# Automated Attacks at Scale

Understanding "Credential Exploitation"

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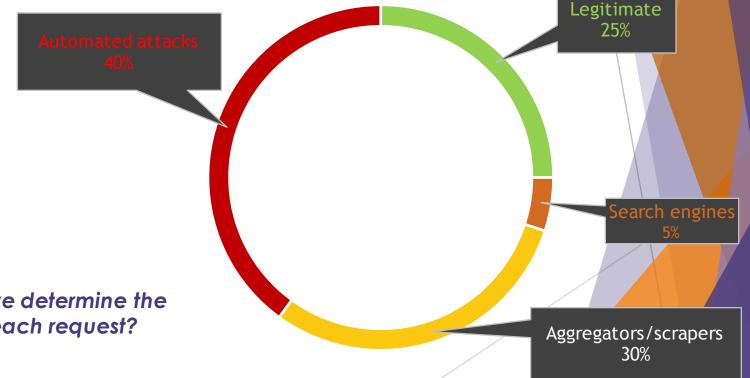
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### What do we mean by an "Automated Attack"? Fundamentally a Bot problem

- Attack toolkits available on underground
- Custom scripts
- Attacks on API endpoints



How do we determine the intent of each request?

# Attacker's Goals

Account Take Over



Fake Account Creation



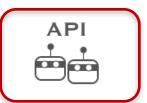
PII / PHI Theft



**Shopping Bots** 



**API** Abuse



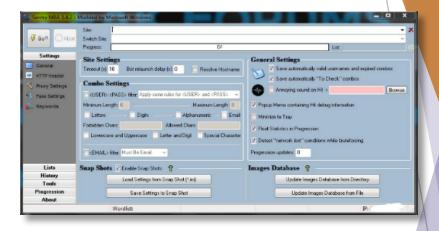
# The Attacker's Perspective

# The 5 Pillars of a credential exploitation attack

- 1) Black Market Attack Tool or Custom tool configured for a target
- 2) Set of Stolen Credentials
- 3) Ability to rotate over many IP addresses
- 4) Compute Power
- 5) Ability to bypass deployed security solutions

# Attack Toolkits & Config Files

- SentryMBA
- Hydra
- PhantomJS
- Medusa
- Curl, Wget
- Ncrack
- Other custom scripts



#### Understanding Config Files...

- Program instructions for how to login and differentiate between failed and successful logins for that particular target. Writing config files is one of the chief ways to monetize in this criminal ecosystem.
- "Capture" setting optional setting enables attackers to understand the value of a compromised account without logging back in again.

# Quick Facts - Underground Ecosystem

- 1,853 unique target sites on sentry.mba
- 10% of Alexa Top 1000 have config files readily available
- 184 API config files roughly 10% of targets
- \$1.73 average cost of a config file.
- Top industries targeted Gaming, Entertainment, E-Commerce

#### POPULAR TARGET SITES

Popular Streaming, Gaming and Social Networking websites are also attackers' favorite targets. This may indicate most attackers are script kiddies.

884 Downloads Reposted 25 times







HBO 289 Downloads Reposted 19 times











SONY

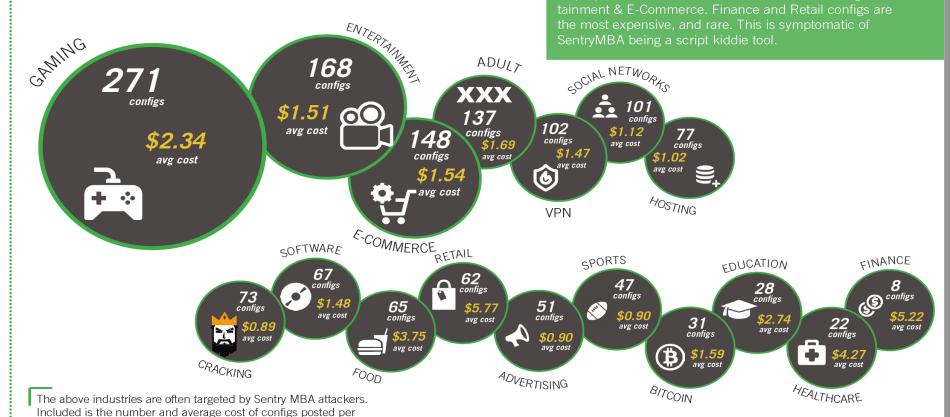
[ target industries ]

All major industries are actively under attack. Some face a

disproportionate volume of attacks such as Gaming, Enter-

industry.

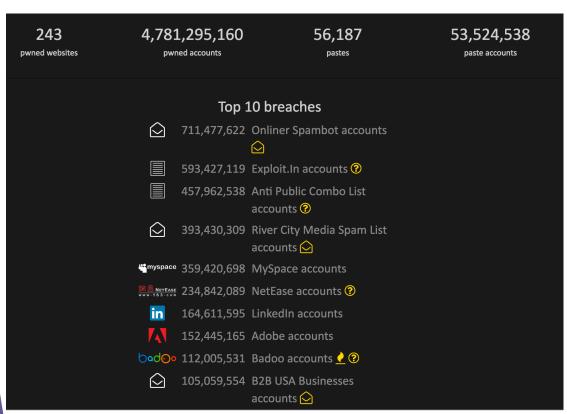
#### TARGET INDUSTRIES



# The 5 Pillars of a credential exploitation attack

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#### Stolen Credentials



- Simple Pastebin
   Crawler harvests
   more than 20,000
   credentials every
   day
- Users average 6.5 credentials per 50 websites

\* Microsoft Research

<sup>\*</sup> https://haveibeenpwned.com/

# Quick aside - How much money can attackers really net?

Social Security number (sold as part of 'Fullz' dossier)	\$30
Date of birth	\$11
Health insurance credentials	\$20
Visa or MasterCard credentials	\$4
American Express credentials	\$7
Discover credit credentials	\$8
Credit card with magnetic stripe or chip data	\$12
Bank account number (balance of \$70,000 to \$150,000)	\$300 or less
Full identity 'Kitz'	\$1,200 to \$1,300

Source: Dell SecureWorks

Attacker tries 1,000,000 credentials - if each stolen account sells for only \$0.25, then a successful login rate of only **0.1%** will net \$250.00

# The 5 Pillars of a credential exploitation attack

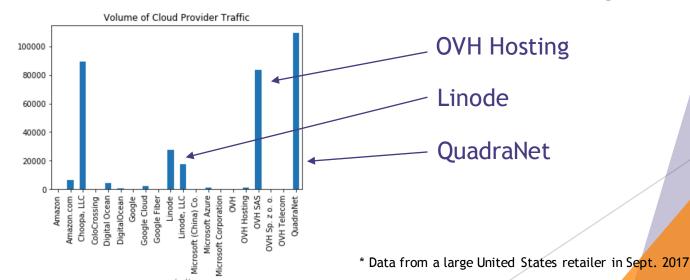
- 1) Black Market Attack Tool or Custom tool configured for a target
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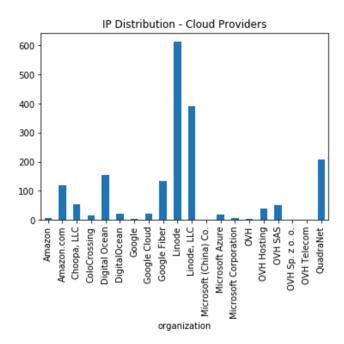
# IP Rotation & Compute Power

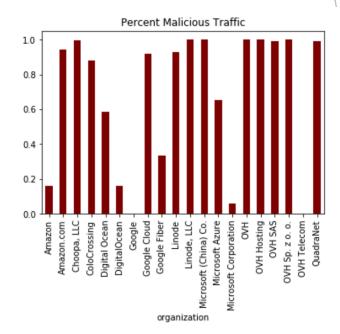
How to gather the necessary infrastructure?

# Option 1: Cloud Hosting Providers

- High reputation AWS & Azure will never get blacklisted
- Virtualization allows easy instance creation programatically



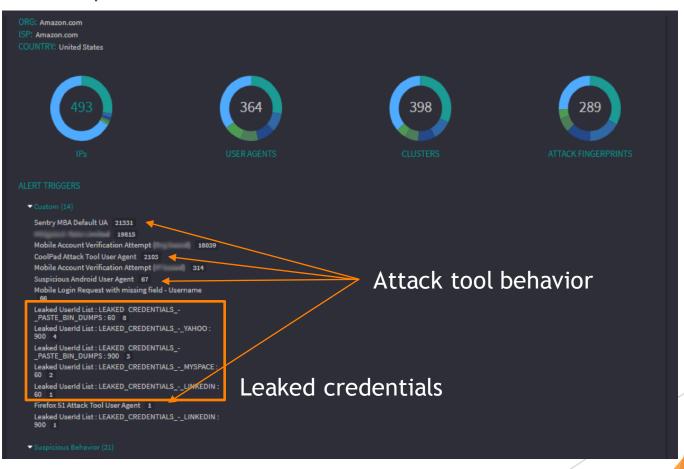




How long do these IP's "stick around" and continue sending malicious traffic before being recycled?

Answer: Surprisingly long...

### Example: AWS



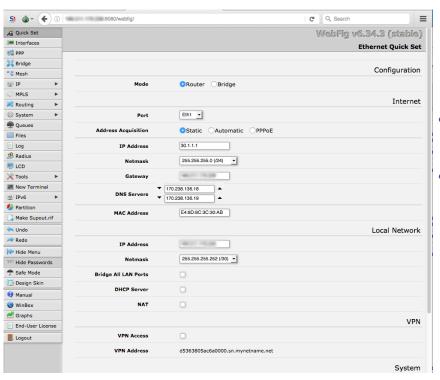
# Option 2: Compromised Devices, IoT Botnets

- Easily exploitable routers, old firmware models & default credentials available with a quick google search
- Client side fingerprinting challenges for defenders
- Available for rent in black market

#### Data Observed December 2016-2017 at large financial institution

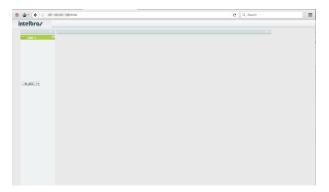
- Device Types: 175 open home routers, 10 DVR/camera systems, 10 web servers (incl. Apache Tomcat), 4 webcams, 1 SCADA system
- Common ISPs Telmex (25%) (Mexico), VDC (Vietnam),
   Claro Dominican Rebublic, Link Egypt, Telefonica del Peru,
   TE Data (Egypt), Qubee (Pakistan)

# Example - Open routers

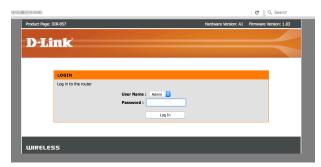


- Admin page open to public on port 8080
- SSH logs showed other attackers trying to brute force login via SSH – "tugof-war" between attackers

# Other device examples:



Intelbras camera system



D-Link, Huawei HG532 and HG8245H, Advantech WebAccess browser-based HMI/SCADA software system (not pictured)

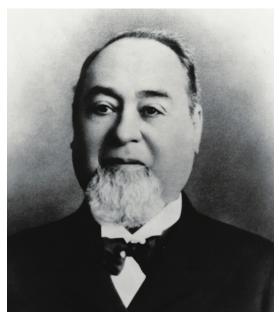




Mikrotic (v6.36.4 and v6.34.3)

# Option 3: An Artificially Geo-Distributed Proxy Farm - "The AWS for bad guys"

Levi Strauss



California Gold Rush of 1848 And the creation of Levi's jeans

#### Who is this actor and what are some indicators?

inetnum:

% Abuse contact for '5.8.44.128 - 5.8.44.255' is 'abuse@pinspb.ru'

5.8.44.128 - 5.8.44.255

#### Orgs, ISPs, ASNs

#### **ISPs**

- Petersburg Internet Network ltd. 38.7%
- Transit Telecom LLC -- 15.6%
- Atomohost -- 15%
- Link Telecom LLC -- 7.5%
- PP Trusov Ilya Igorevych -- 4.8%

#### Orgs

- DepoDataCenter -- 25%
- net for depo40.ru -- 25%
- Atomohost -- 11.5%
- Petersburg Internet Network ltd. 9.5%

#### ASNs

- 50896
- 29802
- 200557
- 44050,32181,44750

```
PanamaNet
netname:
descr:
                Network Panama
country:
admin-c:
                TII34-RIPE
tech-c:
                TII34-RIPE
status:
                ASSIGNED PA
                MNT-DEP040
mnt-bv:
                2015-01-24T21:22:19Z
created:
last-modified:
                2015-01-24T21:22:19Z
source:
                RIPE
```

person: Trusov Ilya Igorevych
address: 249806, Russia, Kaluga region, Moscow Street 258, office 16
phone: +79533100064
abuse-mailbox: abusemail@depo40.ru
nic-hdl: TII34-RIPE

mnt-by: MNT-DEP040
created: 2015-01-24T20:24:14Z
last-modified: 2015-11-20T19:04:51Z

5.189.207.0/25

5.189.207.128/25

source: RIPE 5.189.200.0/21 net for depo40.ru DEPO-NET 5.189.200.0/25 CapeTownNet South Africa Network ZA 5.189.200.128/25 SeoulNetwork South Korea Network SK 5.189.201.0/25 CairoNet Egypt Cairo Network EG 5.189.201.128/25 Jakarta Indonesia Jakarta Indonesia Network ID 5.189.202.0/25 Islamabad Network PK IslamabadNet VN 5.189.202.128/25 Hanoi Network Hanoi-Vietnam-Network 5.189.203.0/25 Singapore Network SG SG-Net 5.189.203.128/25 Las Vegas Network Nevada Las Vegas Network US 5.189.204.0/25 TehranNet Iran Network IR Ogden Utah Network US 5.189.204.128/25 OgdenNet CA 5.189.205.0/25 VancouverNetwork Canada Vancouver Network France Network FR 5.189.205.128/25 ParisNetwork 5.189.206.0/25 Network Alaska US US AlaskaNetwork 5.189.206.128/25 Helsinki Network Helsinki Finland Network FΙ

Network Infrastructure to Miam US

Sweden Network

SE

Miami Network

SwedenNetwork

### More Indicators...

abuse-mailbox:

last-modified:

mnt-by:

created:

source:

organisation: ORG-TII6-RIPE org-name: Trusov Ilya Igorevych org-type: LIR address: Moscow Street 258, office 16 address: 248021 address: Kaluga address: RUSSIAN FEDERATION +79533100064 phone: mnt-ref: RTPE-NCC-HM-MNT mnt-ref: MNT-DEPO40 RIPE-NCC-HM-MNT mnt-by: abuse-mailbox: abusemail@depo40.ru descr: Kaluga Data Center Depo created: 2013-11-08T11,14,03Z last-modified: 2017-03-29T11,44,15Z source: RIPE iluxa85@inbox.ru e-mail: AC28994-RIPE abuse-c: Trusov Ilya Igorevych person: remarks: Depo Data Center Kaluga address: 248021, Russia, Kaluga region, Moscow Street 258, o phone: +79533100064 nic-hdl: TII10-RIPE e-mail: noc@depo40.ru

abusemail@depo40.ru

2013-07-19T09,32,30Z

2017-03-26T13,29,22Z

MNT-DEPO40

RIPE

00	okup Connected Domains		<u>Lookup ti</u>
k	ouy.fineproxy.org		LOOKUP
xa	mple: 65.55.53.233 or 64.233.161.%		
ers	se IP Lookup Results — more than 3 c	domains hosted on IP address 104.25.240.28  View Whois Record	Screenshots
ers			Screenshots
	Domain	View Whois Record	

IP history results for fineproxy.org. -----

198.211.121.105 Amsterdam - Netherlands

IP Address	Location	IP Address Owner	Last seen on this IP
188.166.44.117	Amsterdam - Netherlands	Digital Ocean, Inc.	2017-10-04
104.25.241.28	United States	Cloudflare, Inc.	2017-08-30
104.25.240.28	United States	Cloudflare, Inc.	2017-08-30
188.166.44.117	Amsterdam - Netherlands	Digital Ocean, Inc.	2017-08-12
104.25.241.28	United States	Cloudflare, Inc.	2017-08-10
104.25.240.28	United States	Cloudflare, Inc.	2017-08-10
104.25.42.16	United States	Cloudflare, Inc.	2016-12-25

DigitalOcean, LLC

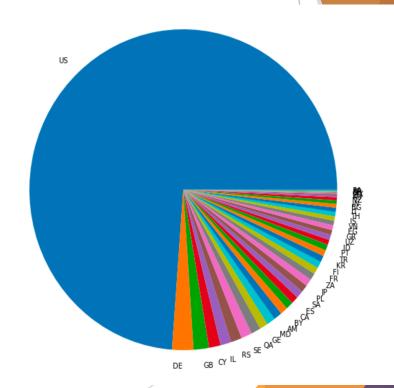
2016-12-17

### Case Study: Large US Retailer

#### **Attack Statistics**

- > 2% of login traffic for over 4 months
- At least 6 unique attack tools used
- 40,000 IP addresses from 61 countries
- Nearly 75% of traffic blending in with US customers
- Thousands of accounts compromised every week

Country Distribution according to MMDB



# Was this traffic really coming from the US?

#### Distributed Traceroute Experiment

RTT from Moscow

RTT from Washington

	N	Max_RTT	IP			Max_RTT	IP
	mean	median	count		mean	median	count
MM_City				MM_City			
Albuquerque	69.753600	3.9270	15	Albuquerque	113.620067	133.8720	15
Anchorage	602.785067	3.9070	15	Anchorage	735.037600	137.0600	15
Baltimore	4.049400	3.8800	15	Baltimore	132.246600	131.7100	15
Cedar Falls	3.695688	3.7005	16	Cedar Falls	135.660375	134.7540	16
Dallas	3.818667	3.8140	15	Dallas	136.748600	133.9290	15
Detroit	356.223118	3.7030	17	Detroit	138.642529	134.1860	17
Honolulu	5.079800	3.8020	15	Honolulu	131.118867	130.6670	15
Las Vegas	318.735211	3.6900	19	Las Vegas	455.526316	139.1030	19
Los Angeles	3.841933	3.8420	15	Los Angeles	134.830067	133.8160	15
Miami	203.213533	3.6720	15	Miami	335.337733	133.4570	15
None	4.649216	3.8670	51	None	136.346333	135.5950	51
Ogden	3.766667	3.7300	15	Ogden	136.639333	136.7310	15
Orlando	3.828545	3.8160	22	Orlando	137.369773	134.7120	22
Portland	3.900625	3.8340	16	Portland	133.129375	131.1810	16
Seattle	3.903438	3.8260	16	Seattle	135.956437	134.1655	16

RTT from Moscow

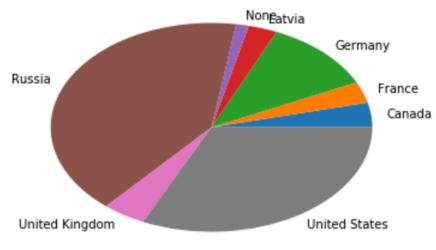
RTT from Washington

		Max_RTT	IP			Max_RTT	IP
	mean	median	count		mean	median	count
MM_Country				MM_Country			
Argentina	3.705111	3.6865	18	Argentina	136.020167	134.8850	18
Armenia	4.905533	3.6880	15	Armenia	136.280800	134.9330	15
Australia	3.690800	3.6910	15	Australia	137.350467	136.7090	15
Belarus	4.098400	3.7400	30	Belarus	135.353300	134.4010	30
Brazil	377.541938	3.7130	16	Brazil	511.674937	136.7345	16
Bulgaria	3.922000	3.6870	15	Bulgaria	135.216600	134.1660	15
Canada	235.931500	3.7685	30	Canada	327.714200	133.7280	30
Chile	4.533000	3.7110	15	Chile	137.067067	138.1580	15
China	3.924812	3.7580	32	China	137.494906	136.5790	32
Colombia	180.153412	3.7380	17	Colombia	313.156765	136.7450	17
			•				

#### Distributed Traceroute Experiment

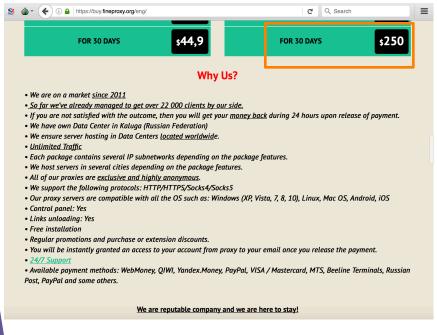
		Chicago	×	Los Angeles	×	Moscow	×
Moscow	×	• 143.327ms		• 213.498ms		_	
Tokyo	×	• 158.802ms		• 109.451ms		• 305.845ms	
Washington	*	• 35.241ms		• 62.305ms		• 136.09ms	
Zurich	*	• 120.878ms		• 147.692ms		• 49.58ms	

<sup>\*</sup> https://wondernetwork.com/pings

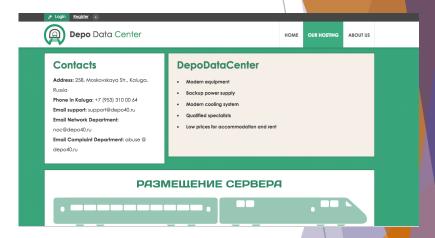


 Country labels according to MMDB for traffic from USA

### How do they monetize?



 Remember that "break even" point of \$250 with a 0.1% successful login rate? Possible to hit that within 1-3 days.



#### <u>Defender's Challenge:</u>

How can we detect these attacking a proactive way instead of reactive?

# The Defender's Perspective

# The 5 Pillars of Detection for protecting against automated attacks at scale

- 1) Analysis of HTTP/HTTPS requests and headers to fingerprint attack tools
- 2) Machine learning models to detect forged browser behavior
- 3) Threat intelligence designed to starve attackers of resources (IP addresses, compute power, stolen credentials)
- 4) Data analytics beyond the individual transaction level need to detect "recon" behavior & "low and slow" attacks
- 5) Technology that covers Web, Mobile & API channels attackers move to wherever there is the least resistance

# Case Study: SentryMBA – the "plug & play" attack tool Pillar 1: HTTP Request Fingerprinting



#### Default User-Agent Strings

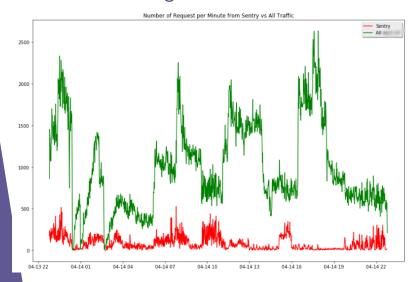
- Mozilla/4.0 (compatible; MSIE 7.0; Windows NT 6.0; NET CLR 1.1.4322; .NET CLR 2.0.50727; .NET CLR 3.0.4506.2152; .NET CLR 3.5.30729)
- Mozilla/4.0 (compatible; MSIE 8.0; Windows NT 6.1; .NET CLR 1.1.4322; .NET CLR 2.0.50727; .NET CLR 3.0.4506.2152; .NET CLR 3.5.30729)
- Mozilla/5.0 (Windows; U; Windows NT 5.1; en-US; rv:1.9.0.11)
   Gecko/2009060215 Firefox/3.0.11
- Mozilla/5.0 (Windows; U; Windows NT 5.1; en) AppleWebKit/522.11.3 (KHTML,, like Gecko) Version/3.0 Safari/522.11.3
- Opera/9.80 (Windows NT 6.0; U; en) Presto/2.2.0 Version/10.00
- Mozilla/4.0 (compatible; MSIE 6.0; Windows NT 5.1) \*\*Testing UA\*\*

#### <u>SentryMBA HTTP Fingerprint observations</u>

- We analyzed over 1500 config files and found that only 12% changed the request fingerprint
- Often missing referrer, accept-language or accept-encoding

#### Traffic Patterns

- Both high velocity and low & slow attacks. Suggesting multiple actors using the tool
- Recon activity w/ successful login ratios <
  .01% and verified credential attacks w/
  successful login ratios > 95%



- 150,000 requests from 3,385 IP's and 1,293 Organizations (1 day).
- Leaked credentials from MySpace, Yahoo, LinkedIN, others



# The 5 Pillars of Detection for protecting against automated attacks at scale

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# Case Study: Drago & Vlad – "Forged Browser Family"

Pillar 2: Forged Browser detection - ML

#### <u>Attack Tool "Vlad"</u>

Mozilla/5.0 (Windows NT 10.0; WOW64; rv:40.0) Gecko/20100101 Firefox/40.0

- Impersonating Firefox 40 on Windows 10
- Behaves similar to a command line tool like Wget or Curl

#### Attack Tool "Drago"

Mozilla/5.0 (Windows NT 6.3; WOW64) AppleWebKit/537.36 (KHTML, like Gecko) Chrome/56.0.2924.87 Safari/537.36

- Impersonating Chrome 56 on Windows 8.1
- Doesn't behave like any other browser in Chromium family

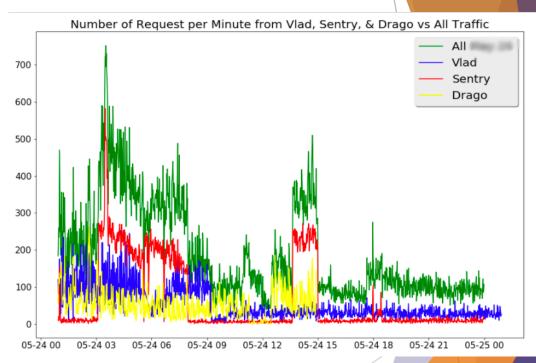
#### Traffic Patterns

#### Drago

More than 3,769 ISPs, 4,160
 Organizations and more than 150 countries, with no single ISP/Organization being responsible for more than 3.5% of the tool's traffic.

#### Vlad

 All traffic claimed to come from the US, yet every request had Accept-language header value equal to "ru-RU"



 Attack tools were responsible for every large spike in traffic, resulting in massive infrastructure overprovisioning.

# The 5 Pillars of Detection for protecting against automated attacks at scale

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### Case Study: Leaked Credentials

Pillar 3: Threat Intelligence targeted at resources attackers need

#### Top Data Breaches Observed per Attack Tool

#### <u>SentryMBA</u>



23%



19%

#### **Adobe**



17%

 Each username tried appeared in an average of 3.5 breaches

#### <u>Vlad</u>



**32**%



25%



22%

#### **Adobe**

 Each username tried appeared in an average of 3.4 breaches

#### Legitimate Traffic



No Breaches 42%



15%



11%

 Each username tried appeared in an average of 2.6 breaches

# The 5 Pillars of Detection for protecting against automated attacks at scale

- 1) Analysis of HTTP/HTTPS requests and headers to fingerprint attack tools
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### Case Study: "CoolPad" & Firefox

Pillar 4: Detection and Visibility across Web, Mobile & API

#### "Coolpad" Attack Tool

- Mozilla/5.0 (Linux; Android 4.4.2; Coolpad 8675 Build/KOT49H)
   AppleWebKit/537.36 (KHTML, like Gecko) Version/4.0 Chrome/30.0.0.0 Mobile Safari/537.36
- Responsible for 97.2% of traffic to a legacy API login
- A popular Chinese mobile device which for a US retailer raised a red flag



#### Firefox 51 Attack Tool

- Mozilla/5.0 (Windows NT 10.0; WOW64; rv:51.0) Gecko/20100101 Firefox/51.0
- Responsible for 40% of web login traffic
- Average of almost exactly 1 login request per unique username for sustained period of time. Legitimate traffic has 1.15-1.3 login requests per unique username.
- Traffic from 210 different countries with accept-language value always "en-US,en;q=0.5,"

# Conclusions & Takeaways

- Easy-to-use attack tools have made barriers to entry lower than ever before
- Sensitive data breaches will continue defenders must pursue this data for preventative measures. Assume all users' info is out there somewhere
- Attackers have a variety of ways to gather the infrastructure they need cloud hosting providers, botnets-for-rent, compromised machines, etc.
- Researching and fingerprinting the network characteristics of these tools is a very effective first step to detecting these attacks.
- Attackers migrate to the channel with the least friction defenders need visibility into their API traffic.

# Thank you!!!

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