Hershel: Single-Packet OS Fingerprinting

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- Introduction
- Background
- Building Hershel
- Simulations
- Internet Scan

Introduction

- The goal of OS fingerprinting is to determine OS of a remote host based on its network behavior
- Stack differentiation is possible due to:
 - Unclear language and lack of response standardization in IETF RFCs
 - No mandated behavior for malformed requests
 - Broken (non-compliant) implementations
- Network administrators and industry analysts have used OS fingerprinting as a tool
 - Identify and secure devices in own network
 - Market analysis of OS usage

Introduction

- Internet measurement studies are important to researchers
 - Detect vulnerabilities
 - Show deployment of new software and protocols
 - Scans have become progressively faster
 - 30 days, 1K pps [Heidemann 2008]
 - 24 hours, 24K pps [Leonard 2010]
 - 45 minutes, 1.4M pps [Durumeric 2013]
 - Large-scale measurement tools need to be fast, low overhead, and accurate
 - OS fingerprinting at large scale has not been explored before, which is our topic here



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Background

- Active OS fingerprinting typically requires open port
- Rooted in banner grabbing, which has many drawbacks
 - Protocol must be known
 - High overhead
 - Defeated by generic software (e.g., Apache)

HTTP/1.1 200 OK Cache-Control: private Content-Type: text/html; Server: Microsoft-IIS/7.5 X-Powered-By: ASP.NET Date: 15 Jun 2014 20:06:22 Connection: close Content-Length: 20559

- Admins can also remove/obfuscate OS-identifying strings
- Nmap is the current state of the art
 - Database of over 4K different OSes
 - Default 1032 probes per target, but no less than 38 in the least-verbose mode

Background

- Why not use Nmap?
 - Not a polite tool, generates complaints
 - Sends malformed probes, performs vertical port scans
 - Slow, infeasible for large scale
 - Packets easily blocked by IDS such as snort
 - Therefore, a more subtle approach is needed
 - p0f, RING, Snacktime are single-packet tools
 - Use header fields and timing of SYN-ACKs
 - Have small OS fingerprint databases (~20 different stacks)
 - Inaccurate when features change (e.g., packet loss)
- As a result, the issue of low-overhead and accurate fingerprinting remains open

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- Our aim is to build a single-packet tool that is robust to network and user modification
 - "Single-packet" means one outbound probe, but multiple responses from the remote OS are allowed
 - Assume remote host responds to TCP SYN
 - Specific port/protocol does not matter
 - A SYN probe provides minimal intrusiveness, along with non-malicious operation
- Suppose each OS j can be described by some fingerprint vector \boldsymbol{y}_j
 - Consists of two types of features network and user

Network features are SYN-ACK RTOs



• Examples:

OS	SYN-ACK RTO	Reset RTO	
Windows 7	3 6	12	
Mac OSX 10.3	2.92 6 12 24	30	
NetBSD 4.0	2.92 6 12 24	-	
Juniper Netscreen	1.67 2 2 2 2 2 2 2 2	2	
Huawei Embedded	0.7 1 1.2 3 4 5	-	

User features are values taken from packet header fields **Options** Do-notvector (RST present, RST

Time-to-(TCP) fragment ACK, RST Seq, RST Win) live field flag (IP) Receiver (IP) Maximum window segment (TCP) size (TCP) OS Win TTL DF **OPT MSS** RST Windows 7 8192 128 **MNWST** 1 1460 1,0,1,0 Mac OSX 10.3 33304 64 **MNWNNT** 1460 1 1,1,1,32768 NetBSD 4.0 **MNWNNTSNN** 32768 64 1 1460 0,-,-,-Juniper Netscreen 8192 64 0 Μ 1380 1,0,0,8192 Huawei Embedded 1536 255 0 Μ 768 0,-,-,never used

M = MSS, **N** = NOP, **W** = Window Scale, **S** = Selective ACK, **T** = Timestamp

before

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- Challenges
 - One-way delay (OWD) jitter (usually zero-mean)
 - Packet loss



- Challenges (cont'd)
 - User modification of default TCP/IP parameters (e.g., OS tuning software, fingerprint scrubbers, NAT, IDS)
 - Unlike OWD, these result in arbitrary value fluctuations
 - <u>Example</u>: Window size is more likely to jump from 8,192 to 65,535 than to 8,193
- Treating all features as volatile, an observed sample can match pretty much any OS

Fingerprint	Win	TTL	DF	OPT	MSS	RST	RTO
Observed	65535	64	1	MNW	1460	1,1,0,0	2.8 6.4
Windows 7	8192	128	1	MNWST	1460	1,0,1,0	3 6 12
Mac OSX	33304	64	1	MNWNNT	1460	1,1,1,32768	2.9 6 12 24 30

- Thus, any observation x can be viewed as a distortion of each original fingerprint y_j from underlying OS j
- Given a sample x, our goal is to determine the most probable y_i that could have produced it:

$$s(x) = \arg\max p(y_j|x) -$$

Which is equivalent to:

probability that observation x comes from OS j

$$s(x) = \arg \max p(x|y_j) p(y_j)$$

probability that y_j became distored into x

fraction of hosts running OS j

- To obtain these probabilities, we need a new model
 - Machine learning techniques don't work due to lossy features
- We develop a stochastic theory of single-packet fingerprinting to account for these random effects
 - See paper for details
- We then build a classifier called Hershel, which can additionally handle OSes with random feature vectors, and construct a database of 116 OSes
- Can distinguish not only between OS families (Windows, Linux, FreeBSD, embedded devices), but also patch levels (SP1 vs SP2)



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Simulations

- Emulate a FIFO queue between server and client
 - Run simulations to classify 2¹⁸ IP samples with random network/user modifications
 - Vary packet loss and user feature modification from 0 to 50%
- First, we perform comparison with Snacktime, which is the most accurate previous single-packet tool
 - Uses only RTO and Win/TTL (Pareto OWD, mean 0.5 sec)

		RTO only	accuracy	+Win/TTL	accuracy
Loss	Feature mod	Snacktime	Hershel	Snacktime	Hershel
0%	0%	12%	22%	58%	86%
3.8%	10%	10%	21%	44%	78%
10%	10%	7%	20%	33%	76%
50%	50%	0.8%	10%	2%	28%

Simulations

- Hershel's RTO classifier doubles Snacktime accuracy at low loss, triples at 10%, and improves an order of magnitude at 50% loss
 - However, Hershel works even better with new features

Hershel accuracy, using Pareto OWD (mean 0.5 sec)							
Loss	Feature mod	RTO Only	+Win/TTL	+DF	+TCP OPT	+MSS	+RST
0%	0%	22%	86%	89%	96%	99%	99.9%
3.8%	10%	21%	77%	79%	91%	94%	95%
10%	10%	20%	76%	77%	91%	94%	95%
50%	50%	10%	28%	35%	54%	57%	60%

 Numerous other scenarios and delay distributions omitted here, but shown in the paper



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- Port-80 SYN scan of the Internet
 - 2.1B IPs in 24 hours, 37.8M responses, 94% with at least one RTO
- **RTOs** Hosts Database 3 9.6M 27 9.0M 2 16 5 7.8M 23 5.0M 4 16 2.6M
- Extensive sanity verification of the dataset
 - Not enough room to show here, see the paper
- We see a lot more values for each header field than we have in our dataset
 - Emphasizes the importance of probabilistic matching
- Run Hershel on all hosts and obtain a non-zero matching probability on 37.4M devices

Internet Scan

Classification results – top 5 OSes and families

OS	Hosts	Family	Count
Linux 2.6 / 2.4	9.6 M	Linux	13.8 M
VxWorks Embedded	4.1 M	Embedded	13.5 M
Windows Server 2003 SP1 SP2	2.3 M	Windows	7.5 M
VxWorks 5.4 / Xerox Embedded	1.8 M	Other (Mac, BSD, Novell, etc)	2.3 M
Linux 2.6 / Debian / CentOS	1.1 M		

- Compared to previous application of Snacktime to this dataset [Leonard10], 9M more embedded devices
- Manual verification vs. Snacktime
 - We pick 1000 random hosts to compare classifications
 - When Hershel and Snacktime disagree, 97% of the time Hershel is correct, 1.8% Snacktime, and 1.2% neither

Thank you!

Questions?