

Dreamlab Technologies

Taking a look at the *.ch zone with a DGA detector



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Shout out

- Jeroen van Meeuwen (Kolabnow.com)
- Sinan Sekerci (Dreamlab.net)



> whoami

- Mischa Obrecht, 34
- IT-Security since 10 years, analyst, engineer, project / security manager, tester
- Pentester and project manager for Dreamlab
- Involved in the SAPPAN-project as project coordinator for Dreamlab

To get in touch please use:

- mischa.obrecht@dreamlab.net



Agenda

[-] Introduction

- *.ch Zonefile
- Domain Generation Algorithms
- NN DGA classifiers

[+] Methodology

[+] Results



Introduction

switch.ch publishes *.ch and *.li zonefiles

The screenshot shows the official website for SWITCH. At the top left is the SWITCH logo. Below it is a button labeled "Open Data". The main heading "Freely available data" is in a large, bold, dark blue font. Below the heading, a subtext states: "SWITCH publishes open data and makes it available to the public on this website in order to create added value and improve transparency, innovation and efficiency." To the right of the text is a blurred background image of a computer monitor displaying a dashboard. At the bottom of the page, there are three callout boxes: "Open data on domain names", "Contact SWITCH-CERT", and "Top 1000 Blog".

Freely available data

SWITCH publishes open data and makes it available to the public on this website in order to create added value and improve transparency, innovation and efficiency.

Contact SWITCH-CERT

cert@switch.ch

Top 1000 Blog

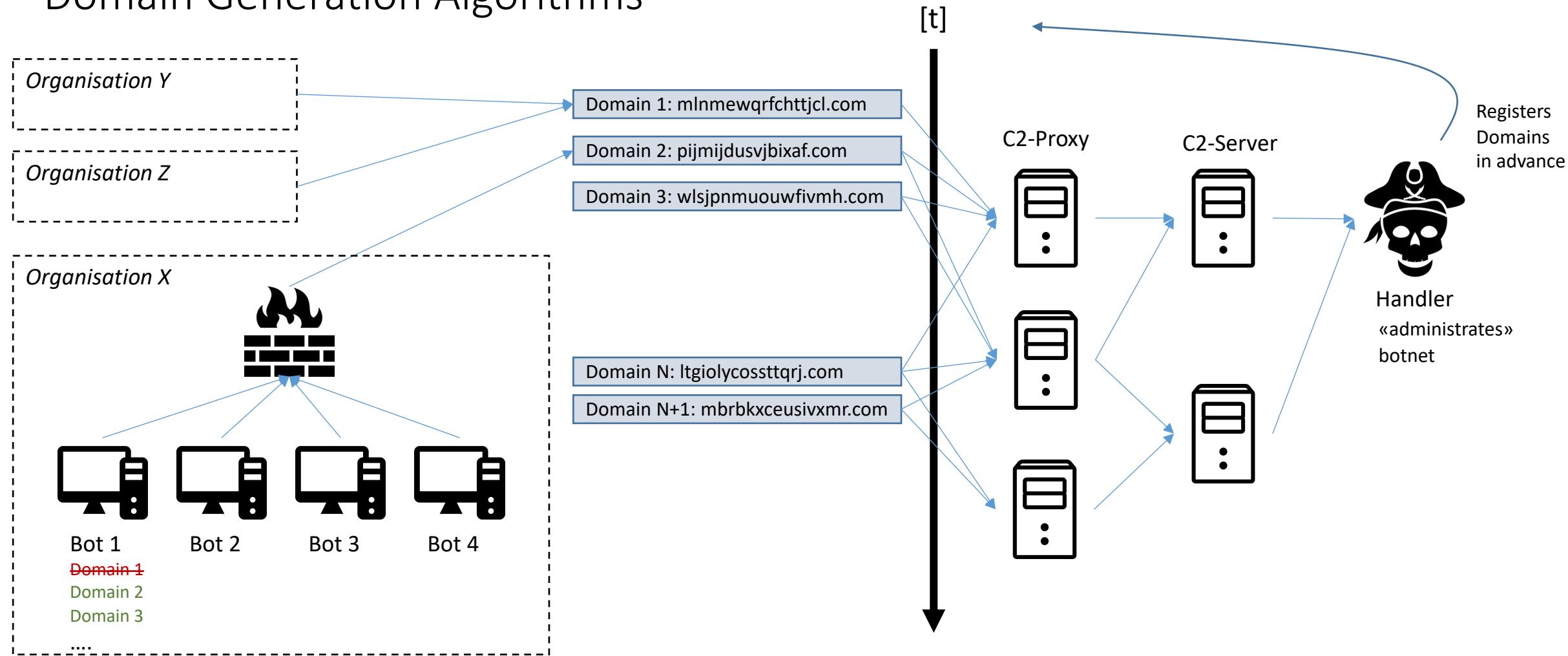
This short blog article describes the technical background of the Top 1000 ranking in more detail.

2260606
Entries for .ch



Introduction

Domain Generation Algorithms





Introduction

(n-ary) NN based DGA detectors are investigated in context of SAPPAN-

→ Bin Yu, Jie Pan, Jiaming Hu, Anderson Nascimento, and Martine De Cock.
2018. Character Level Based Detection of DGA Domain Names. In International Joint Conference on Neural Networks. IEEE,
<https://faculty.washington.edu/mdecock/papers/byu2018a.pdf>

Analyzing the Real-World Applicability of DGA Classifiers

Arthur Drichel
RWTH Aachen University
drichel@itsec.rwth-aachen.de

Samuel Schüppen
Siemens CERT
samuel.schueppen@siemens.com

Ulrike Meyer
RWTH Aachen University
meyer@itsec.rwth-aachen.de

Dominik Teubert
Siemens CERT
dominik.teubert@siemens.com

ABSTRACT

Separating benign domains from domains generated by DGAs with the help of a binary classifier is a well-studied problem for which promising performance results have been published. The corresponding multiclass task of determining the exact DGA that generated a domain enabling targeted remediation measures is less well studied. Selecting the most promising classifier for these tasks in practice raises a number of questions that have not been addressed in prior work so far. These include the questions on which traffic to train in which network and when, just as well as how to assess ro-

1 INTRODUCTION

Bots need to be able to establish a connection to a command and control (C2) server in order to obtain updates. To achieve this end, they often rely on domain generators, which generate a large amount of pseudo-random domains using a seed. The botnet herder knows the algorithm used to predict the algorithmically generated domain names and to register a small subset of these domains with a legitimate domain name generator (DGA) one-by-one trying to obtain a valid response from the C2 server. The majority of these queries result in a failure.

Table 2: Binary Classification: Mixed DGAs

Classifier	ACC	TPR	TNR	FNR	FPR
FANCI	0.99764	0.99744	0.99784	0.00256	0.00216
B-Endgame	0.99891	0.99969	0.99813	0.00031	0.00187
B-NYU	0.99907	0.99976	0.99838	0.00024	0.00162
B-ResNet	0.99916	0.99978	0.99853	0.00022	0.00147



Introduction

Goal: Find malicious/untrustworthy domains in .ch-zone

```
$head malicious.txt  
nymaim chvkhnfqisg.net  
nymaim fnqqhplfm.com  
nymaim mwjvbwj.info  
nymaim ridfga.com  
nymaim xzzwrbnhd.com  
nymaim znrijgzfg.com  
nymaim guqthb.com
```

Assumption: DGA classifiers introduced above can be used to detect malicious / Untrustworthy domains

Available data:

- Available datasets for DGA samples mostly feature *.com, *.org, *.biz, “.info, etc.
- There is no list of DGA samples for .ch-TLD, we can:
 - Create our own
 - Try to generalize (or rather specialize) from existing lists

➔ What will we find applying classifiers that were trained on “generic” data to the *.ch zone?

➔ Can we find a self-contained way to specialize from existing lists and add some locality to the classifier?



Agenda

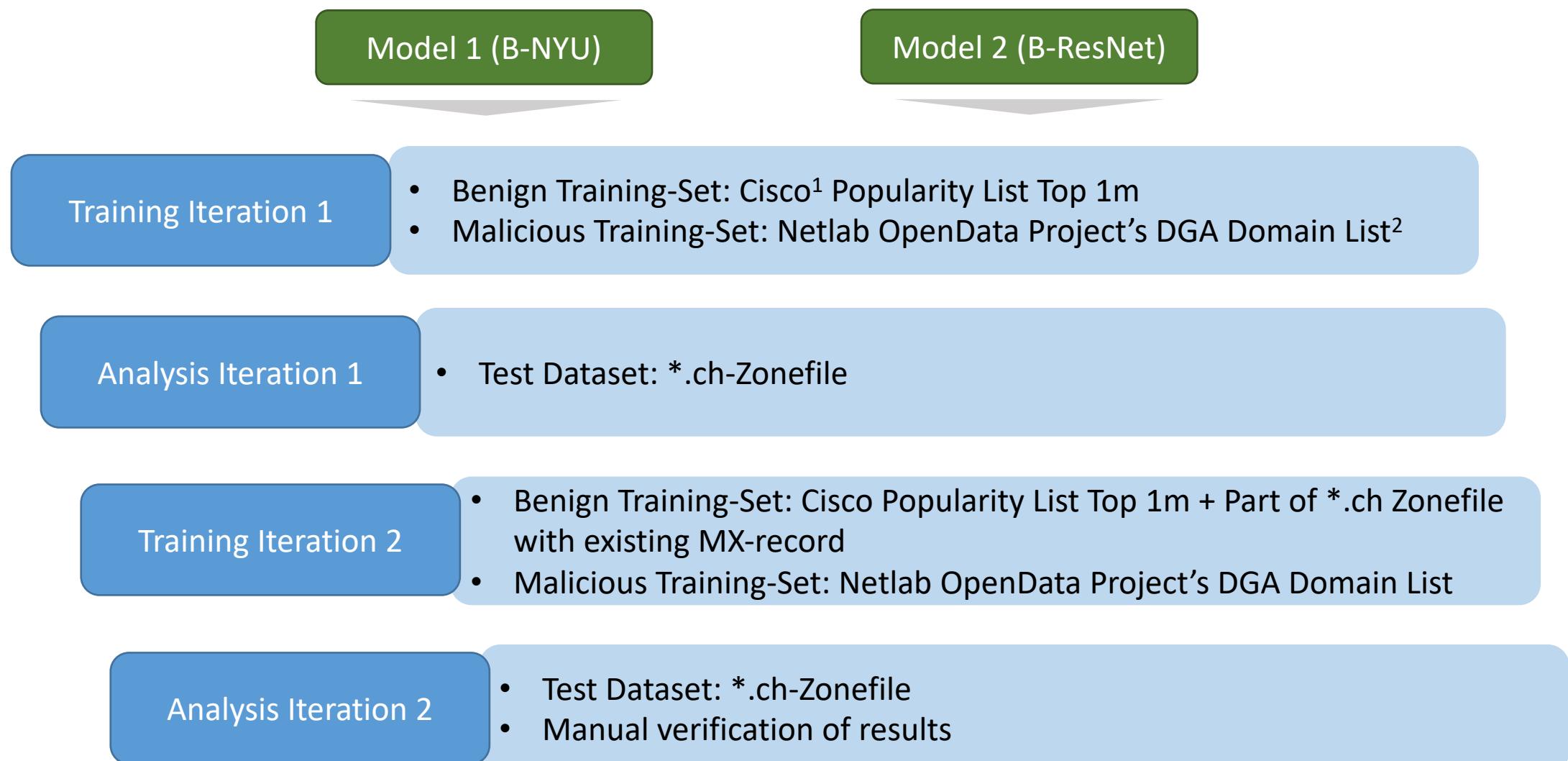
[+] Introduction

[+] Methodology

[+] Results



Methodology



¹Cisco Umbrella Project - <https://s3-us-west-1.amazonaws.com/umbrella-static/index.html>

²<https://data.netlab.360.com/dga>



Methodology

Metrics on Datasets

Dataset	\$ wc -l
Cisco Popularity List Top 1m	999'999
Netlab DGA Domain List	1'369'253
*.ch Zonefile	2'260'606
*.ch-Domains with MX-records	1'640'145

```
$ grep ".ch" cisco-benign.txt | wc -l  
51567
```

```
$ grep ".ch" ch.zone.txt | wc -l  
2260606
```



Agenda

[+] Introduction

[+] Methodology

[-] Results

- Analysis Iteration 1
- Analysis Iteration 2
- Conclusion / Outlook



Results

Analysis Iteration 1 – Distributions of results

B-NYU Distribution:

(0, 25) 2257416

(25, 50) 1347

(50, 75) 675

(75, 100) 1024

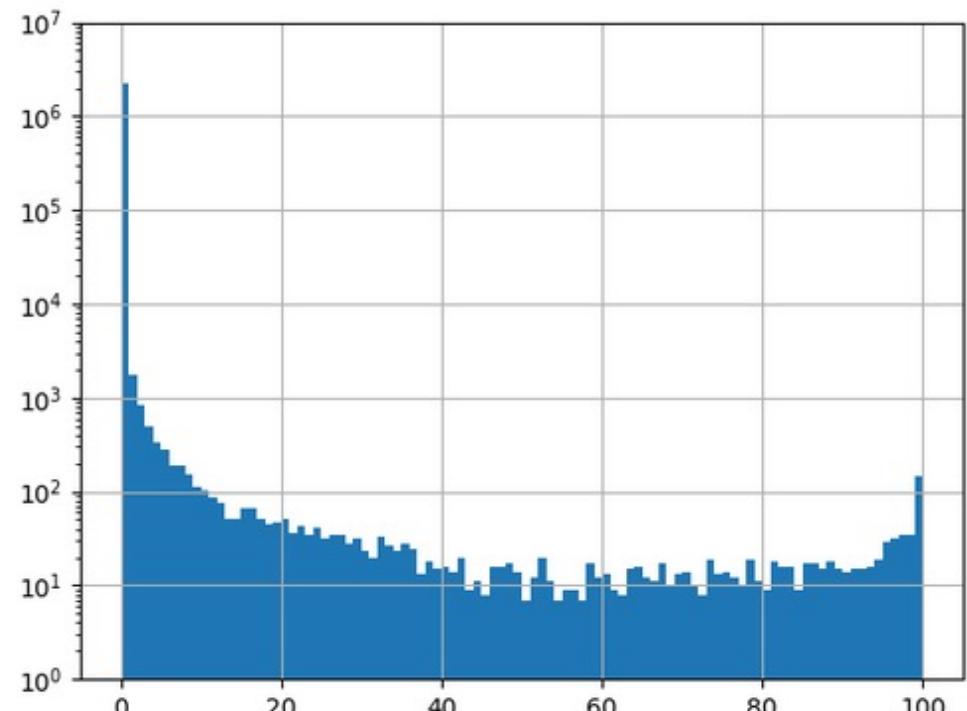
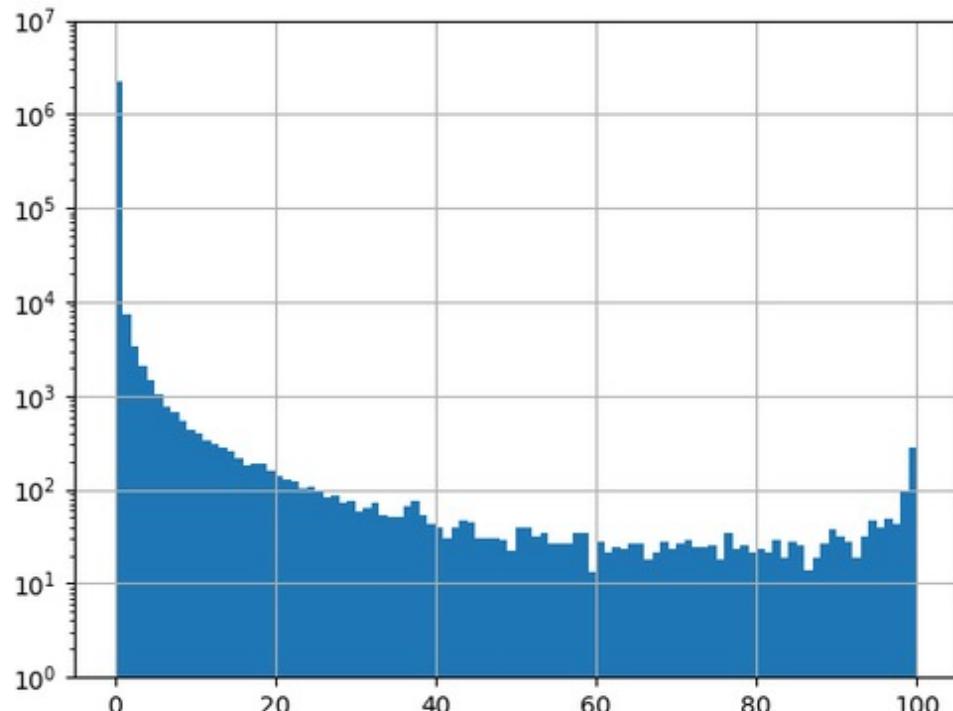
B-ResNet Distribution:

(0, 25) 2259070

(25, 50) 524

(50, 75) 299

(75, 100) 569





Results

Analysis Iteration 1 – Obvious False Positives

B-NYU Distribution:

(0, 25) 2257416

(25, 50) 1347

(50, 75) 675

(75, 100) 1024

B-ResNet Distribution:

(0, 25) 2259070

(25, 50) 524

(50, 75) 299

(75, 100) 569

Has MX Record	NO MX Record
08d749013a84d810edacd03c338e2b2e.ch	0767676767.ch
19zweiundsiebzig.ch	1augustbrunch.ch
1nf02m4c0253845714.ch	2millefeuille.ch
365prosicherheitsmanufaktur.ch	3nkh5crxol.ch
4897055461203223.ch	4aumf9m40jpe.ch
50jahrefcteufen.ch	59dc8ad64e5d3b35fb590.ch
6822beb0714d4417a5090fddaec2fa5f.ch	6qs8d6zd3d69s6.ch
aachmuehle.ch	718281828459045235360287471352662497757247.ch

...

Has MX Record	NO MX Record
08d749013a84d810edacd03c338e2b2e.ch	0000000019d6689c085ae165831e934ff763ae46a2a6c
19zweiundsiebzig.ch	125jahrefeuerwehrverbandluzern.ch
1nf02m4c0253845714.ch	3epilier-assurance.ch
365prosicherheitsmanufaktur.ch	3epilier-prevoyance.ch
4897055461203223.ch	4aumf9m40jpe.ch
50jahrefcteufen.ch	6qs8d6zd3d69s6.ch
6822beb0714d4417a5090fddaec2fa5f.ch	718281828459045235360287471352662497757247.ch
aachmuehle.ch	7hkjuf95uenl.ch
aaeeiioouu.ch	8qswldnsrvb73xkczdyj.ch

...



Results

Analysis Iteration 2 – Distributions of results

B-NYU Distribution:

(0, 25) 2260430

(25, 50) 13

(50, 75) 6

(75, 100) 12

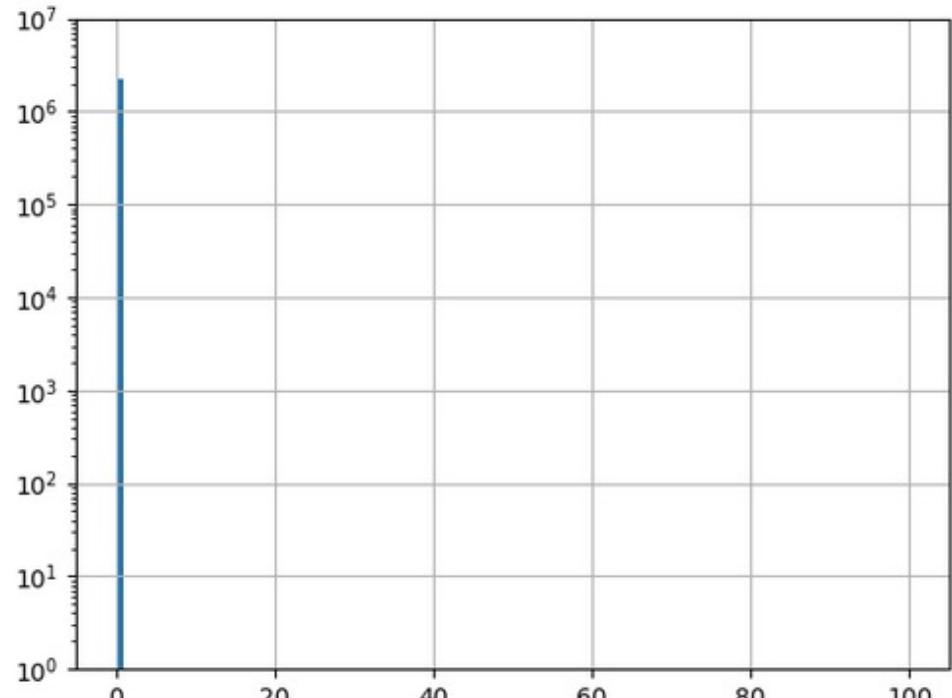
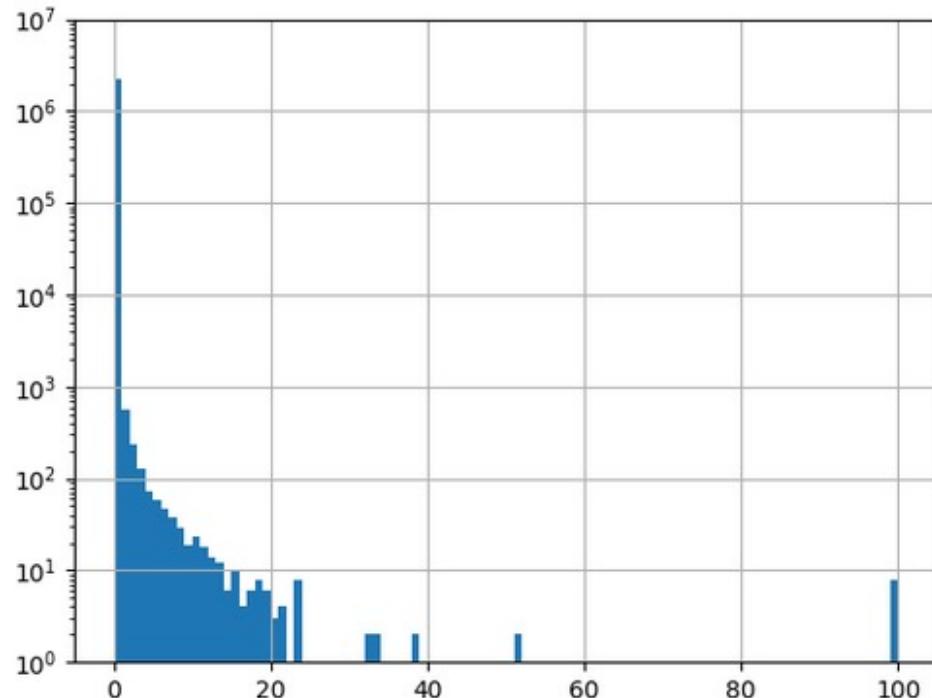
B-ResNet Distribution:

(0, 25) 2260462

(25, 50) 0

(50, 75) 0

(75, 100) 0





Results

Analysis Iteration 2 – Interesting detections 1/2

Domain	Certainty	Analysis / Conclusion
abcdefghijklmnoprstuvwxyz.ch	100%	Likely malicious , currently resolves to suspicious IP
adslkfalkfjlkfjdsalkfafljflsa.ch	100%	Unclear , no IP resolution
8qswldnsrvb73xkczdyj.ch	99.9%	Likely malicious , Currently resolves to suspicious IP, used to resolve to another suspicious IP involved in coronavirus-scam
rgdfgdfgdfgdf.ch	99.9%	Likely benign , no suspicious observations
utitan101310bgfhnythjdukfdyjt.ch	99.8%	Likely benign , no suspicious observations
sfdfgdfgdfgdfgdfg.ch	99.8%	Unclear , no IP resolution
n7q9ipiddq9ihtx.ch	99.1%	Likely malicious , currently resolves to suspicious IP
testhgfjdgdfxhgxdfhx12.ch	99.1%	Likely malicious , currently resolves to suspicious IP, IP has been flagged as botnet C2
oigweurpui345345jk.ch	94.1%	Likely benign , no suspicious observations
ymfvrcnwyw.ch	92.5%	Unclear , no IP resolution
aqdddwdxszedc.ch	84.8%	Unclear , no IP resolution
ihjj8qlfyfe.ch	82.2%	Likely malicious , domain flagged by Kaspersky, currently resolves to suspicious cloudflare IP
asdfjkhdsfajdfsajhsadf.ch	77.1%	Likely benign , no suspicious observations
7as6q796d6s98q6qd6sdq.ch	72.6%	Likely malicious , Currently resolves to suspicious IP
rggrgrgrgrgr.ch	66.5%	Unclear , no IP resolution



Results

Analysis Iteration 2 – Interesting detections 2/2

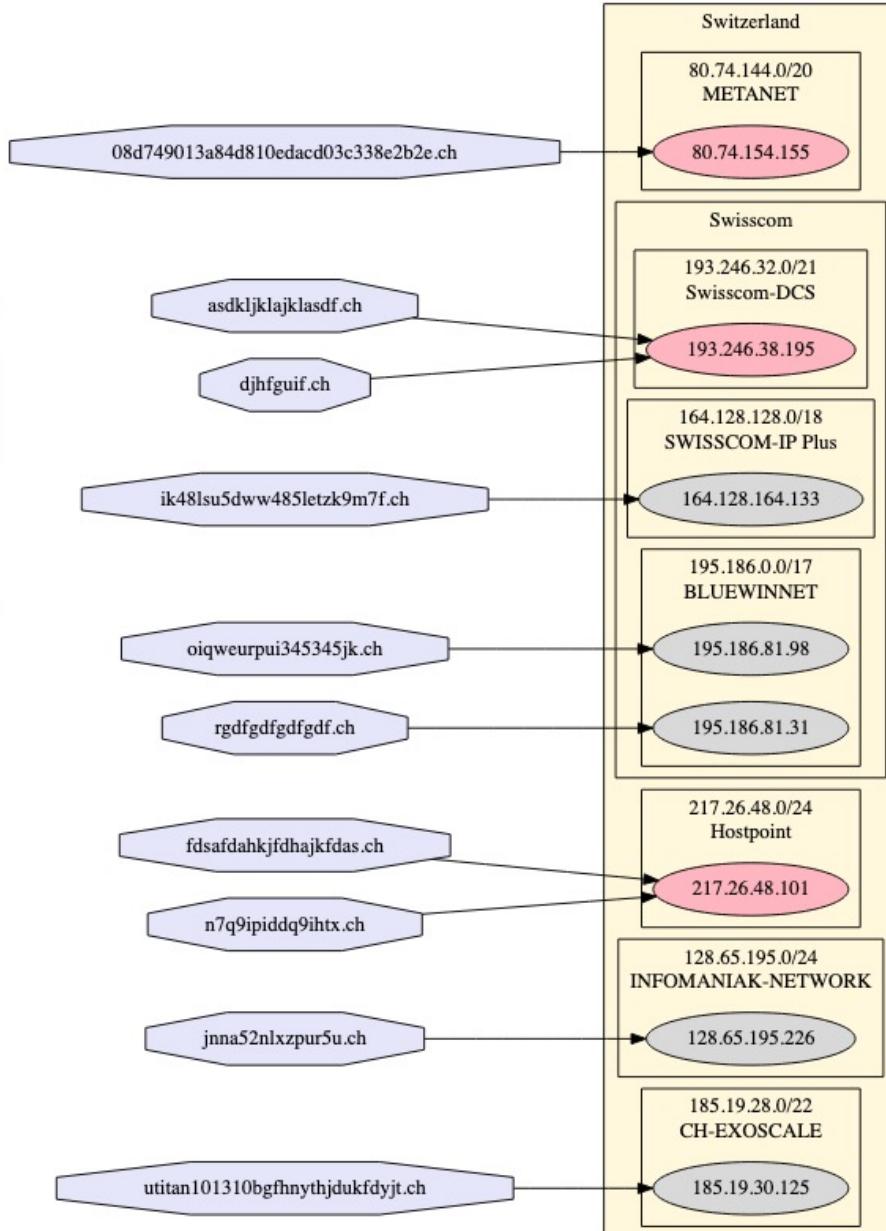
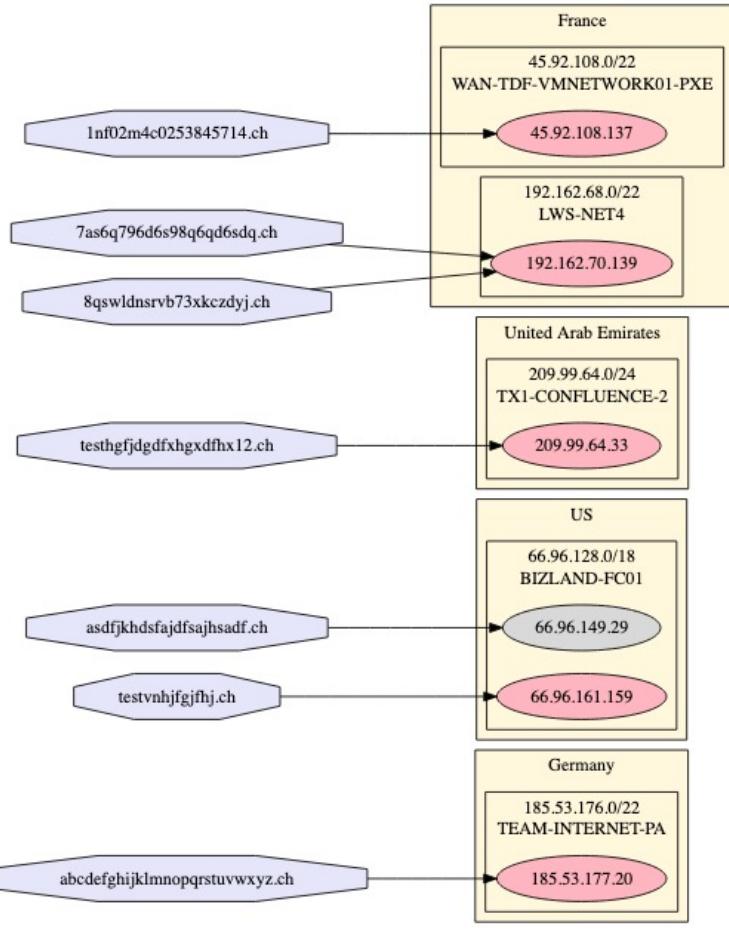
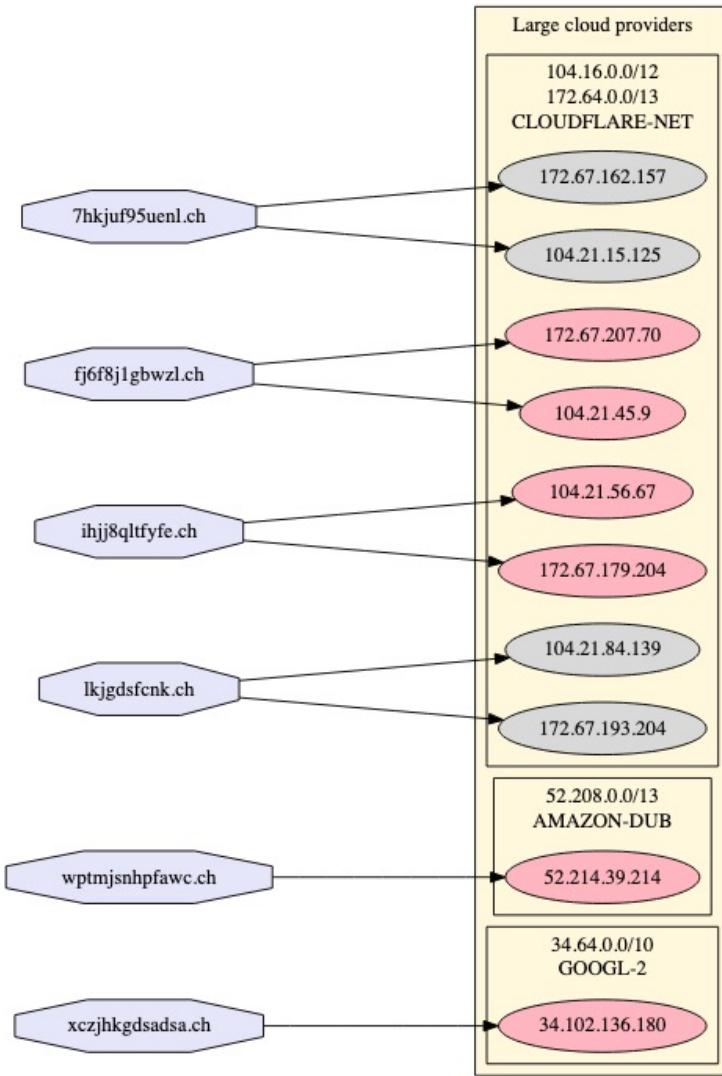
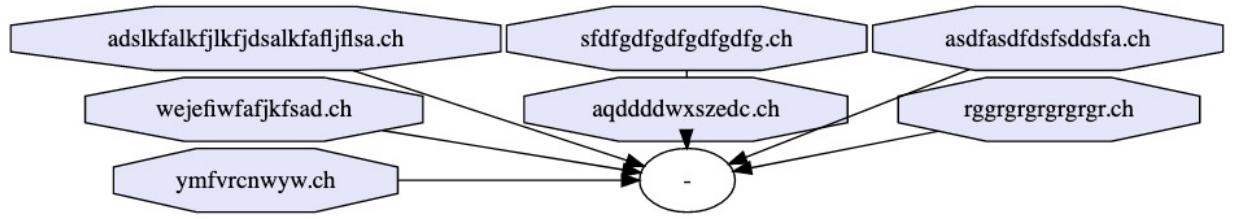
Domain	Certainty	Analysis / Conclusion
fj6f8j1gbwzl.ch	54.6%	Likely malicious , used to resolve to suspicious cloudflare IP (with very little associated URLs)
fdsafdahkjfdhajkfdas.ch	52.2%	Likely malicious , Currently resolves to suspicious IP
xczjhkgdsadsa.ch	51.3%	Likely malicious , Currently resolves to suspicious IP
ik48lsu5dww485letzk9m7f.ch	51.1%	Likely benign , no suspicious observations
asdkljklajklasdf.ch	48.3%	Likely malicious , Currently resolves to suspicious IP
1nf02m4c0253845714.ch	47.9%	Likely malicious , Currently resolves to suspicious IP
jnna52nlxzpur5u.ch	43.4%	Likely benign , no suspicious observations
wptmjsnhpfawc.ch	38.9%	Likely malicious , Currently resolves to suspicious IP
wejefiwfafjkfsad.ch	38.8%	Unclear, no IP resolution
asdfasdfsfsddsfa.ch	33.8%	Unclear, no IP resolution
testvnjhfgjfhj.ch	33.7%	Likely malicious , Currently resolves to suspicious IP
08d749013a84d810edac....e2b2e.ch	32.8%	Likely malicious , Currently resolves to suspicious IP
djhfguif.ch	32.6%	Likely malicious , Currently resolves to suspicious IP
7hkjuf95uenl.ch	30.1%	Likely benign , no suspicious observations
lkjgdsfcnk.ch	28.8%	Likely benign , no suspicious observations
neue-webermuehle-neuenhof.ch	25.9%	Likely benign , no suspicious observations



Results

Analysis Iteration 2 – Conclusion

Certainty of URL being malicious	Number of URLs	Verification result
75%-100%	13	5 urls are likely malicious , 4 urls do not resolve, 4 urls are likely benign , all urls do not "look" trustworthy
50%-75%	6	4 urls are likely malicious , 1 urls do not resolve, 1 urls are likely benign , all urls do not "look" trustworthy
25%-50%	12	6 urls are likely malicious , 2 urls do not resolve, 4 urls are likely benign , all but one urls do not "look" trustworthy
0%-25%	2260430	not inspected





Conclusion / What's next?

Conclusion:

- Applying classifiers based on generic training data to a language specific zone results in too many false positives
- Extending the training dataset with MX-domains helps to specialize a generic list of benign and malicious domains for a given country-TLD, this can be done in a self contained way
- The CNN based binary classifier works well for identifying suspiciously-looking domains, more than 50% of which are very likely malicious, thresholding remains tricky

Outlook:

- Sweden also publishes zonefile → repeat analysis to confirm results
 - Initial results indicate, that method seems to work even better in the case of .se-TLD
- Further improve accuracy by using more features

Thank
You!

