

A Practical Approach for Taking Down Avalanche Botnets Under Real-World Constraints

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Davy Preuveneers, Andrzej Duda, Wouter Joosen, Maciej Korczyński

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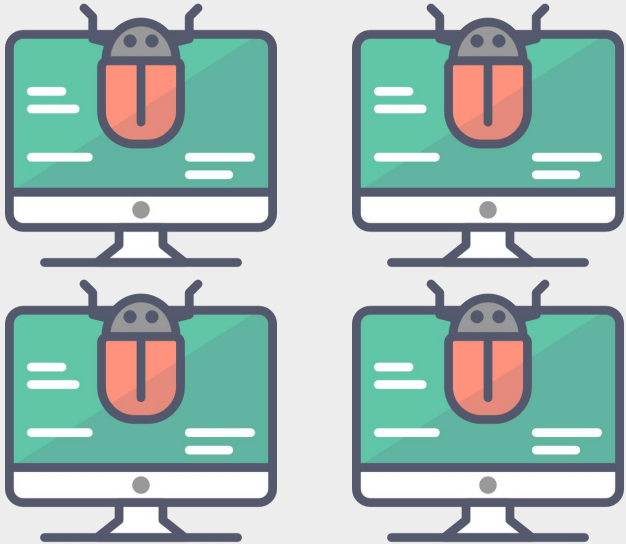
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“the world’s largest and most sophisticated cybercriminal syndicate law enforcement has encountered”

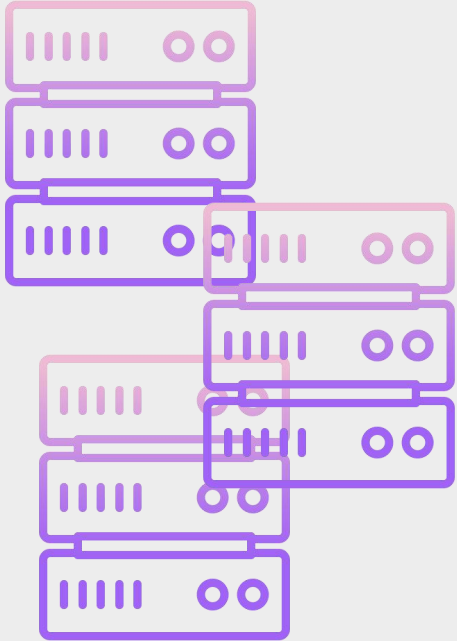
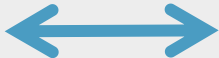
[Wai17]

Avalanche operated an advanced infrastructure

Infected client



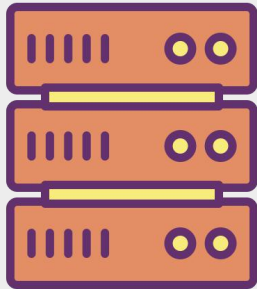
Infected hosts
serving as entrypoints

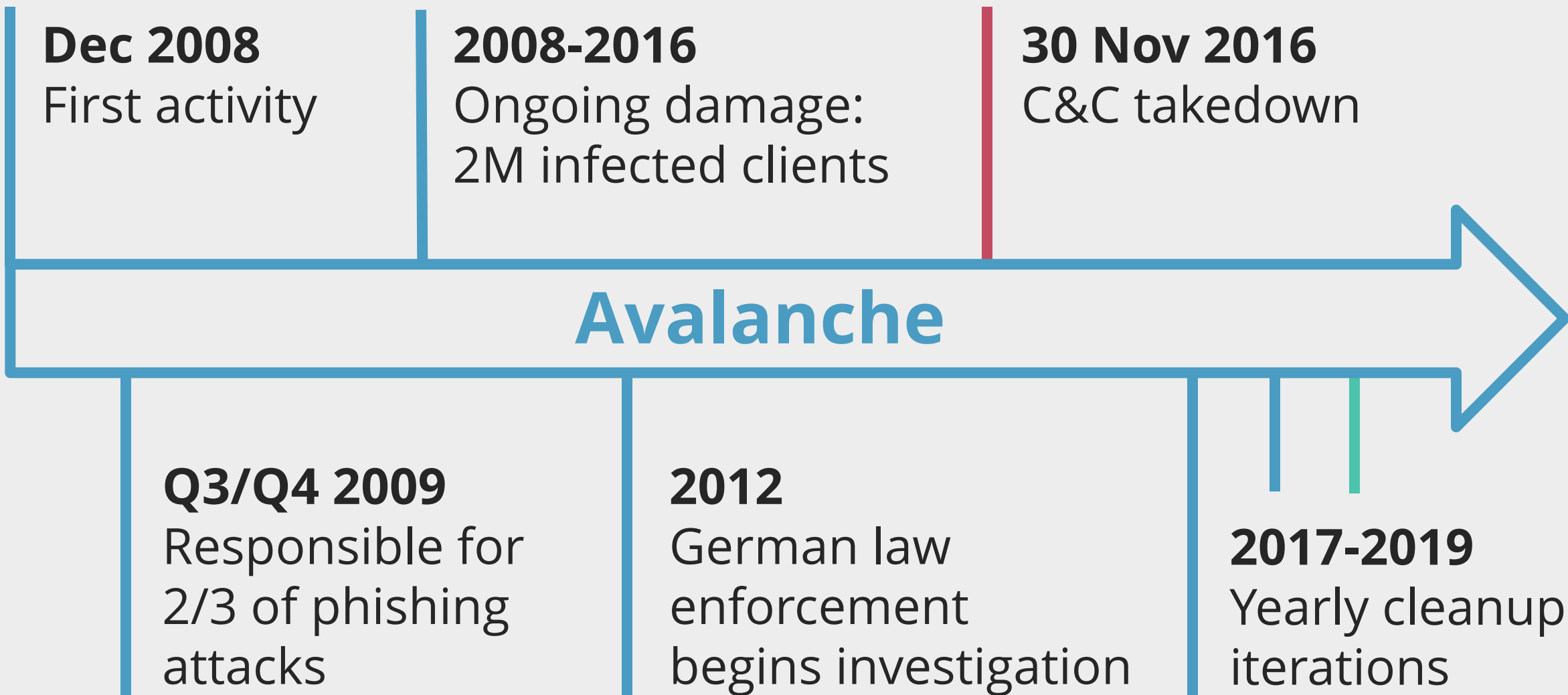


Layered network
of proxy servers



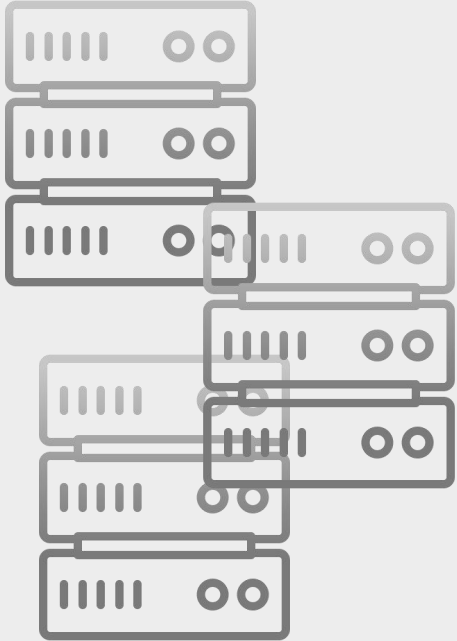
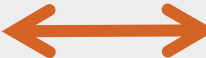
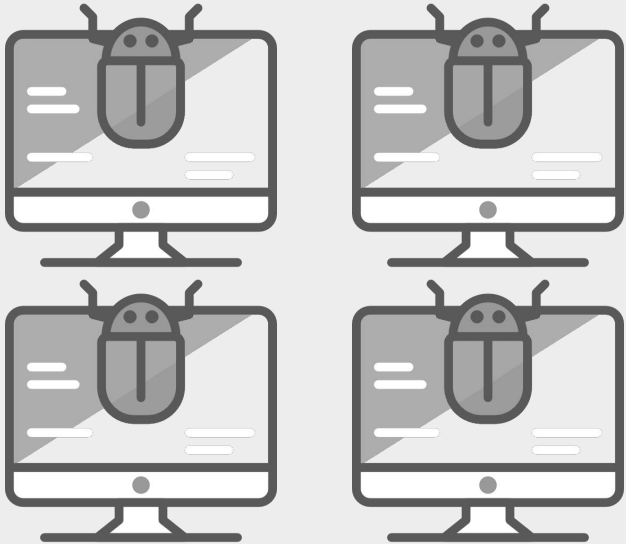
Core C&C
server



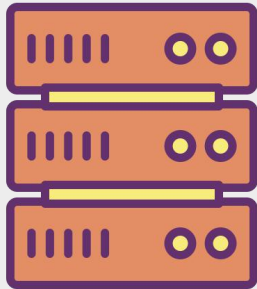


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Core C&C server



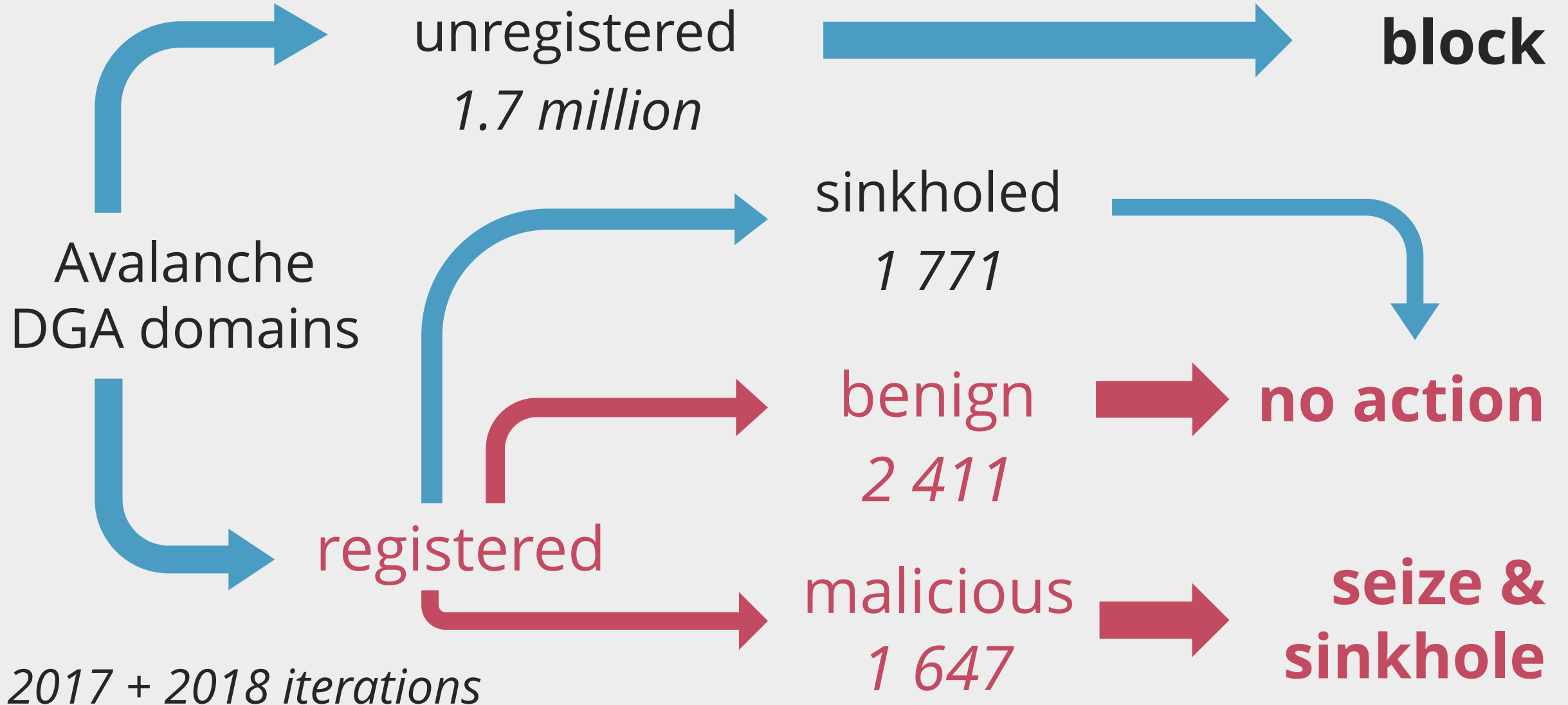
Domain Generation Algorithms

0a85rcbe2wb5n5f.com

researchmadness.com

arbres.com

Law enforcement has to classify registered domains



We evaluate on a **real-world** takedown: Avalanche

- › Design a more **automated** approach to reduce extensive **manual classification effort**
- › and assist in making **accurate** decisions
 - ›› Take down a *benign* domain: service interruption
 - ›› *Not* take down a *malicious* domain: botnet can respawn
- › leveraging (*limited*) **real-world ground truth**
 - › synthetic data sets may not be representative [Küh14, LeP19]

A Practical Approach for Taking Down Avalanche Botnets Under **Real-World Constraints**

Constraints affect available indicators

Individual patterns

in contrast to

bulk registration

[Hao16, Spo19]

bulk lexical patterns

[Woo16, Sch18]

Proactive analysis

in contrast to

presence/detection
of malicious activity

[Bil11, Ant12]

No active connections

in contrast to

active collection
of web content

[Khe14]

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Our experimental protocol mimics real takedowns

- › Enrich with **comprehensive** feature sets (*within constraints*)
- › Collect **historical data** as of iteration (*if possible*)

› Some domains have **missing data**

› Classify **all** domains
using **ensemble** model

<i>Data set</i>	<i>Missing</i>
WHOIS	14.6%
Passive DNS	8.7%
Active DNS	19.5%

train on

test on

accuracy

*F₁
score*

*Effort
saved*

	<i>train on</i>	<i>test on</i>	<i>accuracy</i>	<i>F₁ score</i>	<i>Effort saved</i>
Base	2017	2018	84.3%	73.4%	100%

▶ Concept drift

	<i>train on</i>	<i>test on</i>	<i>accuracy</i>	<i>F₁ score</i>	<i>Effort saved</i>
Base	2017	2018	84.3%	73.4%	100%
Extended A priori	2017 + 15% of 2018	Remaining 85% of 2018	86.4%	78.6%	85.0%

► Hybrid model: Human oracle

	<i>train on</i>	<i>test on</i>	<i>accuracy</i>	<i>F₁ score</i>	<i>Effort saved</i>
Base	2017	2018	84.3%	73.4%	100%
Extended A priori	2017 + 15% of 2018	Remaining 85% of 2018	86.4%	78.6%	85.0%
Base A posteriori	2017	2018	97.3%	95.3%	70.3%
Extended A posteriori	2017 + 15% of 2018	Remaining 85% of 2018	97.6%	95.8%	66.2%

*2019: 76.9%

We analyze influences on our model

<i>Set</i>	<i>Feature</i>
1 WHOIS	Time between creation...
2 WHOIS	Time between creation...
3 Passive DNS	Time between first seen...
4 Passive DNS	Time between first and...
5 WHOIS	Time between creation...
6 WHOIS	Renewal of domain ...
7 Active DNS	Days DNS record seen ...
8 WHOIS	Renewal of domain ...
9 Active DNS	Time between first seen...
10 Joint	Number of pages found...

› Important
time-based features
are **hard to evade**

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- › Important **time-based** features are **hard to evade**
- › **Data availability** affects **performance**
 - › Some **redundancy** exists

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We evaluate on a **real-world** takedown: Avalanche

- › Automating **classification** of registered DGA domains
- › Real-world setting yields **unique opportunity**
but also imposes **constraints**
- › **Hybrid** model: synergy between model and analyst
- › **Insights** for real-world takedowns

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