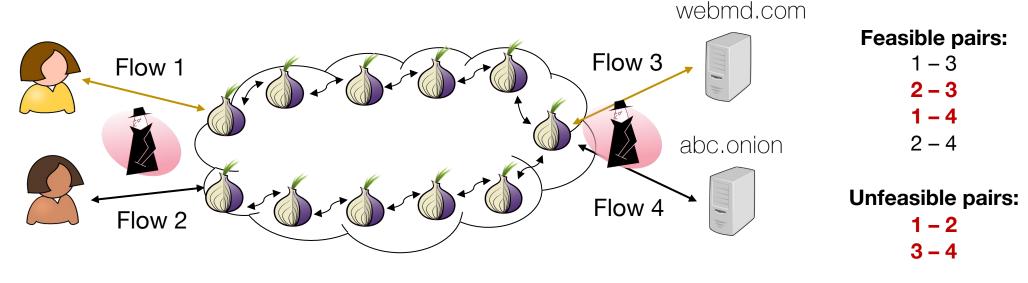
Challenges for onion service flow correlation (I)

- Onion services connect to the Tor network very much like a client
 - A naïve correlation method may try to match flow pairs that are surely uncorrelated



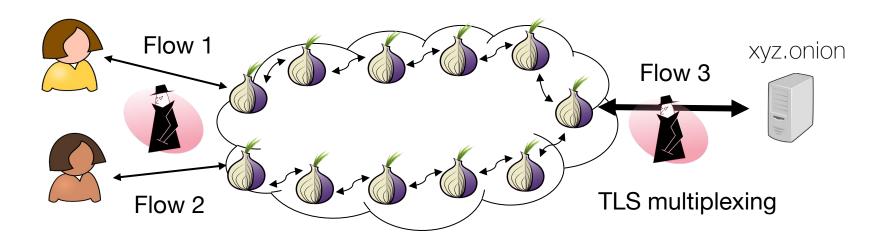
Challenges for onion service flow correlation (II)

- Onion services connect to the Tor network very much like a client
 - A naïve correlation method may try to match flow pairs that are surely uncorrelated
- Identify and discard client requests towards the clearweb



Challenges for onion service flow correlation (III)

- Onion services connect to the Tor network very much like a client
 - A naïve correlation method may try to match flow pairs that are surely uncorrelated
- Identify and discard client requests towards the clearweb
- Untangle concurrent client requests/responses at the onion service guard



Our recent efforts

- Introduce SUMo, a novel classification pipeline that enables efficient and accurate flow correlation for Tor onion service sessions
- Collect a large dataset for evaluating flow correlation on Tor, encompassing accesses both to clearnet and onion service websites
- Provide an implementation and evaluation of SUMo

[shameless advertising spot]

Flow Correlation Attacks on Tor Onion Service Sessions with Sliding Subset Sum











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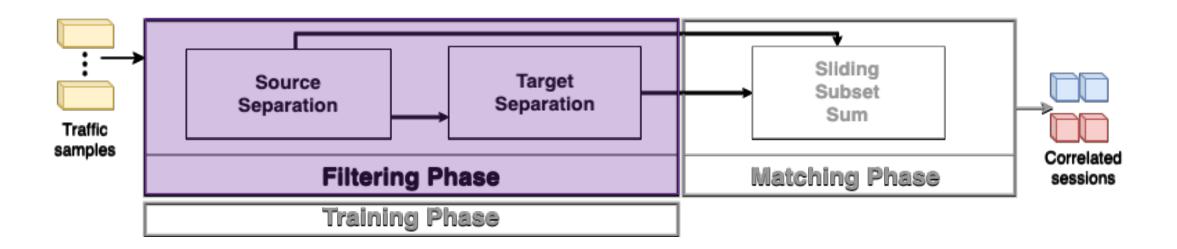


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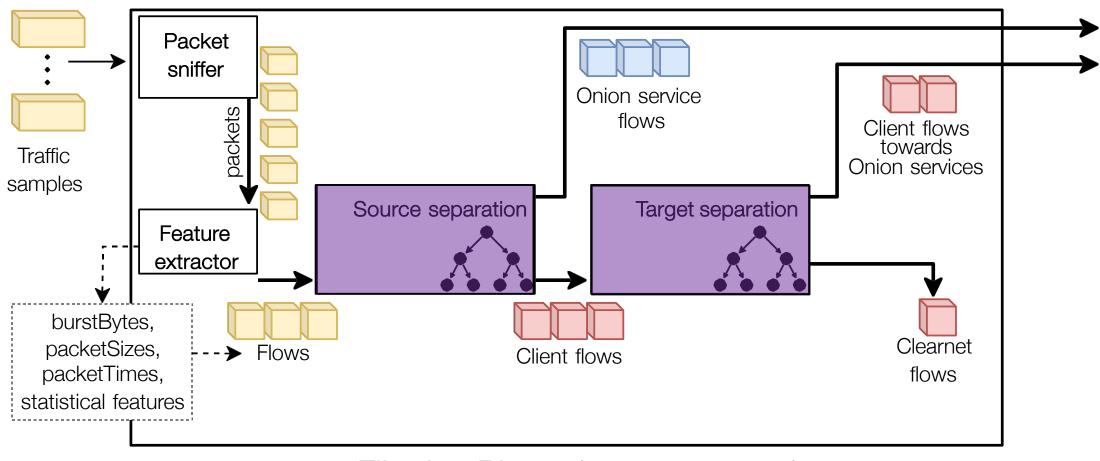
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The SUMo Pipeline

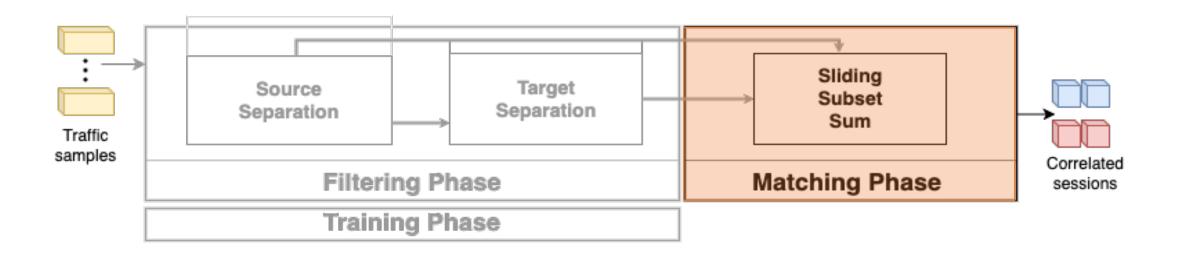


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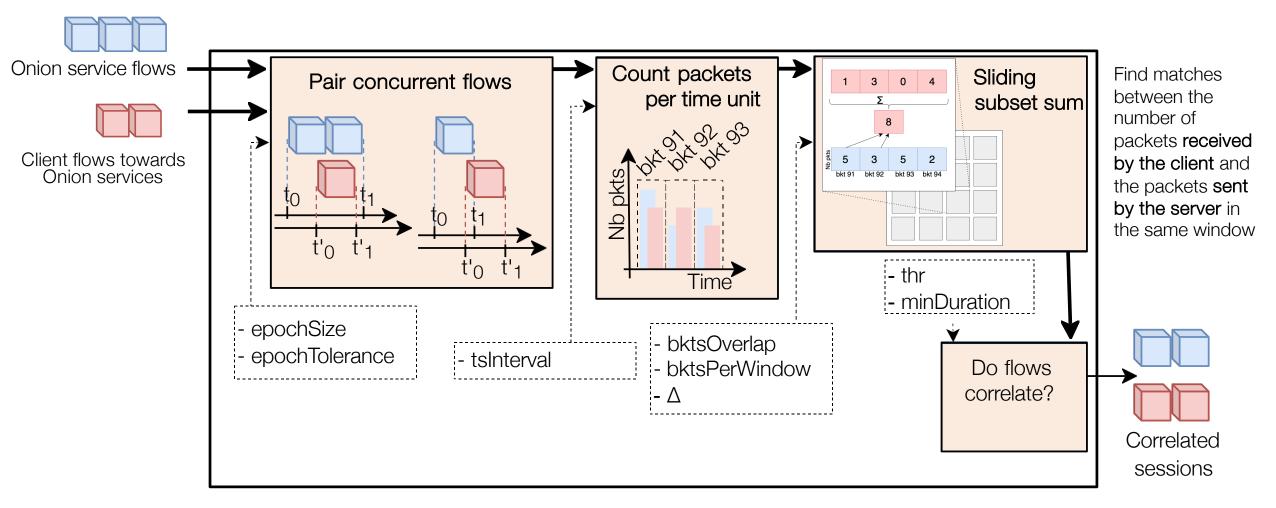


Filtering Phase (on local probes)

The SUMo Pipeline



Match clients with onion services

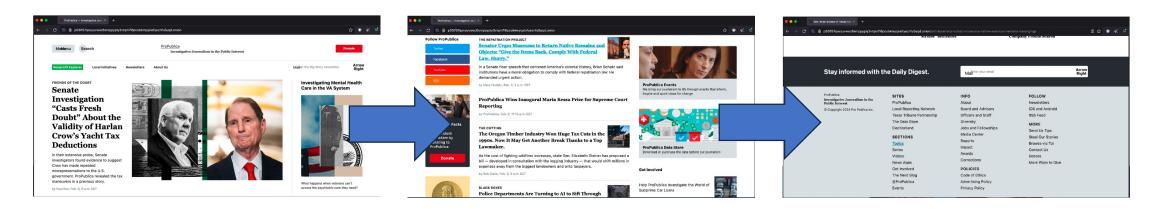


Matching Phase (on correlator)

How did we evaluate SUMo?

Experimental testbed

- Developed a framework to generate network traces (> 80,000 pcaps)
 - Geographical distribution of live nodes (clients and servers)
 - Hosted a set of webpages scraped from actual onion services
 - Modeled client requests' concurrency to .onions and Tranco top 150 sites
 - Emulated typical browsing behaviour to collect browsing sessions



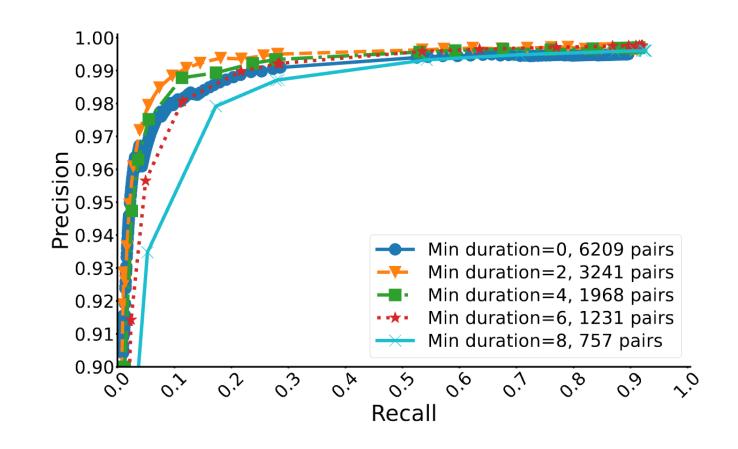
SUMo can effectively correlate flow-pairs

Sessions of any duration

- 99.5% precision
- 89.6% recall

Sessions > 6 minutes

- 99.8% precision
- 92.1% recall



SUMo's pipeline is efficient

Phase	Stage	Training time (full data)	Inference Time (per flow)
Filtering	Source separation	4.3s	44.7ms
	Target separation	1.7s	35.0ms
Matching	Session correlation	-	32.6ms

Trains fast

Filtering classifiers are fast to train

Filters fast

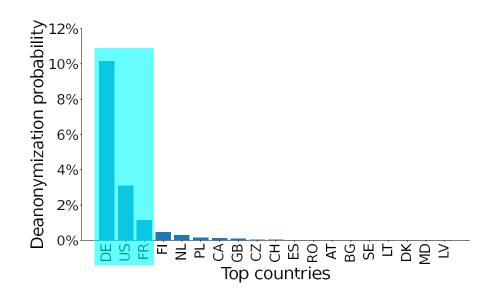
Filtering classifiers are fast to predict

Correlates fast

Sliding subset sum is fast to match (100x faster than SOTA correlation)

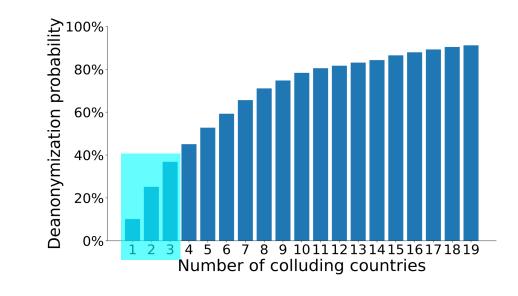
Traffic correlation on Tor is a realistic threat

- We verified client and onion service guard locations
 - for 40,000 random 6-hop circuits between our clients and Oses



Probability of both guard nodes being under the same jurisdiction

Guard node distribution is heavily skewed



Visibility over guard nodes under country collusion

Chances of deanonymization increase

Takeaways

- Onion services are instrumental for anonymity online
- The threat of traffic correlation on Tor is getting increasingly realistic
 - colluding entities can intercept a large fraction of Tor traffic
 - correlation algorithms are increasingly efficient and effective

Countermeasures:

- Increase guard node's geographical diversity
- Pluggable transports (e.g. Brik [CoNEXT'23])
- Client concurrency (e.g., fake client traffic)
- Concurrent "multi-tab" requests



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Thank You!