

Large Scale DNS Analysis

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- 1 Motivations
- 2 What data to analyse and how?
- 3 Related work
- 4 DNSSM
- 5 Conclusion

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- ▶ DNS (Domain Name System) is the service that maps a domain name to its associated IP addresses
 $\text{www.example.com} \implies 123.45.6.78$
- ▶ DNS allows to find any information about a domain :
 - ▶ A : IPv4 address
 - ▶ AAAA : IPv6 address
 - ▶ MX : Mail server
 - ▶ NS : Authoritative DNS server
 - ▶ TXT : any information

Why DNS monitoring?

- ▶ DNS: critical and essential Internet service
- ▶ Used by attackers to enhance malicious activities

Misuse

DNS scanning

cache poisoning

typosquatting

fast/double-flux

DNS tunnelling

Malicious activity

worms (spreading)

phishing

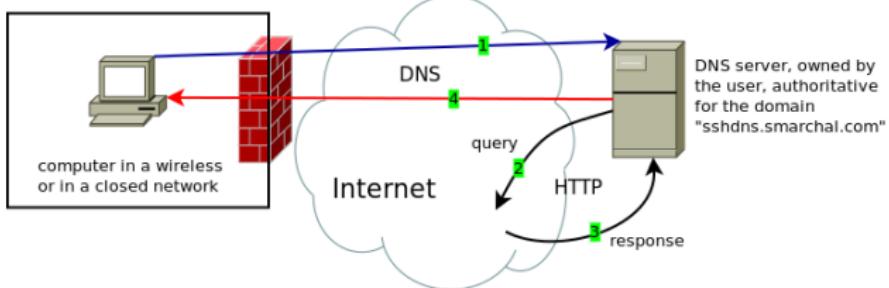
spam

botnet C&C communications

covered channel

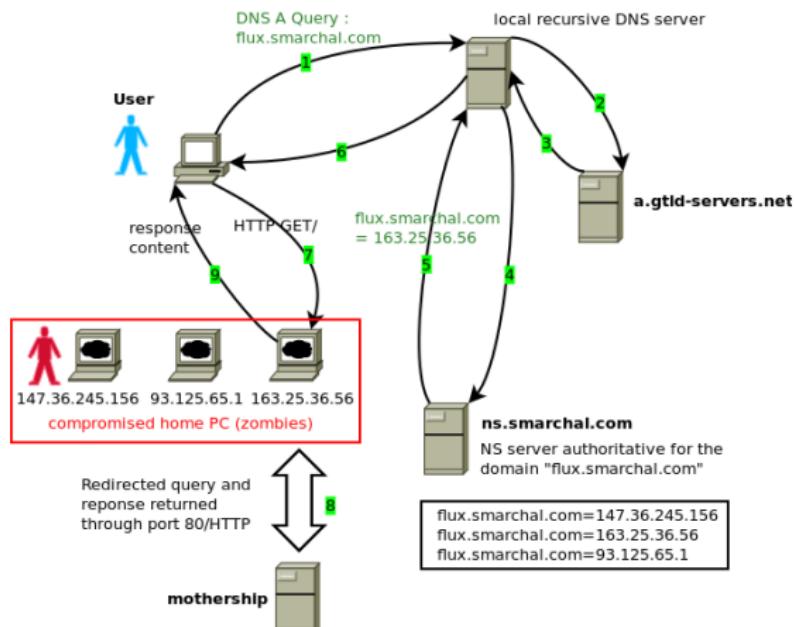
DNS A Query :
 fvwrk4tufv3damk5ygl3ttonuc4y3pnuwgky3eonqs2431mez23tjon2h.amzyggqwwg2lsoqwxmmbribxxazlloonzwqltdn5wsyzldmrzwc1tnbqtelio.nfxzi4bvgiys2y3foj2c25rqgfag64dfnzzxg2bomnxw2ldtonuc2zdt.55153-0.id-3907.up.sshdns.smarchal.com: type A, class IN

DNS TXT Query :
 472-55153.id-3907.down.sshdns.smarchal.com: type TXT, class IN



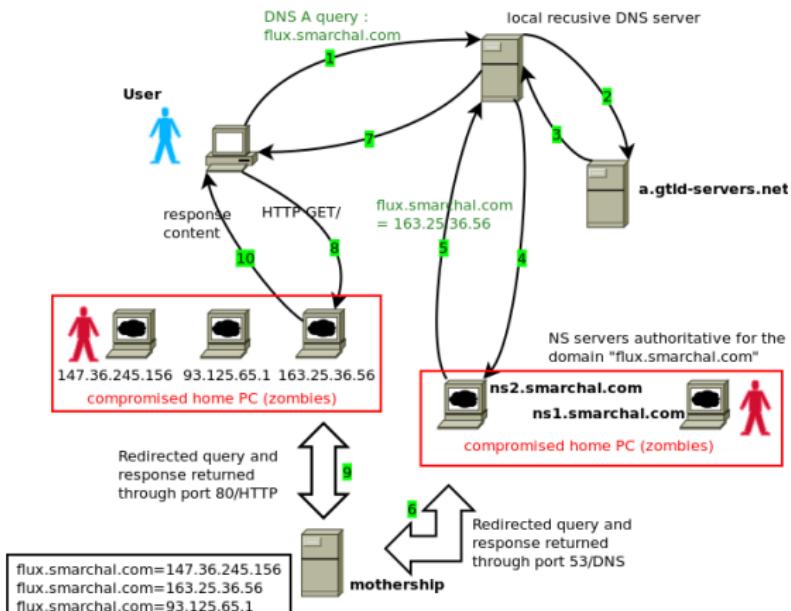
DNS TXT Response :
 472-55153.id-3907.down.sshdns.smarchal.com: type TXT, class IN
 Text: 4P7rSAb6lACp7loEm4+VNWhPRZECBFxtM1ZqpT7C7npsmNDbMrPVzMRtA3V4TsVfyAWOBKhOT0Rk7efFbCxWEYk/xbwGYTKzaEDdACz4daWutOoDAGxHbA7FTnTQ5sB3Z5oAnalnsAn+INxWFBw=
 Text: 4NbAGnqBcd53QBVBVscpU3lioCkS5GNz/xqgdMVLdxGBXSDn4oylnz4IK8XZtyrMjZxh7N4VZTFXA
 KXs61eHQvtbHtgEz11sWMPc=

- ▶ huge number of DNS A queries for the same domain
- ▶ abnormal number of DNS TXT queries and responses over time



- ▶ several A records over time
- ▶ RRs with low TTL

Double Flux



- ▶ several A and NS records over time
- ▶ RRs with low TTL

Identify such behaviour (DNS tunnelling, fluxing domains, DNS scanning, etc.) \Rightarrow detection:

- ▶ **worm** infected hosts
- ▶ malicious **backdoor** communication
- ▶ **botnet** participating hosts
- ▶ **phishing** websites hosting
- ▶ **spamming** domains
- ▶ etc.

\Rightarrow Only based on DNS features

Legitimate activities can have the same characteristics as malicious activities:

⇒ Fluxing domains / Content Delivery Network (CDN):

- ▶ RRs with low TTL
- ▶ several IP addresses for the same domain name
- ▶ IP addresses scattered over several IP ranges
- ▶ algorithmically generated subdomains

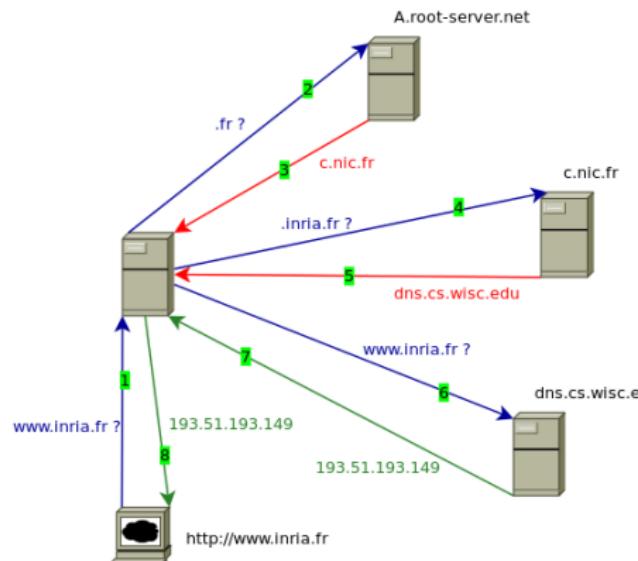
⇒ **Refine features selection to discriminate malicious from legitimate activities**

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- ▶ Extract relevant features:
 - ▶ TTL
 - ▶ Number of Resource Records (A, NS, TXT, etc.)
 - ▶ DNS querying behaviour for a single domain (number of requests)
 - ▶ Number of subdomains (both in requests and responses)
 - ▶ features from lexical analysis of domain name
 - ▶ etc.
- ▶ Information contained in :
 - ▶ domain names
 - ▶ DNS requests (name and behaviour)
 - ▶ fields in DNS responses

- ▶ Gather both DNS requests and responses
- ▶ Where must we place probes ?

- ▶ end user machine (no privacy, redundancy)
- ▶ recursive DNS server (no specific user behaviour)
- ▶ authoritative DNS server (targeted domains)



- ▶ Probe locally or worldwide

- ▶ Large scale analysis ⇒ recursive DNS server
- ▶ Multiple locations all over the world
- ▶ Observations (probes set up in recursive DNS servers of 2 ISPs in Luxembourg)
- ▶ 1 GB of data per day

⇒ **Address the problem of data storage and data processing for scalability**

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- ▶ Apply supervised classification to DNS features \Rightarrow 2 classes: legitimate / malicious
- ▶ Targeted detection of malicious activity (only phishing, botnets, spam, etc.)
- ▶ Use ISC (Internet System Consortium) Passive DNS Database for learning step \Rightarrow only DNS responses : domain names, RR type, @IP, first and last seen
- ▶ Data storage addressed using binary tree

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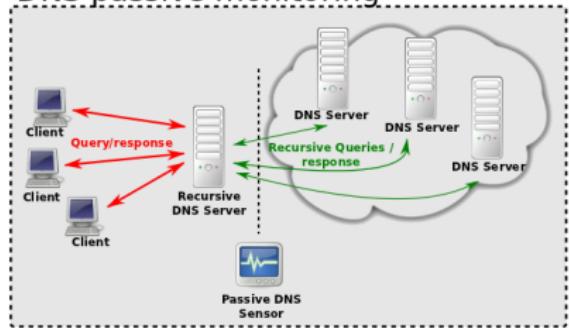
Automated clustering technique for DNS on-line analysis

- ▶ Passively collected data at recursive DNS server level (based on Florian Weimer work)
- ▶ Extraction of 10 relevant features
- ▶ MySQL database storage
- ▶ K-means clustering : 8 clusters
- ▶ Group domains regarding their activity
- ▶ Tested on 2 datasets (\neq location, \neq type of network, \neq users, \neq quantity)

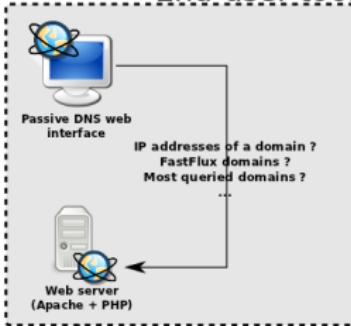
Distributed data storage

- ▶ Meet scalability requirement (10s of Gigabytes of data per month)
- ▶ Centralized MySQL database \implies distributed architecture
- ▶ Hadoop cluster implementing MapReduce design pattern:
 - ▶ distributed data storage
 - ▶ distributed computations

DNS passive monitoring

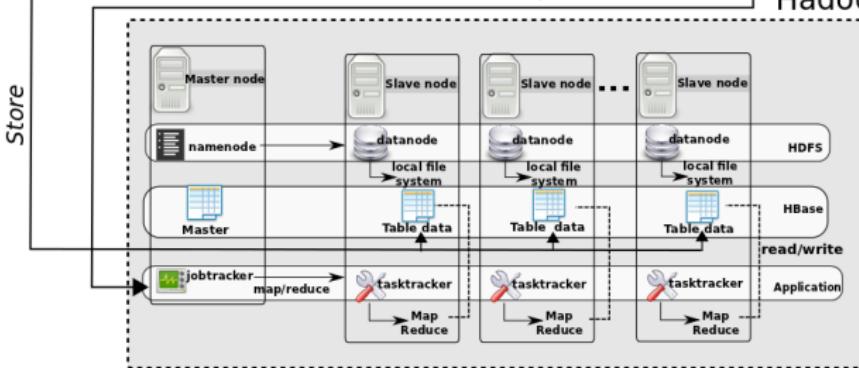


End-user tool



Data access and process

Hadoop



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- ▶ Challenges for large scale DNS analysis
 - ▶ Probing, features selection, data storage, data mining
- ▶ Passive DNS monitoring solution (DNSSM)
 - ▶ Architecture for data collection, storage and mining (leveraging Hadoop)
 - ▶ Relevant features selection
 - ▶ Unsupervised clustering techniques ⇒ domain activity
- ▶ Future Work:
 - ▶ Apply technique to bigger datasets
 - ▶ Selection of new relevant features
 - ▶ Explore lexical and semantic composition of domain names

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