Empirical Analysis of Factors Affecting Malware URL Detection

Marie Vasek & Tyler Moore SMU, Dallas, TX mvasek@smu.edu

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So Much Malware



So Much Malware



So Many AntiVirus Products























Research Questions

Our hypothesis: At least part of the difference in performance across blacklists can be explained by characteristics of the malware URL or the blacklisting service.

- 1. Impact of type of malware URL on blacklisting?
 - Payload type (exploit kit, Fake AV, etc.)
 - URL features (IP address, path)
- 2. Impact of type of blacklisting service?
 - Blocking or advisory
 - Cost

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Malware Domain List

- Publicly Available
- Low Volume

8 of 36

- o 722 URLs in our collection
- BUT it's free, open, & no data-sharing issues

ATMzIDfvqzMwEmMjFjM1

MALWARE DOMAIN LI

Homepage | Malware Domain List | Forums | RSS update feed | Contact us

WARNING: All domains on this website should be considered dangerous. If you not know what you are doing here, it is recommended you leave right away. Th website is a resource for security professionals and enthusiasts.

Date (UTC)	Domain	IP	Reverse Lookup	Descrip
2013/07/11_12:54	www.keepsaketributes .com/calc/images/ac0094cbbe/? ==wMw1mLulWYt9VbzXXO5IDMXIzN3QD O5UDO2w3LlbmYjRTOwAz Yh9ycldWYtl2LjxWYj95 bvNmLzVGd1JWayRXZrF2 cwVWZr5yd3d3LvoDc0RHa8NnZ		host2.jbatkins.com.	exploit kit
	www.lowes-pianos-and- organs.com/images/d521c2a038/?	67 222 100 112	d15 alteerver com	ycuF2Zy9WLk

VirusTotal

- Publicly Available
- API accessible
- Google-Owned
- Checks each URL against 38 blacklists



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What We Do

- 1. Check MDL for new URLs.
- 2. Scan new URLs through VT.
- 3. Scan new URL every other hour for 48 hours after.
- 4. Afterwards, scan every day for 14 days.

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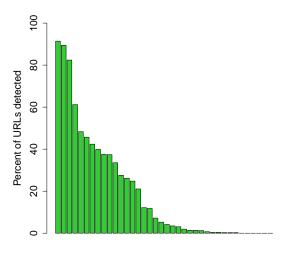
Data

Factors

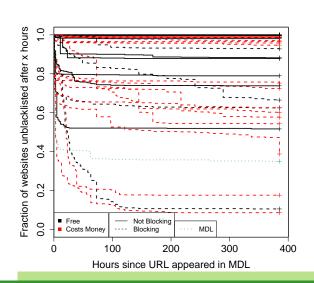
Results

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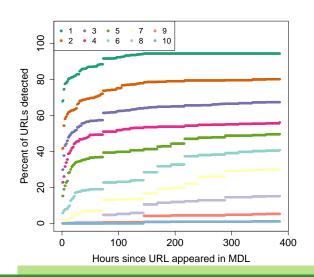
How Blacklisting Detection Rates Vary



How Blacklisting Detection Rates Vary



Percentage of URLs Detected by at least *n* Services



Blacklisting over Time

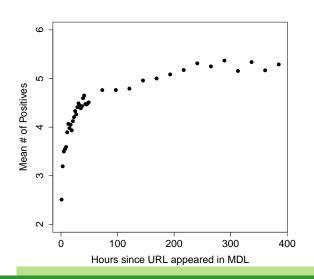


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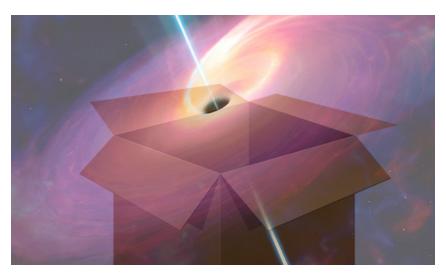
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Malware Type



IP/Domain

127.0.0.1/bad.html

icanhazcats.com/milk.php

Path?

fluffybunnies.org/

goldphish.net/bubbles.php

Type of Malware

Malware Type	#	%	IP/Domain	#	%	Path?	#	%
Executable Fake AV Styx Blackhole Lnd. Other	175 65 51 149 282	24% 9% 7% 21% 39%	IP Address Domain	124 598	17% 83%	Has Path No Path	675 47	93% 7%

Blocks?

Web Page Blocked

Access to the web page you were trying to visit has been blocked in accordance with company policy. Please contact your system administrator if you believe this is in error.

User: 129.119.9.166

URL: www.malwareconference.org/index.php?option=com_docman_task=cat_view_gid=79_Itemid=51

Category: malware-sites

Costs Money



Blacklist Types

osts? Descrip	otion
✓ Paid A	V
X Free A	V & Web Browser Blacklists
✓ Malwai	e Site Checkers
X Malwa	e URL lists
	✓ Paid A\ ✓ Free A\ ✓ Malwar

Blacklist Types

Blocks?	#	%	Costs?	#	%
Blocks Users	22	58%	Costs Money		45%
Doesn't Block	16	42%	Free		55%

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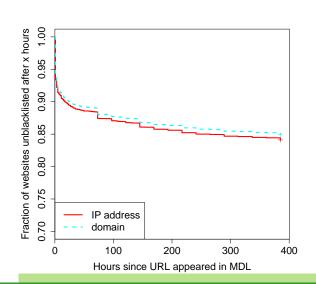
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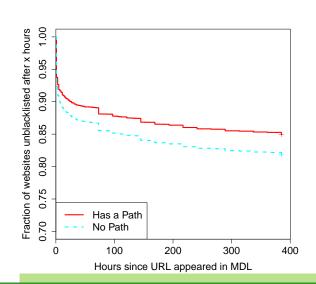
Logistic Regression: URL ever Blacklisted by Service?

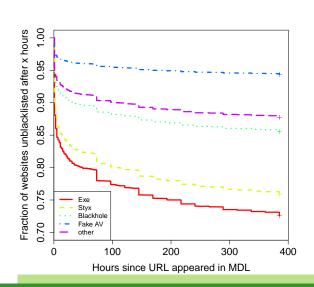
	Coefficient	Odds Ratio	p value				
Intercept	-2.304	0.100	0.000				
URL features							
IP address	-0.004	0.996	0.945				
Has a Path?	-0.196	0.822	0.031				
Executable	1.017	2.765	0.000				
Fake AV	-0.838	0.433	0.000				
Styx	0.746	2.109	0.000				
Blackhole Landing Page	0.196	1.217	0.000				
Malware Blacklist Features							
Blocks Users?	0.611	1.843	0.000				
Costs Money	0.298	1.347	0.000				
$\chi^2 = 1144.501$, p value $= 0.000$							

Survival Regression: Blacklist Timing

Coefficient		Odds Ratio	p value			
URL features						
IP address	0.056	1.058	0.210			
Has a Path?	-0.207	0.811	0.012			
Executable	0.896	2.449	0.000			
Fake AV	-0.814	0.443	0.000			
Styx	0.750	2.118	0.000			
Blackhole Landing Page	0.179	1.196	0.000			
Malware Blacklist Features						
Blocks Users?	0.538	1.713	0.000			
Costs Money	0.300	1.351	0.000			
$R^2 = 0.055$						







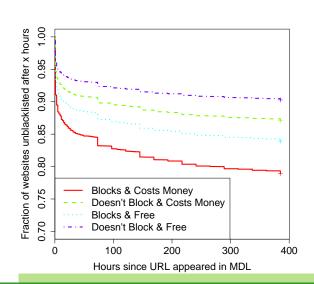


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Summary

- Known Exploit kits more likely to be blacklisted
 ... but Fake AV is not ..?
- · Executables more likely to be blacklisted
- ullet Blocking users \longrightarrow more likely to blacklist
- ullet Costing money \longrightarrow more likely to blacklist

Limitations

- MDL → low volume
 - o other exploit kits
 - AS / DNS features
 - Malicious vs. Hacked?
- ullet MDL \longrightarrow used as input to blacklists
 - Blacklisted b/c found on VT or found organically?
 - Are things on MDL still bad on day *n*?
- VT → not representative sample of malware blacklists
 - Blacklists not on VT?
 - Blacklists on VT that didn't blacklist anything?

Questions?

