CERN openlab Major Review Meeting

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FRN

Networking by HP

openlab

ProCurve

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Agenda



Flow analysis

- Statistical analysis methods
- What about the signatures?
- Results
- Time series mining preliminary stage
 - Introduction
 - Future plans
- Conclusions





Statistical Flow Based Analysis (1)

Sampled Flow Table





Statistical flow based analysis (2)

Moving Average approach

- Monitor a certain parameter and report hosts violating a given threshold, e.g:
 - Monitoring UDP connections within CERN from portable hosts
 - Measuring the number of different destination addresses contacted by a given source address
- Isolating anomalous traffic is challenging:
 - Very noisy traffic (many different protocols):
 - exclude flows from 'well-known' services (e.g. dhcp, ldap, AFS, etc)
 - the excluded 'well-known' flows should be analysed separately
 - Payload inspection might be necessary

Example of noise reduction



Top flow analysis



	#	SIG_NAME	PAYLOAD
Т	606	CINBAD BitTorrent 1	64313Ab164323Ab9 <mark>1</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
•	607	CINBAD BitTorrent 1	64313A6164323A69 <mark>6</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	608	CINBAD BitTorrent 1	64313A6164323A696432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	609	CINBAD BitTorrent 1	64313A6164323A69 <mark>6</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F OI OEI
	610	CINBAD BitTorrent 1	64313A6164323A696432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	611	CINBAD BitTorrent 1	64313A6164323A69 $_{6}$ 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F $_{5}$ in the
Ī	612	CINBAD BitTorrent 1	64313A6164323A696432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	613	CINBAD BitTorrent 2	64313A/264323A696432303A939F70DF1D5C16A4EB9A909EA844276F429
	614	CINBAD BitTorrent 2	64313A7264323A696432303A939F70DF1D5C16A4EB9A909EA844276F429
	615	CINBAD BitTorrent 1	54313Ab164323Ab96432303A900B3E8310D2A9C1C582173EAA1ADF6D3RVOrO
	616	CINBAD BitTorrent 1	64313A6164323A69 <mark>6</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	617	CINBAD BitTorrent 1	64313A6164323A69 <mark>6</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
ĺ	618	CINBAD BitTorrent 1	64313A6164323A69 <mark>6</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	619	CINBAD BitTorrent 1	64313A6164323A69 <mark>6</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
Ч	620	CINBAD BitTorrent 1	64313A6164323A69 <mark>6</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	621	CINBAD BitTorrent 2	64313A7264323A695432303AA7F3318F65FF036B3549023F311B6E2934E
- 1	622	CINBAD BitTorrent 2	64313A7264323A69 <mark>6</mark> 432303AA7F3318F65FF036B3549023F311B6E2934E/OOITS
	623	CINBAD BitTorrent 1	04313Ab164323Ab56432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
Ī	624	CINBAD BitTorrent 1	54313A6164323A69 <mark>5</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	625	CINBAD BitTorrent 1	54313A6164323A69 <mark>5</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	626	CINBAD BitTorrent 1	54313A6164323A69 <mark>5</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3 F}@M@
	627	CINBAD BitTorrent 1	54313A6164323A69 <mark>5</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	628	CINBAD BitTorrent 1	54313A6164323A69 <mark>5</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	629	CINBAD BitTorrent 1	64313A6164323A69 <mark>6</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F
	630	CINBAD BitTorrent 1	54313A6164323A69 <mark>5</mark> 432303A900B3E8310D2A9C1C582173EAA1ADF6D3F

How to use signatures



simple_filter

- CINBAD tool based on libpcap for filtering collected data
- Signatures can be written as pcap filter strings
- Snort
 - Rule based network traffic monitoring system
 - Rules for detecting numerous anomalies available
 - Does not work with sampled traffic out-of-the box



Snort customisation



- Porting to work with sampled data
 - workaround for truncated payloads
 - snort rules translated into stateless ones (if applicable)
- Results of the analysis logged to Oracle
- 7000+ rules with daily updates
 - cinbad rules
 - e.g. bittorent, zatto, QQ ...
 - http://www.emergingthreats.net
- Campus and Internet traffic analysis



Some signature analysis results

- ~45% alerts compared to the Snort analysis on the central firewall
 - we expect this ratio to increase when we add more switches
 - traffic rate at the firewall is high and Snort cannot process every packet
- internal and external traffic inspected
 - more p2p applications, instant messengers
 - two trojan likely infections
 - Password Stealer
 - Win32/Alureon.gen!J



Results are promising, but...

Presented methods require a significant amount of manual work

That is why we want to look at the Time Series Data Mining techniques...

Which hopefully will allow to increase the automatisation



Time series and why do we care?

- What are the Time Series?
 - A time series is a sequence of data points, measured at successive times
 - Time series are ubiquitous, more and more data is being measured and collected
- Examples:



Time series analysis



- Time series carry much information about the generating processes, for example:
 - What kind of system generated the data?
 - Temporal behaviour of the phenomenon
 - What is likely to happen in the future?
 - Relation with other data sources
 - Occurrence of the anomaly (novelty/surprise)
 - · · · · ·
- Unfortunately the information is usually buried deeply within the multivariate data





- Time series are the most natural representation for the state of the network:
 - Amount of traffic
 - SNMP counters
 - Number of connections
 - Number of distinct hosts contacted
 - Number of different TTL values
 - Number of BitTorrent signatures
 - Average packet roundtrip
 - . .



What can we find in time series data?







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Time series mining – plans

 Much ongoing research in the area of time series analysis, for example:

- Financial data analysis (we all want to be rich...)
- Bioinformatics, genomics (i.e. DNA analysis)
- Medicine (i.e. attempts to build brain-computer interface)
- • • •
- Network traffic analysis (i.e. detecting traffic volume anomalies)
- Look at the current state of the time series data mining
- Develop methods useful for the CINBAD project

Time series mining – plans



- Try to combine time series data mining with automatic signature extraction:
 - 1. An anomaly (novelty) is being detected by the mining procedure
 - 2. Feed the relevant traffic to the extraction engine
 - 3. Attempt to identify common patterns (longest common substrings) in the sampled packets
 - 4. Manual assessment of usefulness of the rules seems unavoidable

Conclusions



- Starting with the statistical flow analysis we have discovered typical patterns and signatures
- We obtained encouraging initial results from SNORT signature analysis
- Very good feedback received from HP ProCurve and its Architects' Forum
- We need to investigate the time series mining techniques