Let a Thousand Filters Bloom:

privacy-preserving long-term collection of DNS queries

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Introduction

- Privacy of DNS traffic between client and resolver currently has a lot of attention in the Internet community, e.g.:
 - DPRIVE working group in the IETF, standardised DNS-over-TLS
 - Deployment of DNS-over-TLS by e.g. 1.1.1.1, 8.8.8.8, 9.9.9.9 and others (including SURFnet)
 - Upcoming DoH standard (DNS-over-HTTPS), which has a big push from the browser community
- Note the focus is on privacy of traffic in-flight

Elephant in the room

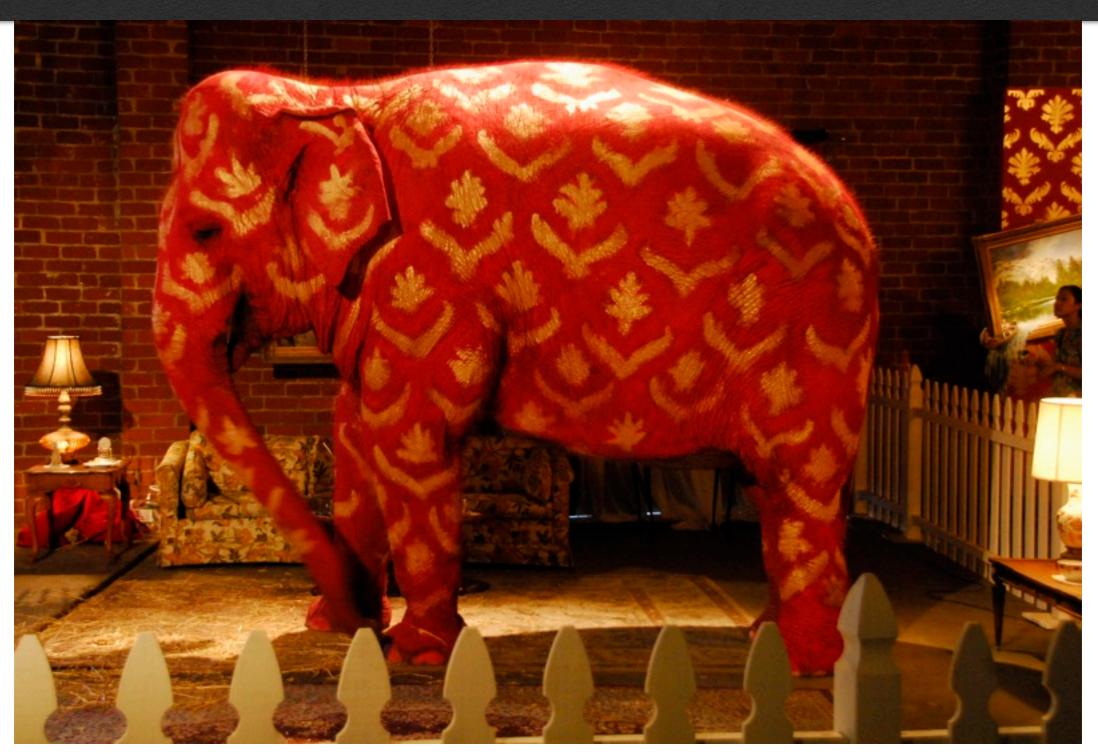


image © Jdcollins13@Flickr

Elephant in the room

- Resolver operators can still observe and collect DNS query traffic
- And they have legitimate reasons to do so
- For example: to detect indicators of compromise in DNS traffic



Goal

- Privacy is a strongly held value at SURFnet
- Yet we also need to ensure the security of our network and the users on it
- Simply logging DNS queries on our resolvers is unacceptable
- We want to take strategic and tactical decisions based on the presence of DNS queries associated with indicators of compromise, so we are not interested in queries per user
- So we asked ourselves:

How can we detect if certain DNS queries were performed, while respecting the privacy of users?

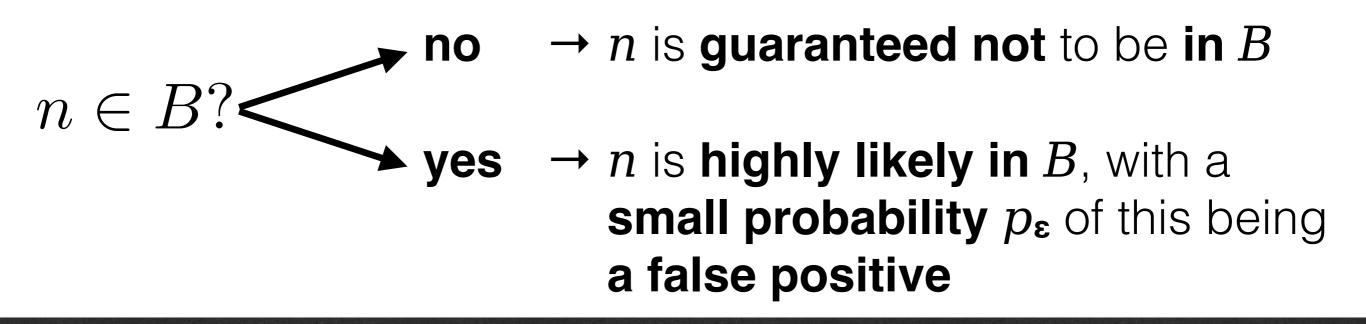


- We worked with Dutch security company
 Quarantainenet to develop a possible solution
- We want to use Bloom filters as a privacyenhancing technology to record all DNS queries
- This talk explains what Bloom filters are, how we intend to use them, and what we have learned so far



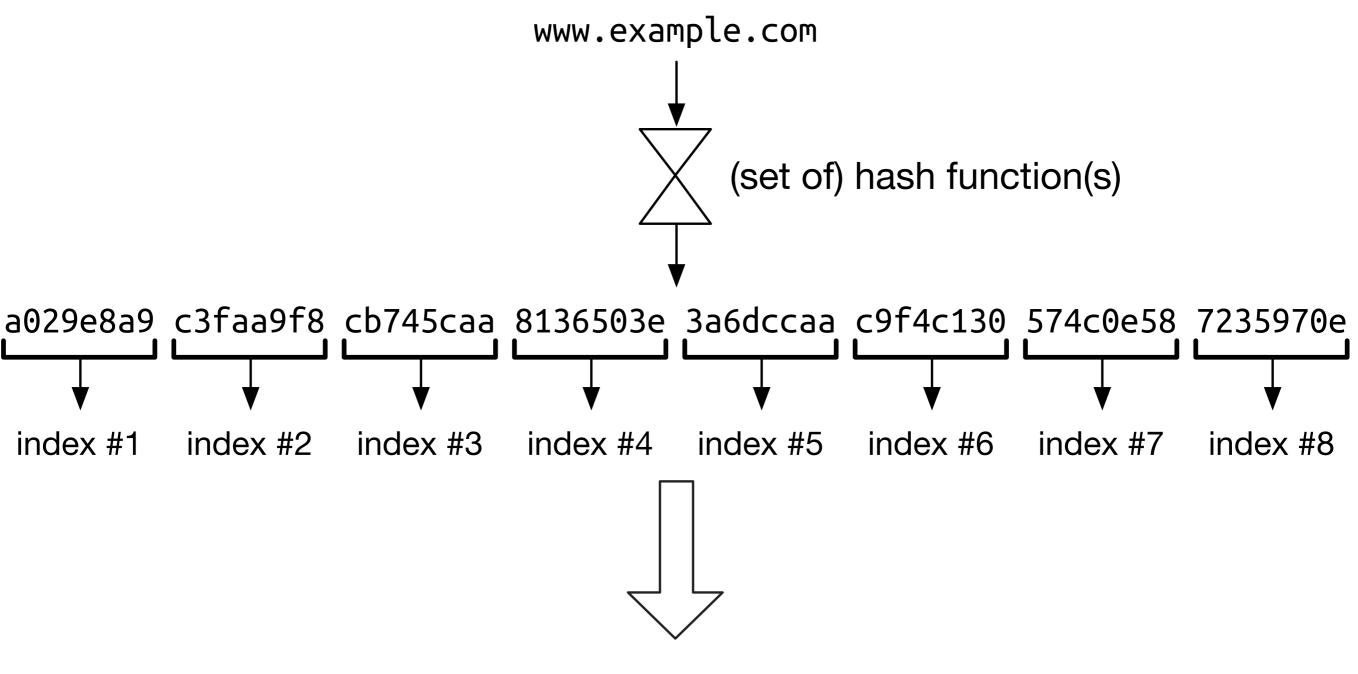
What is a Bloom filter?

- Originally designed in 1970 as a space-efficient way to optimise indexing of data
- Think of Bloom filters as unordered sets of unique elements with probabilistic membership tests
- For a Bloom filter *B* and an element *n*, if we test membership:



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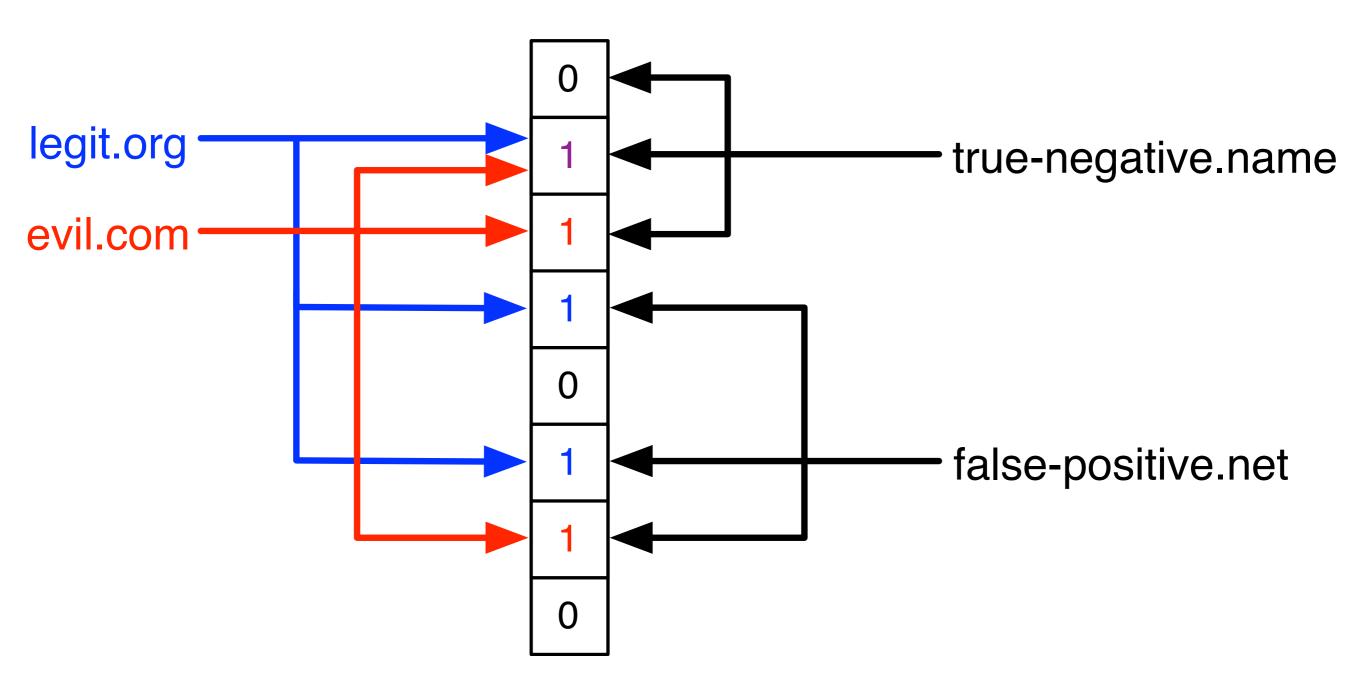
Bloom filter in pictures



set bits to 1 in bit array using indices

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Bloom filter in pictures



(image courtesy of Quarantainenet)

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Bloom filter parameters

- Tune to achieve a certain (low) false positive rate at a reasonable filter size
- Parameters:
 - Number of hash functions $k \rightarrow$ number of indices
 - Size of bit array m
 - Expected number of distinct elements *n*
- The formula below approximates the probability of a false positive p_ε:

$$p_{\epsilon} \approx \left(1 - e^{-\frac{kn}{m}}\right)^k$$

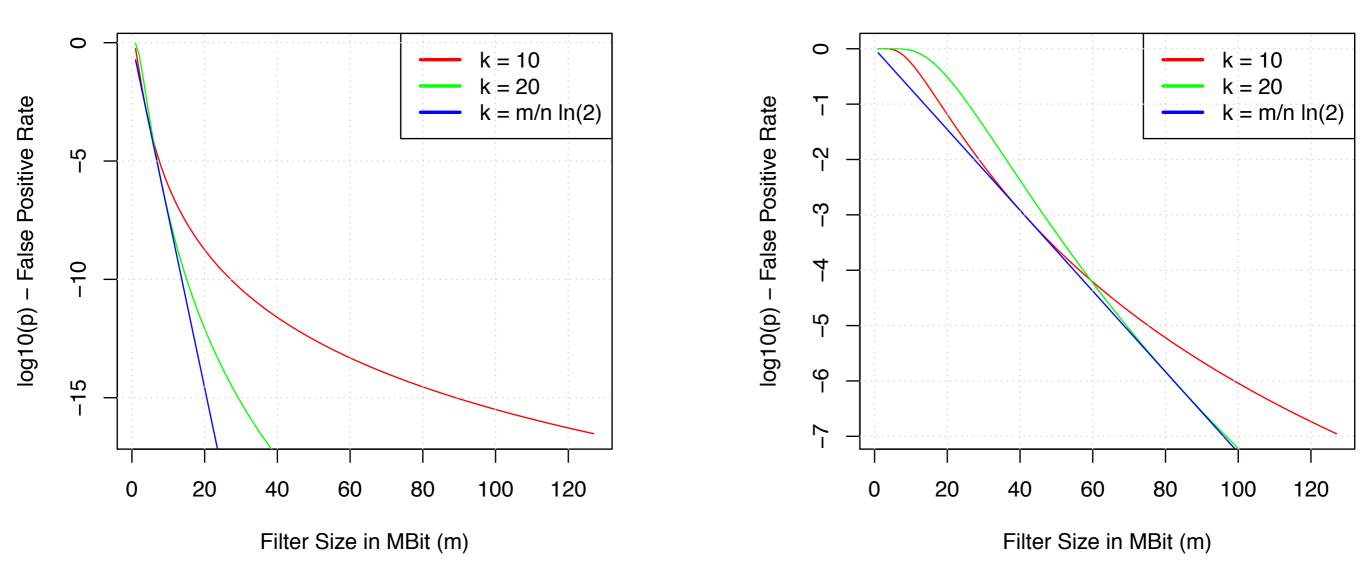


False positive rate

Bloom Filter False Positive Rate: n = 300K

Bloom Filter False Positive Rate: n = 3M

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Privacy properties

- Filters do not store original query names and are non-enumerable; lookup only possible if you know exactly what you are looking for
- By mixing queries from multiple users in a single filter, tracking individual users becomes even hard(er)
- We can combine the state of filters with the same parameters into a new, aggregated filter (with possibly a higher false positive probability, but data over a longer period and/or for more users combined)



Other considerations

- Privacy risk: if I know a query that unambiguously identifies a certain user (e.g. name of personal server), I can still track them, but impossible to correlate with other queries if more than one user in the filter
- Bloom filters have additional benefits:
 - Space efficient (filters have a fixed, reasonable size)
 - **Time efficient** (lookups are fast)

What to store?

- The most important design decision is what information from a query to store
- We considered the following query attributes:
 1. Full query name (canonicalised)
 - 2. Individual labels in a name (e.g. 'www', 'example', 'com')
 - 3. Queried **type**
 - 4. Response data
- For the moment, we are **focussing on 1 and 2**



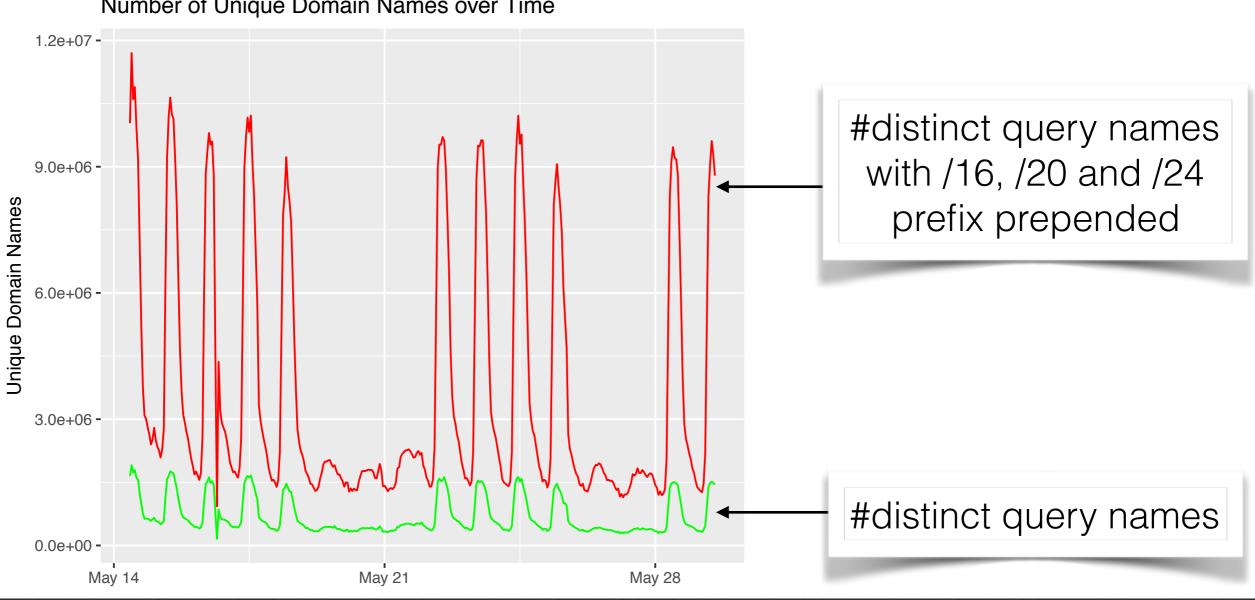
Distributing users over filters

- The second important design decision is how to distribute users over filters
- We want to learn if queries were made by users from certain institutions (again, not interested in individuals)
- Two options:
 - 1. Separate filter for each institution
 - 2. One big filter, and prepend institution name to data inserted into filter



Distributing users -- numbers (1)

Ideally, we want to collect queries per hour; so how many distinct queries do we get?



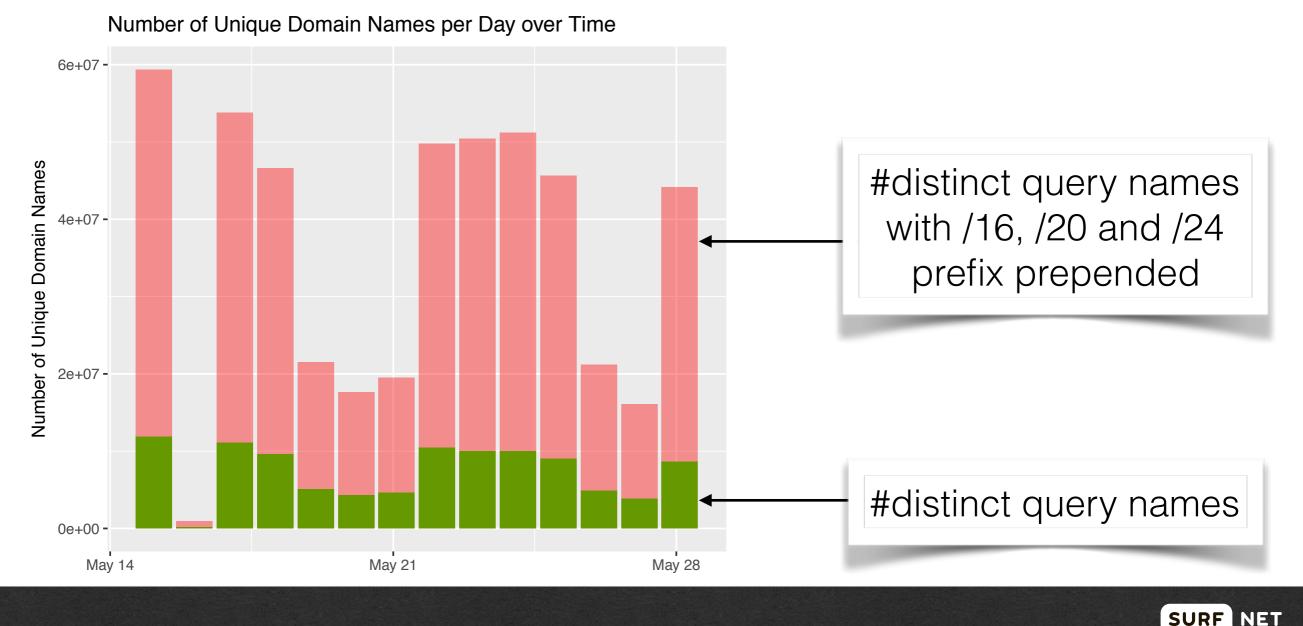
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Number of Unique Domain Names over Time

Distributing users -- numbers (2)

 Also, we would like to aggregate from hourly to daily filters, while maintaining a reasonable false positive rate



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Work in Progress

- We have a master student who is building a working prototype to test the use of Bloom filters for detection of indicators-of-compromise (IoCs) in DNS queries
- His main focus:
 - What IoCs can we detect using this approach, but also: what can't we detect?
 - Designing an architecture for filling and querying filters (e.g. how do we group users, how do we store and query filters?)



Prototype design

- Based on measurements and experiments, we decided to use a single large(r) Bloom filter to store all query information
- Advantages: (very) space efficient, and single set of parameters so filters can be combined for aggregation
- Disadvantage: a single user can pollute the filter with random query names and raise the false positive rate

• We will store:

- <institution>+<full query name>
- <institution>+<individual labels from FQDN>
- <full query name>
- <individual labels from FQDN>
- (possibly also <prefix>+<(parts of) FQDN>)

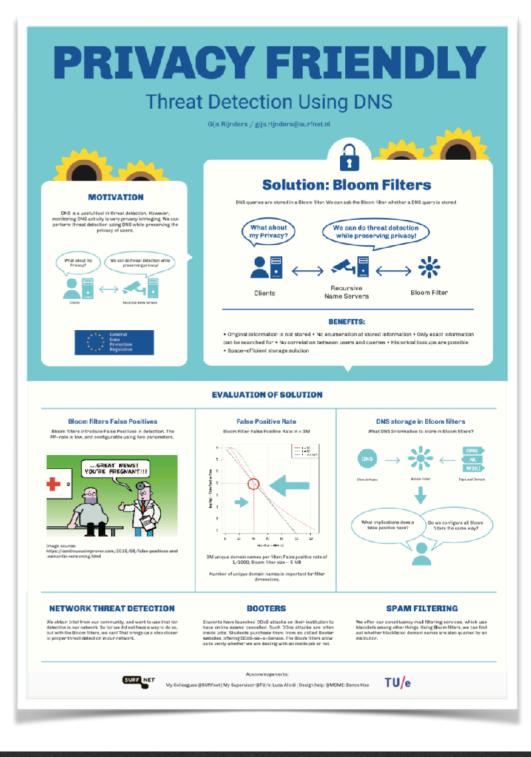


Prototype design

My student, Gijs Rijnders, who is working on our prototype, was one of the winners of the TNC Poster Pursuit, so go see his poster for more info on the prototype!

(also: vote for him ())





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Testing the prototype

- We will deploy Unbound with Bloom filter integration on SURFnet's production resolver infrastructure
- Relatively busy resolvers (order of 5-10k queries per second), that between them see roughly
 150-200k unique client IPs per day
- Ideally, we want to group by customer, challenge: we have ±200 customers
- Goal is also to see how well all of this scales



Use cases

- The master student will look at three use cases in particular:
- Detection of (high value) IoCs that we receive from the Dutch National Detection Network (IoCs received from, a.o., intelligence agencies)
- 2. **Detection of** queries for "DDoS-as-a-Service" providers (aka **Booters/Stressers**)
- 3. Analysis of **blacklist hits from** our **e-mail filtering** service



- Bloom filter library we use developed as open source by Quarantainenet, funded by SURFnet (BSD 3-clause license)
- SURFnet also provided funding for integration in Unbound (will be DNSTAP) in collaboration with NLnet Labs
- Expecting to release prototype code somewhere this year, no definitive date yet (come talk to me if you would like to play with it)



Conclusions

- We set out to find a privacy-conscious way to collect information on DNS queries, with the goal of looking for certain queries for security purposes
- In collaboration with Quarantainenet and NLnet Labs, we are implementing a solution based on Bloom filters, that will be released in open source
- We expect to publish results of our prototype experiments at the end of this summer (late August)



Thank you for your attention! Questions?

acknowledgements: with many thanks to my student, Gijs Rijnders for supplying nice graphs :-)







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