



# **Modern Botnets**

and the Rise of Automatically Generated Domains

Joint work with

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### Who I am

#### Federico Maggi, PhD

Post-doctoral Researcher



### **Topics**

Android malware, malware analysis, web measurements

### **Background**

Intrusion detection, anomaly detection

### www.red-book.eu



#### The RED BOOK

A Roadmap for Systems Security Research

#### Audience

Policy makers

Researchers

**Journalists** 

#### Content

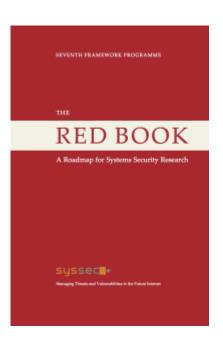
Vulnerabilities

Social Networks

Critical Infrastructure

Mobile Devices

Malware



Free PDF



## Roadmap

- 1. Botnets
- 2. Communication channels
- 3. Domain generation algorithms (DGAs)
- 4. Detecting DGA-based botnets
- 5. Results



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### **Botnets: from malware to service**

#### **Botnet**

- Network of (malware infected) computers
- Controlled by an external entity (e.g., cybercriminal)

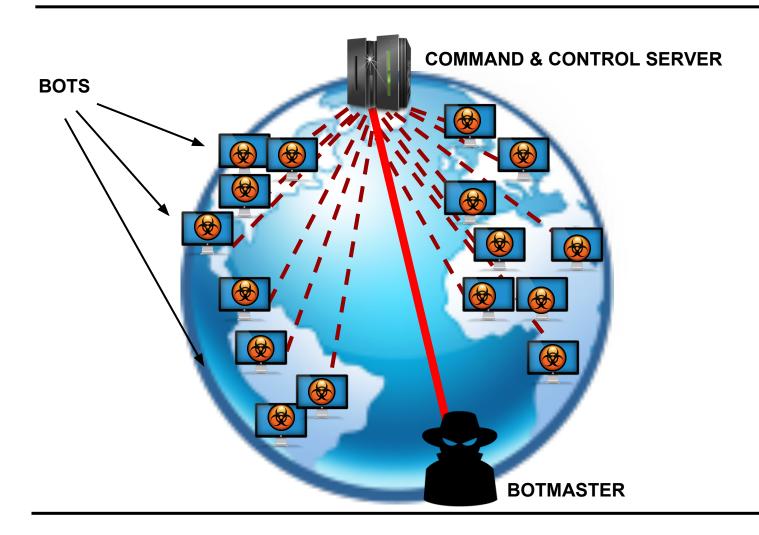
#### **Bot**

- Computer member of a botnet
- Infected with malicious software

#### **Botmaster**

Person or group managing the botnet

# Centralized topology example



### Infected machines = \$\$\$

#### Steal sensitive information

- harvest contacts
- online banking credentials

#### Run malicious activities

- send spam, phishing emails, click fraud
- denial of service

#### Make money

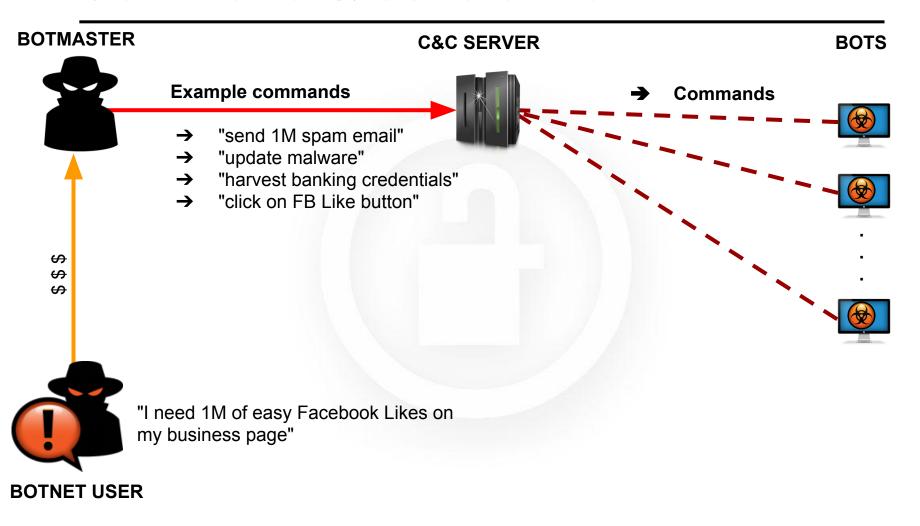
rent the infrastructure as a service

#### **Maintenance**

update the malware

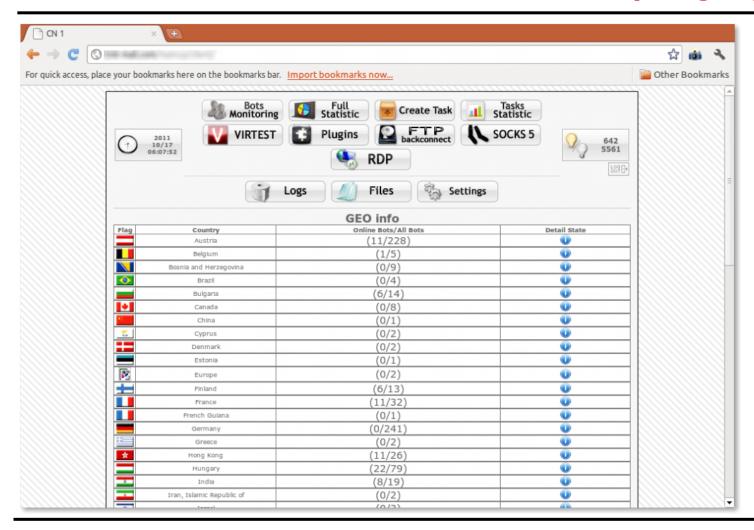


### **Command & control flow**





# Administration dashboard (spyeye)



Source (webroot.com)



# Some notable examples

#### Flashback (2012–today)

- 600K compromised Macs (so, it's not just Windows)
- credentials stealing

#### Grum (2008–2012)

- 840K compromised devices,
- 40bln/mo spam emails

### **TDL-4 (2011–today)**

- 4,5M compromised machines (first 3 months)
- known as "indestructible".

### Cryptolocker (October 2013–today) NEW



## Roadmap

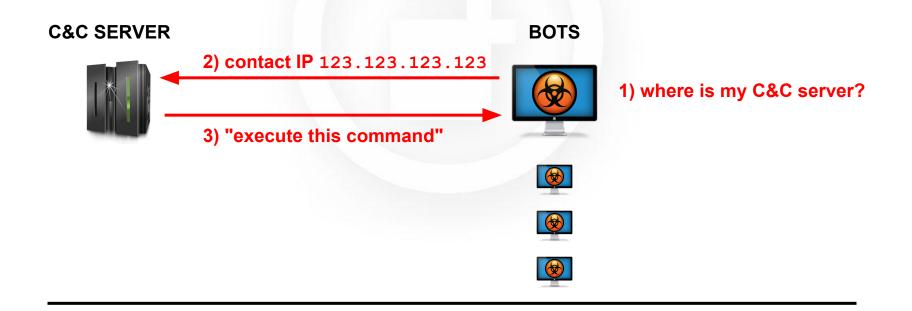
1. Botnets

### 2. Communication channels

- 3. Domain generation algorithms (DGAs)
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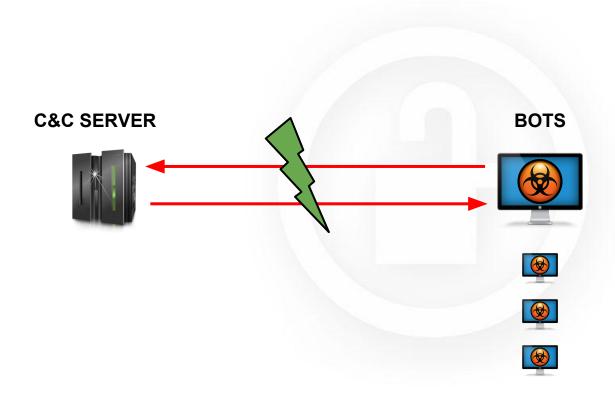
# Where is the my C&C server?

- 1. Where is my C&C server located?
- 2. Contact the C&C server
- 3. Receive command





# **C&C** channel: single point of failure



P2P is the natural answer.

We focus on **centralized botnets** because they're still a **majority**.



### Centralized C&C mechanisms

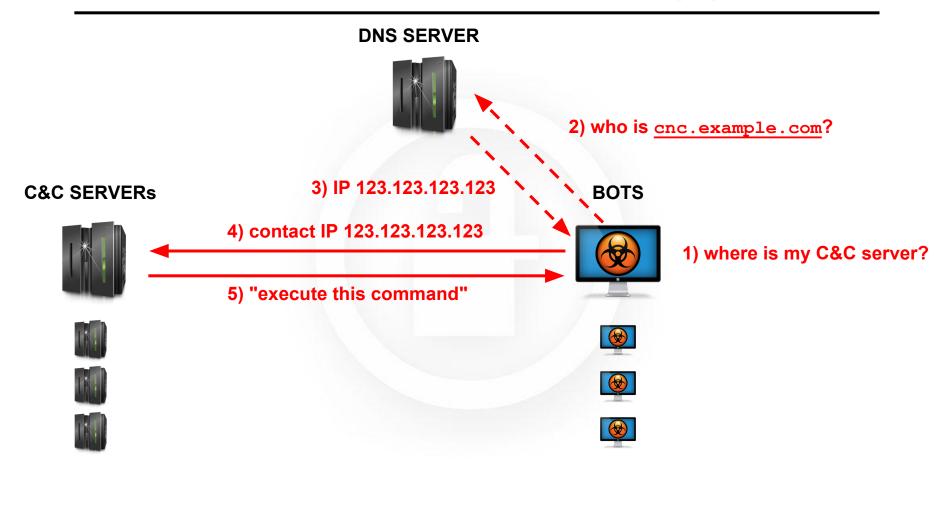
#### Hardcoded IPs (e.g., 123.123.123.123)

- Bot software (malware) ships with the IPs
- Botmaster can update IPs regularly
- Knowing the IP makes takedown easy

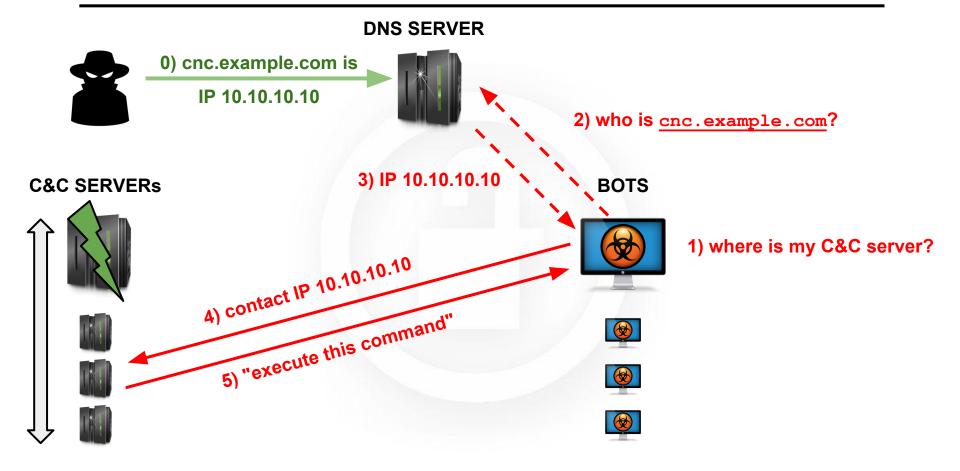
#### Hardcoded domain names (e.g., cnc.example.com)

- Decouple IP from domain
- Botmaster free to change domain names and IPs
- Frequently changing IPs make takedown harder
- Botmaster must own many IPs

# Hardcoded domain names (2)



# Hardcoded domain names (1)



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### Game-changing approach

#### Goals of the botmaster

- Make the C&C server harder to locate
- Make the C&C channel resilient to hijacking

Reversing the malware binary should not reveal the location of the C&C nor any useful information toward that.

## Single domain vs. Domain flux

vljiic.org

yxipat.cn

f0938772fb.co.cc rboed.info

jyzirvf.info

79ec8f57ef.cc

hughfgh142.tk

gkeqr.org

fyivbrl3b0dyf.cn

xtknjczaafo.biz

vitgyyizzz.biz

yxzje.info

nlgie.org

ukujhjg11.tk

aawrqv.biz

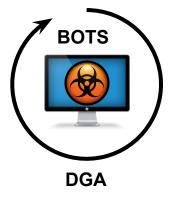


cnc.example.com

predictable easy to leak

#### THOUSANDS OF DOMAINS PER DAY

unpredictable impossible to leak





## Domain of the day

#### **BOTMASTER**



Domain of the day

Register only one domain every day (week) that resolve to the true IP of the C&C

vljiic.org

f0938772fb.co.cc

jyzirvf.info

hughfgh142.tk

fyivbrl3b0dyf.cn

vitgyyizzz.biz

nlgie.org

aawrqv.biz

yxipat.cn

rboed.info

79ec8f57ef.cc

gkeqr.org

xtknjczaafo.biz

yxzje.info

ukujhjg11.tk

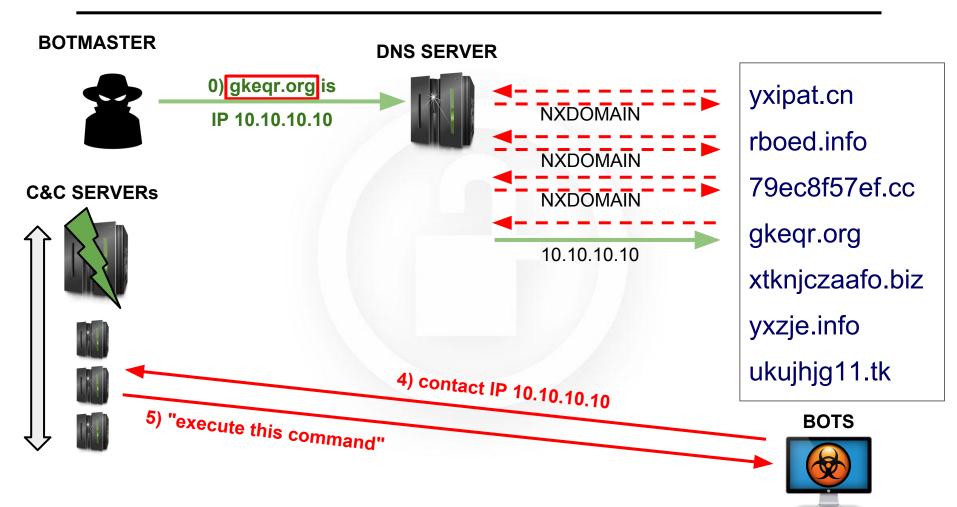
. . .

#### THOUSANDS OF DOMAINS PER DAY

unpredictable impossible to leak



# Where is my C&C server?



# **Leveraging DNS**

- Only the botmaster knows the active domain
- The DNS protocol does the rest
- The DGA can be made more unpredictable (e.g., Twitter trending topic)

Reversing the malware binary

only reveals the generation algorithm

not the active domain of the day!



# Message in a bottle



(Source)

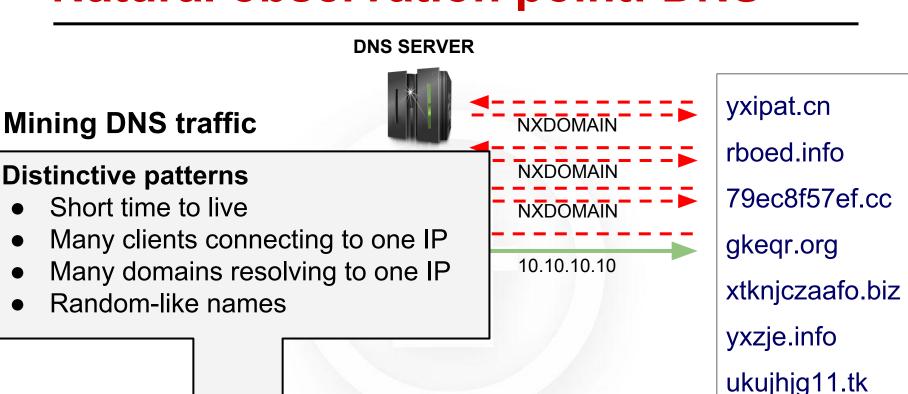


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# Natural observation point: DNS



gkeqr.org is malicious

ukujhjg11.tk

**BOTS** 





## **Domain reputation systems**

#### **Notos**

• [Antonakakis et al., 2010]

#### **KOPIS**

• [Antonakakis et al., 2011]

#### **EXPOSURE**

- [Bilge et al., 2011]
- http://exposure.iseclab.org

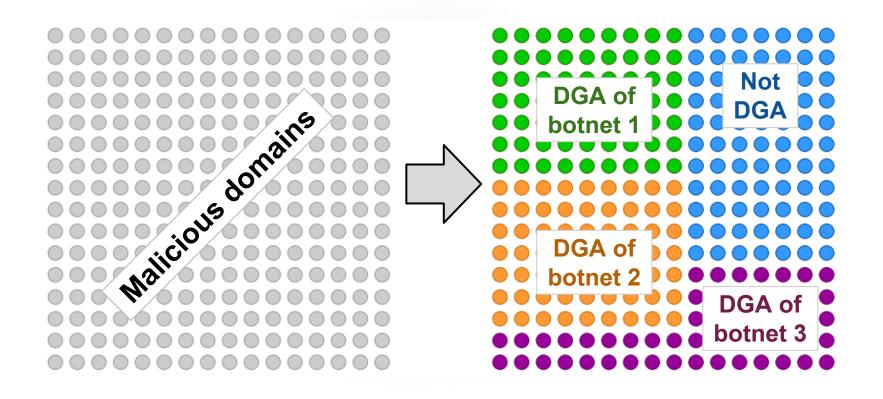
### **Drawbacks**

### They tell malicious vs. benign domains apart

#### No insights on what is the purpose of the domain

- C&C of what botnet?
- Could the same C&C be used for multiple botnets?
- Is the domain malicious for other reasons?
  - Phishing
  - Spam
  - Drive-by download

# More insights needed

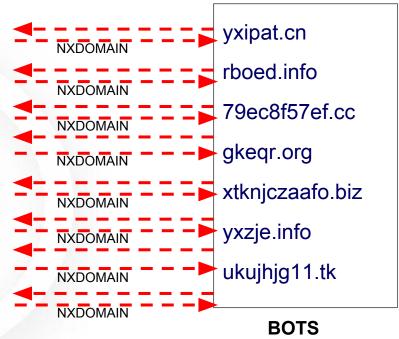


### **NXDOMAINs**

**Infected clients try many domains** 

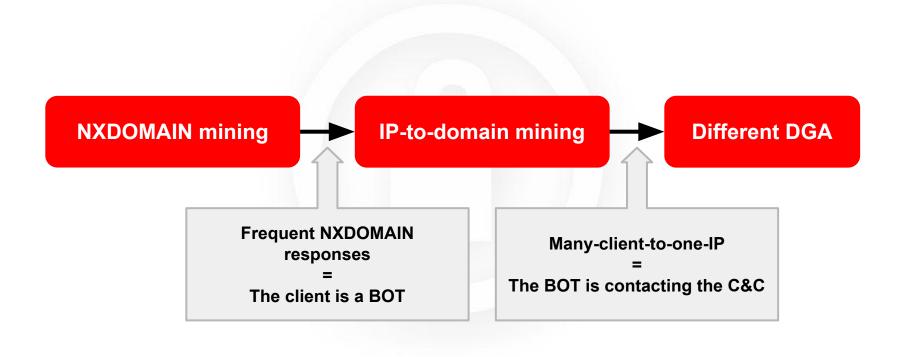
Many NXDOMAIN responses

**Distinctive pattern of DGA** 





# Finding distinct DGAs



### **Drawbacks**

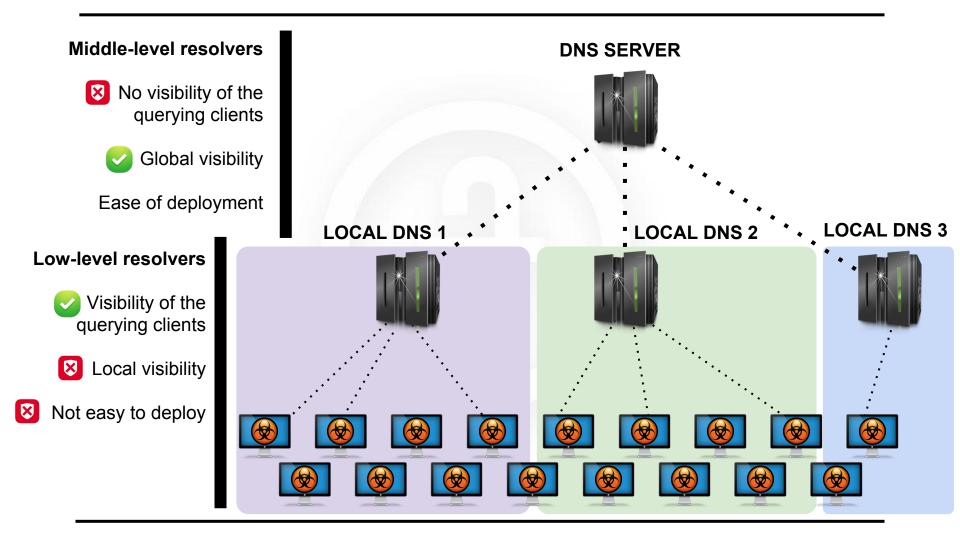
#### Needs an unpractical observation point

- No global view
- Hard to deploy

#### **Needs the IP of the clients**

Privacy of the clients is not enforced

### Lower level DNS servers

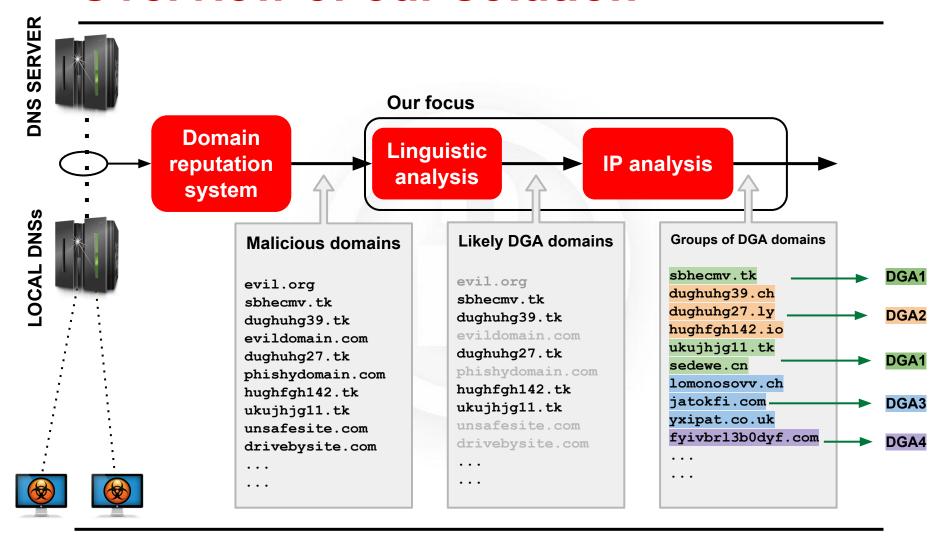




# **OUR SOLUTION**



### Overview of our solution





# Step 1: Linguistic analysis

We measure the "randomness" of the strings with respect to non-DGA-generated domains

malicious.cn fyivbrl3b0dyf.cn yxipat.cn

f0938772fb.co.cc evildomain.com evilrot.org

jyzirvf.info nlgie.org gkeqr.org

hughfgh142.tk aawrqv.biz xtknjczaafo.biz

Feature 1: meaningful word ratio
Feature 2: n-gram popularity

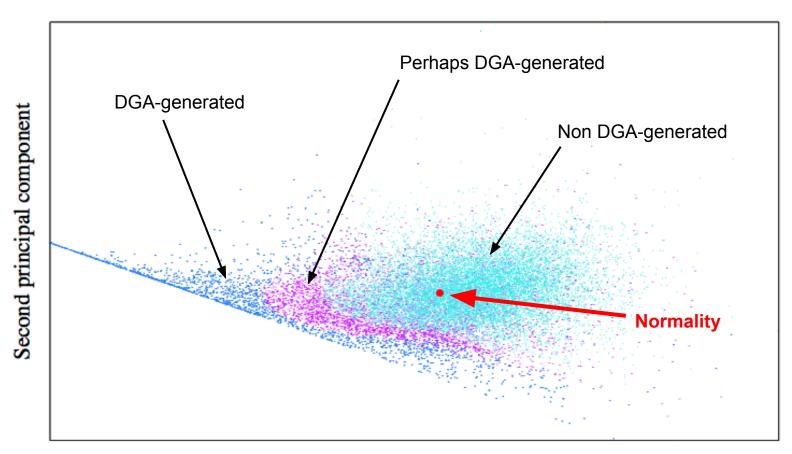
(with respect to a given language)

Likely non-DGA-generated

Likely DGA-generated

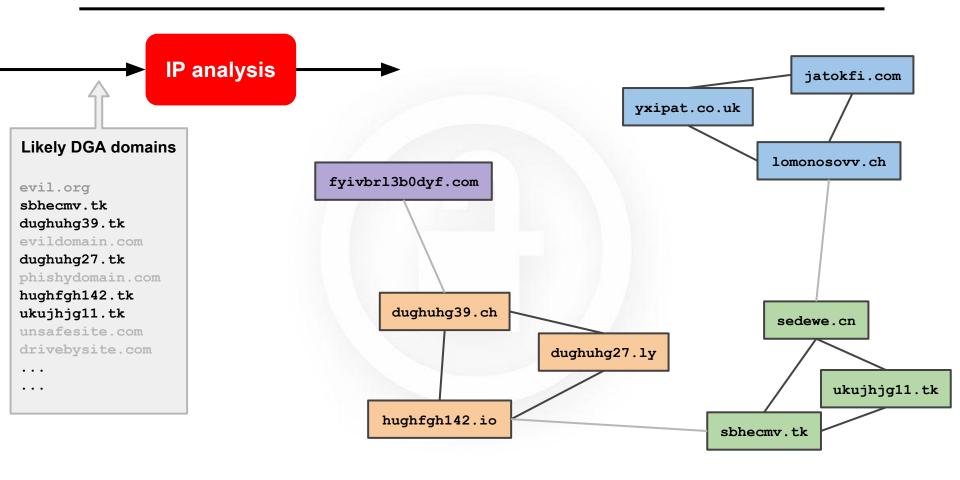
jyzirvf.info nlgie.org gkeqr.org hughfgh142.tk xtknjczaafo.biz aawrqv.biz Feature 1: meaningful word ratio Feature 2: n-gram popularity (with respect to a given language) **Likely DGA-generated** Likely non-DGA-generated Feature 1  $\begin{array}{ll} \textbf{HIGH} & 1 = \frac{4+6}{10} = \frac{|\text{`evil'}| + |\text{`domain'}|}{|\text{`evildomain'}|} = LF1 = \frac{|word_1| + \cdots + |word_N|}{|\text{domainname}|} = \frac{|\text{`pat'}|}{|\text{`vxipat'}|} = \frac{3}{6} = 0.5 \ \text{Lower lower low$ Feature 2 (n = 2) $= \texttt{`ev'} + \texttt{`vi'} + \cdots + \texttt{`ai'} + \texttt{`in'} = LF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = \sum \text{popularity}(\text{n-gram}_i) = \texttt{`yx'} + \cdots + \texttt{`at'} = CF2 = CF2$ **HIGH** HIGH LOW Feature 3 (n = 3)HIGH Feature N (n = N)LOW

# Linguistic features (2D PCA)



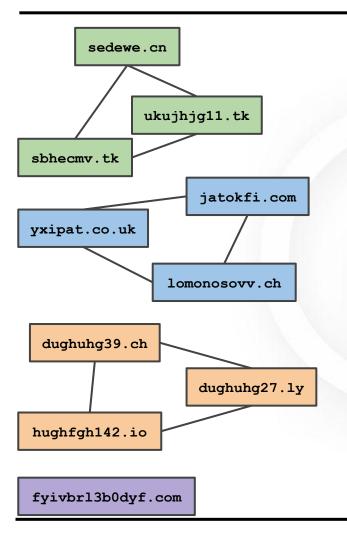
First principal component

## **Step 2: IP analysis**





# **Step 2: DBSCAN Clustering**



#### Cluster 1

Domains that, in their lifetime, have resolved to the very same IPs

#### Cluster 2

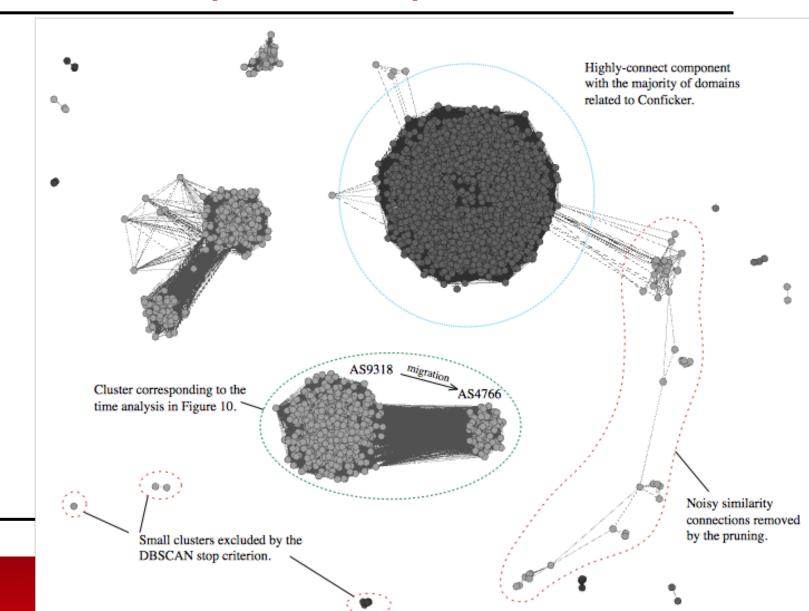
Domains that, in their lifetime, have resolved to the very same IPs

#### Cluster 3

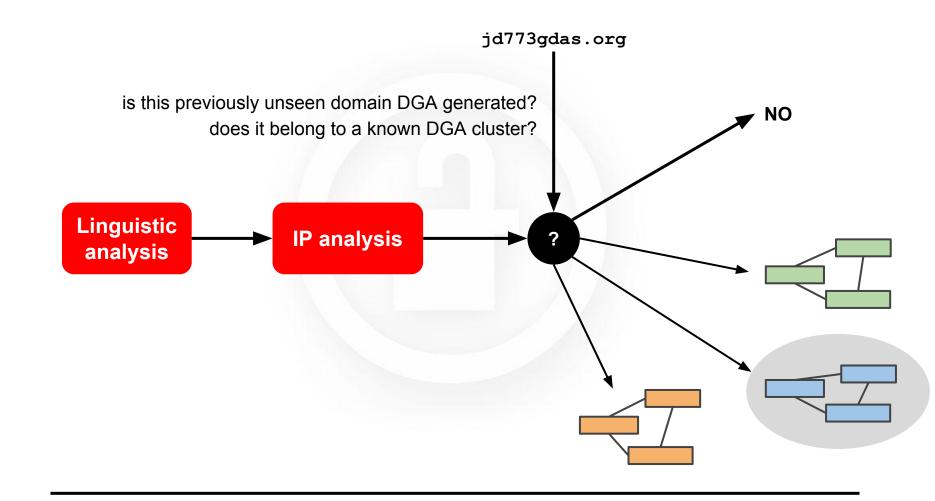
Domains that, in their lifetime, have resolved to the very same IPs

Singleton (removed)

# Real output (example)



# **Classifying new domains**



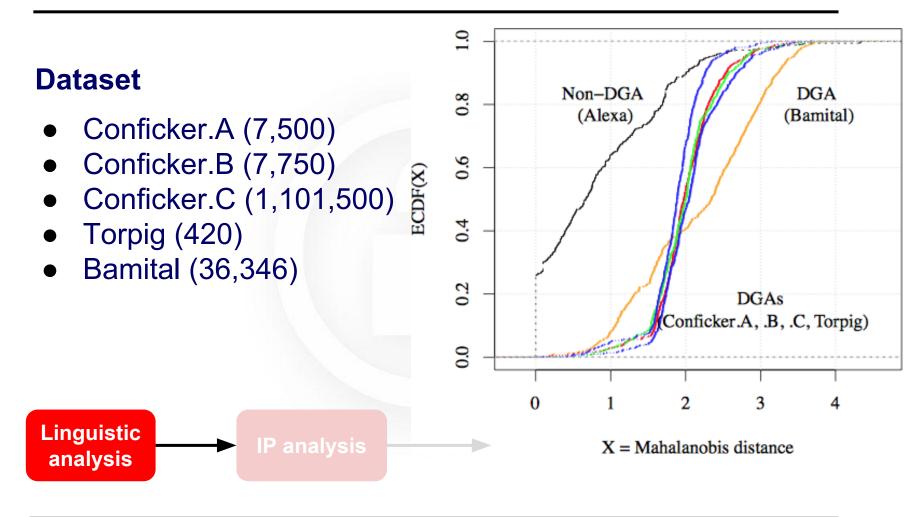
## Roadmap

- 1. Modern cybercrime
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### 6. Results



## Step 1 on real data





## Step 2 on real data

hy613.cn 73it.cn 5ybdiv.cn dky.com ejm.com eko.com 69wan.cn hy093.cn 08hhwl.cn efu.com elq.com bqs.com hy673.cn onkx.cn xmsyt.cn bec.com dpl.com eqy.com watdj.cn dhjy6.cn dur.com CCZ.COM pjrn3.cn 3dcyp.cn x0v7r.cn dky.com ejm.com eko.com 0bc3p.cn hdnx0.cn 9q0kv.cn efu.com elg.com bgs.com 5vm53.cn 7ydzr.cn fyj25.cn bec.com dpl.com eqy.com qwr7.cn xq4ac.cn ygb55.cn dur.com bnq.com ccz.com

Correct clusters found: Conficker, Bamital, SpyEye, Palevo



# **DEMO** (come talk to me offline)

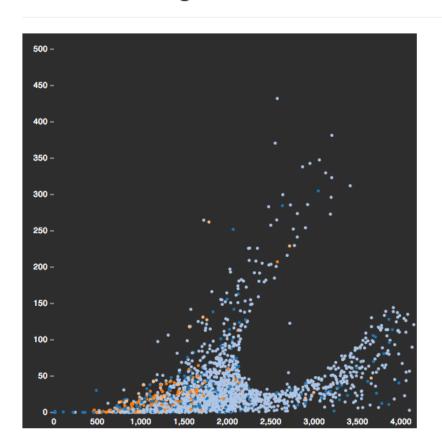








#### **DGA Clustering**









## Ongoing research

### Non-english baseline

- Italian domain names? Swedish domain names?
- Non-ASCII domains?
  - п.соm
  - · 葉留心.io
  - **♥★**₹♥.tk

#### **Word-based DGAs**

- concatenate random, valid words instead of letters
  - also-is-dom-yesterday-a-new.com







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