CSC 790



Fall 2014

Intrusions, Events, and Detection

- Intrusion is a set of actions that attempt to compromise the integrity, confidentiality, or availability of any resource on a computing platform
- Attacks manifest themselves in terms of events
 - Events can have different granularity (from packets to logs)
 - Each attack step/phase/action has some associated event
- Intrusion Detection Systems (IDS) monitor the system
 - Analyze information about system and network activities
 - Looks for evidence of malicious behavior
 - Goal for IDS is to analyze one or more event streams and identify manifestations of attacks

IDS Categories Based on Events

- IDS can be categorized based on the use of event streams
 - Anomaly detection or misuse detection
- Anomaly detection attempts to find abnormal behavior
 - Must first define *normal behavior* (based on history)
 - System attempts to identify patterns of activity that deviate
 - Recognize normal events, not an attack
- Misuse detection is the complement of anomaly detection
 - Have known attack descriptions (signatures)
 - Events stream are constantly matched against the signatures
 - Don't recognize normal, know attack events

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IDS Categories Based on Scope

- Can further categorize IDS based on scope: network or host
- Host implemented on a single machine
 - Only responsible for the host on which it resides
 - Maintains/obeserves audit files, system calls, etc...
 - For example tripwire
- Network implemented in a centralized or distributed fashion
 - Only responsible for the network
 - Measures traffic and/or scans packet data
 - For example Network Flight Recorder (NFR)
- Neither category is comprehensive... only applicable to certain types of attacks

Basic IDS Operation

Regardless of the category/type of IDS, they all do the following

- 1. Data Collection Collect system data
 - Network based Collect traffic using a sniffer software
 - Host based Process activity, memory usage, and system calls
- 2. Feature Selection Reduce data, create feature vectors
 - Network based Packet header information, payload, ...
 - Host based User name, login time and date, duration...
- 3. **Analysis** Determine if vector contains signature (misuse detection) or whether the data is anomalous (anomaly detection)
- 4. **Action** Alert and possibly automatically stop/minimize attack

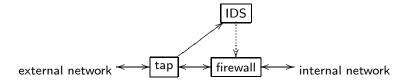
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Misuse Detection

- Known attack patterns create a library of attack signatures
 - Data that matches a library entry is considered an attack
 - An example signature based NIDS is **snort**
- Snort supports header and payload inspection of network traffic
 - User can define a rule and action that is applied to packets
 - Library is rule list applied to packets, there is a x match policy
 - Actions include: alert, log, pass, activate, dynamic
 - Rules are ordered based on the action...
 - Older versions of Snort had stateless inspection

Snort Placement

- Snort (and most IDS) is most commonly used in stealth mode
 - Use a network tap and send duplicate traffic to IDS Why use a tap?
 - Snort in-line is placed in the traffic stream...
- Integrating firewall and IDS, Instrusion Protection System (IPS)
 - Have the IDS send rules to the firewall



- Can also use the firewall to avoid IDS for legitimate traffic

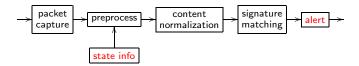
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Misuse Detection Advantages/Disadvantages

- Very low false alarm rate (only attacks match)
 - If an alert given, high probability it is an attack
- Only applicable to known attacks
 - Attack variations can defeat detection (neighboring attack)
 - Novel attacks are not detected
- Can be resource intensive since inspecting every attack
 - Stateful systems are very slow
 - Methods for rule optimization? High-speed IDS?

What happens when all traffic is encrypted?

Snort Operation



- Packet capture (libpcap... unfortunately)
- Preprocessing performs various operations
 - Flow detection, reassembly, and manage state
- Content normalization
 - Change content to common form (e.g. '%41' to 'A')
 - Otherwise think about all the signature variations
- Detection engine applies the set of rules to packet streams
 - Scan the payload for a certain signature (string match)
- Alert engine performs the matching rule action

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Snort Rules

- Snort rules have two parts, rule header and rule options
 - Header describes the action and packets to consider
 - Options provides more details packet attributes (if needed)
- For example, consider the following snort rule

```
alert tcp any any -> 10.1.1.0/24 222 (content:"|00 11 22 33|"; msg:"rpcd request")
```

- Rule header *alerts* when TCP traffic is observed originating from any network with any source port, destined for network 10.1.1.x to destination port 222
- Keyword content in option field requires the payload to be searched for the pattern

More Snort Rules

```
alert tcp any any -> any 80 (content:"abcde"; nocase; offset:5
depth:15; content:"fghi"; distance:3 within:9);
```

- Example above contains multiple contents
 - Additional keywords specify the content, case, and locations

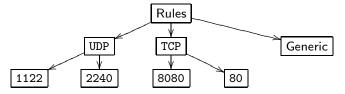
```
alert tcp any any -> any 80 (content:"mode=admin";
uricontent:"/newsscript.pl");
```

- Above example contains URI content
 - URI content is normalized and processed separately
- There are over 5000-ish Snort rules, will it ever decrease?

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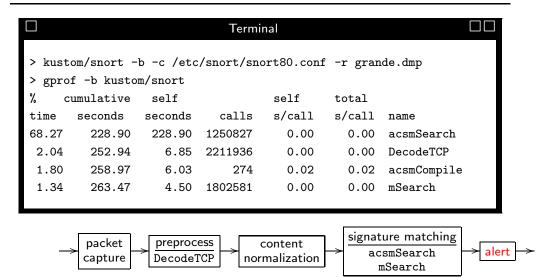
Rule Groups

- Do not necessarily want to compare a packet against every rule
- Snort divides rules into groups
 - Snort supports rules for TCP, UDP, IP, and ICMP protocols
 - Within each protocol rules are divided into groups
 - Each rule is placed in a group based on source/destination port



- When a packet arrives the content is compared against rules in
 - Port groups associated with the packet
 - Generic port group

Specifically, What is the Problem?



- Content searching is very time consuming
 - Others have reported from 40% to 75%

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Improving IDS Performance

- Improve the content search/match algorithms
 - Quickly search payloads for multiple signatures
 - Consider signature length when designing algorithms
 - Good, but perhaps the improvement is not enough
- Parallelize certain IDS components
 - Parallelization is possible at different granularities
 - Must consider the overhead of multi-threaded applications

Content Matching Algorithms

- Essential for any signature-based IDS
 - Algorithms were not necessarily motivated by IDS
 - It is just string searching
- Snort has incorporated various searching algorithms over time
 - Initially a simple brute force search, repeat for each signature
 - Replaced by Boyer-Moore, but still sequential
 - Snort 2.0 added Aho-Corasick and Wu-Manber (multi-pattern)
 - Snort 2.x added refinements to the existing algorithms
- Multi-pattern search has significantly increased performance

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Boyer-Moore Overview

- Used to quickly find a single pattern in a text
 - Compare to last character of pattern and shift tables
- ullet Assume pattern is length m and t is the text
 - Compare the last character of pattern to t_m
 - If not a match and t_m not in the pattern, look at t_{2m}
 - If t_m matches $\mathbf{4}^{th}$ character in pattern then look at t_{m+4}

a	n		е	x	a	m	p	1	е
a	m	р							
			a	m	p				
					a	m	р		

- "Longer the pattern the faster the search", possibly sublinear
- Unfortunately, IDS needs to search for thousands of patterns

Wu-Manber Overview

- Used to quickly find a group of patterns in text
 - Use shift table from Boyer-Moore, but with multi-patterns
 - Creates hash tables for pattern look-up
- Assume smallest pattern is length m and t is the text
 - 1. Call shift table on t_m , which return s
 - 2. If $s \neq 0$ then shift and go to step 1
 - 3. If s = 0 then (potential match) call hash
 - 4. If entries in hash table then sequentially match pattern(s)
- Best average case performance
 - "Short patterns inherently makes this approach less efficient"
 - Maximum shift m is the shortest pattern in the group

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Aho-Corasick Overview

- Linear time algorithm for multiple patterns
 - Based on an automata approach
 - Builds a FSM based on the characters in the patterns
 - Refinement of a **keyword tree** (*trie*)
- Best worst case performance (linear)
 - Requires more memory than other algorithms
 - Wu-Manber has a better average case due to skips

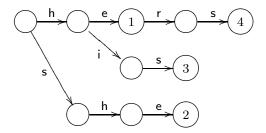
Keyword Tree

- Keyword tree (or a trie) for a set of patterns P is a rooted tree K
 such that
 - Each edge of K is labeled by a character
 - Any two edges out of a node have different labels
- The label of a node v is the concatenation of edge labels on the path from the root to v, and denote it by L(v)
 - For each $p \in P$ there is a node v with L(v) = p
 - The label L(v) of any leaf v equals some $p \in P$

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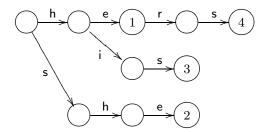
Example Keyword Tree Construction

• Keyword tree for $P = \{\text{he, she, his, hers}\}$



- Construction of the tree for $P = \{p_1, p_2, ..., p_k\}$, starting at the root, follow the path labeled by chars of p_i
 - If the path ends before p_i , continue it by adding new edges and nodes for the remaining characters of p_i
 - Store identifier i of p_i at the terminal node of the path
- Requires $O(|p_1| + |p_2| + ... + |p_k|) = O(n)$ time

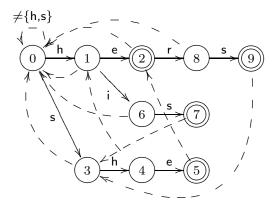
Keyword Tree Lookup



- Lookup of a string s: start at root, follow the path labeled by characters of s as long as possible
 - If path leads to node with identifier, s is keyword in the dictionary
 - If the path terminates before s, the string is not in the dictionary
- Takes O(|s|) time in the best case, very efficient lookup method
- Worst case? We can create an automaton to keep linear...

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String Matching



- Automaton consists of normal and failed (dashed) transitions
 - Double circles indicate matched strings
 - Node labels are the longest proper suffix
- Consider processing the string "ushers"

What does Snort Use?

- Snort selects an algorithm based on the number of rules
 - If there are fewer than 5 rules, then sequential Boyer-Moore
- If more than 5 rules, then use a multi-pattern algorithm
 - The default algorithm is ...
- Change /etc/snort.conf

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if(pattern found) validate

- Regardless of the algorithm, if pattern found then validate
 - Initial search uses Boyer-Moore, Wu-Manber, or Aho-Corasick
 - Search for the longest content string in the rule (good idea)
- Second phase attempts to validate the initial match
 - Snort rules may contain multiple keywords, including content

```
alert tcp any any -> any 80 (content:"abcde"; nocase; offset:5
depth:15; content:"fghi"; distance:3 within:9);
```

- mSearch verifies the remainder of the rule

Rule Groups and Small Signatures

- Assume there is a group that contains r > 5 rules
 - Furthermore, assume smallest signature in the group is 1-byte
- Snort will use Wu-Manber No Bad-Character Shift algorithm
 - Builds two hash tables, one-byte and multi-byte
 - Helps distribute the hash entries
- ullet When processing packet of length n
 - Will make n calls to the *one-byte* hash table
 - Will also make n-1 calls to the *multi-byte* hash table
 - As a result, small signatures are generally avoided

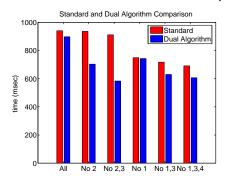
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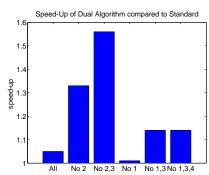
Proposed Dual Algorithm

- If a group has 1-byte and multi-byte patterns, then separate
 - Sub-group for 1-byte patterns and another for multi-byte
 - Original thought was to process groups in parallel
- Process the payload twice
 - For the 1-byte group, use B-M or W-Mnbcs
 - For the multi-byte group, use W-Mbcs
 - Yes, the payload is processed twice
- As a result, we hope...
 - Greatly reduce the number of multi-byte hash calls

Experimental Results

- Patrick has extensive results comparing dual to standard
 - 12 pages of results varying different parameters
- Used Snort 2.5.4 match component with traffic traces





- ullet As ASL increases, the dual algorithm should perform better
 - Greater the gap, the better the performance

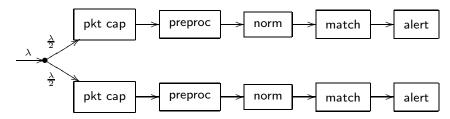
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Parallel IDS

- IDS in a high-speed environment is very difficult
 - Algorithm improvements may not be sufficient (sorry Patrick)
 - Faster hardware is only a temporary solution
 - Would like to parallelize IDS, parallel \neq distributed
- Given an array of processing elements, you can
 - Data parallel divide data across equal processing elements,
 reduces the arrival to any one element (improves throughput)
 - Function parallel- divide work across the processing elements, reduces the work at any one element (reduces latency)
- ullet Reducing latency is the objective o function parallel
 - Difficult to implement... where/what?

Parallel Taxonomy

- Three levels where parallelization is applicable
 - System, component, and sub-component
- System level parallelization
 - Just duplicate the entire IDS and split the traffic... easy

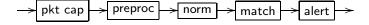


- Scalable and robust, but difficult to maintain state info
- This is data parallel, what is the function parallel version?

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Component Level

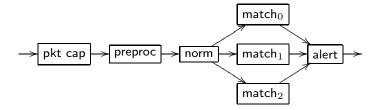
• Parallelize the different **components** of one IDS



- Isolate components, more like pipelining
- Preprocessing, matching, and alert are good possibilities
 - Processing can be divided into reassembly and port-scan
- Must consider the support and design required for threads
 - Data structures, semaphores, and context switching
 - Will the threads be serialized?

Sub-Component Level

• Parallelize (duplicate) components of one IDS



- For example, parallelize matching component (original idea)
- Duplicate all groups or different rule groups per thread
- Does not work with all components, consider reassembly
 - Same problem as system level, must maintain state information
- Are these examples of data or function parallel?

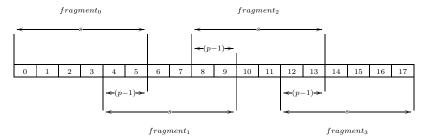
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Data Parallel String Matching

- Standard data parallel approach
 - Distribute packets across processing elements
 - Reduces the arrival to any one processing element
 - Requires load balancing
- An alternative is Divided Data Parallel (DDP)
 - Divide data of one packet across processing elements
 - Each processing element has a smaller part (fragment)
 - Shorter inputs are better, right?
 What are the problems associated with dividing the payload?

Dividing the Payload

- DDP divides the packet payload into fragments
 - Fragments given to processing elements, complete payload scanned
- Signature (malicious content) may span fragment
 - Single processor may not see complete signature
 - Must overlap fragments to prevent false negatives

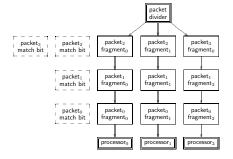


- Overlap dependent on largest signature, p, (example above p=3)
 - Overlap is (p-1) with leftmost fragment

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Match This

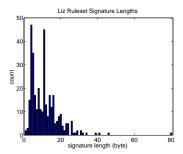
- Only considering the initial match
 - Once a match made, can skip remaining packet
 - Original DDP does not



- Once a match is made, set match bit for the packet
 - Match bit indicates do not process remaining fragments
- Allows processors to skip fragments

DDP Performance

- Compare different parallel match performance
 - Data parallel and forms of distributed data parallel
- Created a multi-threaded match for Snort 2.6
 - Traffic traces from a university
 - Rules from government agency (345 total web-rules)

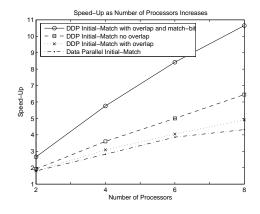


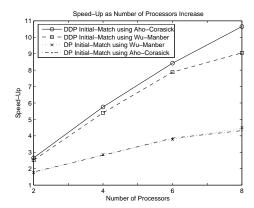
Recorded processing time and speed-up

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Some Results

• DDP always performed better than traditional data parallel





• Speedup using match bit was the best

Types of Parallelism

	Data Parallel	Function Parallel
System	replicate IDS	?
Component	?	isolate IDS components
Sub-Component	duplicate match	distribute rules across match

- Of course multiple levels/types of parallelization are possible
- Must consider overhead associated with parallelization
 - Queues for moving data from one thread to another
 - Inter-thread communication and context switching
 - Semaphores and potential serialization

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Problems with Signatures

- IDS discussed thus far relies on signature matching
 - Look for suspicious packet headers and/or payloads
- Great for finding known threats
 - Not so great for zero-day threats...

Detection Using Machine Learning

- It may be possible to use machine learning to find malware
 - Assumption is malware looks different than non-malware
- Supervised learning could be used
 - Train IDS to identify normal/legit conditions
 - Should be able to identify never-before-seen conditions

Of course the problem is?

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Anomaly Detection

- The normal behavior of the system is modeled
 - Patterns that deviate from normal are attacks
 - Premise is malicious activity is a subset of anomalous activity
 - Applicable to network attacks, such as DoS
- Most systems use a form of change point monitoring (CPM)
 - Determine if the observed data is statistically homogeneous, and if not, determine when the change happened
 - Collect statistics about system under normal usage (history)
 - Once statistics change, then it is/was possible attack

$$r_n = \alpha \times r_{n-1} + (1 - \alpha) \times r_n$$

What to Measure

- Number of unique IP addresses
 - Large number of unique IP addresses indicate DDoS attack
- Number of TCP SYN packets
 - 90% of the DoS attacks use TCP, measure the number of SYN requests to a certain server
- Compare the number of TCP SYN and SYN/ACK packets
 - Distributed Reflector DoS (DRDoS), use routers as reflectors to send SYN/ACK packets
- Compare the number of TCP SYN and FIN packets
 - Should be the *relative* same number of SYN and FIN packets

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Anomaly Detection Advantages/Disadvantages

- Possibility of very high false alarm rate
 - What is normal? What is a significant change?
 - Usage change over time, can anomaly detection differentiate?
 - Would consider a *flash crowd* and attack
- Does not depend on specific attack signatures
 - Attack variations can be detected and possibly novel attacks
- Not as resource intensive since measuring aggregates
 - May still have difficulty in high-speed environments
 - Can one IDS see a DDoS?

IDS Performance Metrics

- True Positives (TP) is number of correct malware classifications
 - True Positive Rate (TPR) is $\frac{TP}{TP+FN}$
- True Negatives (TN) is number of correct non-malware classifications
- False Positives (FP) is number of incorrect classifications of non-malware as malware
 - False Positive Rate (FPR), or false alarm rate, $\frac{FP}{FP+TP}$
- False Negatives (FN) is number of incorrect classifications of malware as non-malware

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A Few More Performance Metrics

- Recall is the fraction of correct instances among all instances that actually are positive (malware)
 - Calculated as $\frac{TP}{TP+FN}$ How can you get perfect recall?
- Precision is the fraction of correct instances (malware) that algorithm believes are positive (malware)
 - Calculated as $\frac{TP}{TP+FP}$ Is perfect precision always best?

Base Rate Fallacy

- Base rate fallacy (base rate bias), is an error in thinking
 - When given related general, generic information (base rate info)
 and specific information, we tend to focus on the specific
- Suppose you develop a vampire test that falsely indicates a vampire in 5% of the cases; however, it never fails to detect a real vampire
 - Assume in a town, 1/1000 of people are vampires
 - Suppose you stop a random person and the test indicates vampire What is the probability the person is a vampire?

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Bayes' Theorem

- ullet Need to use Bayes' theorem, which can find the probability of B given A is also true
 - For us, we want the probability of the person is a vampire given the test indicated vampire
- The equation for our problem is

$$p(vampire|test_{vampire}) = \frac{p(test_{vampire}|vampire)p(vampire)}{p(test_{vampire})}$$

where

- p(vampire) = 0.001

- p(normal) = 0.999

 $-p(test_{vampire}|vampire) = 1$

 $- p(test_{vampire}|normal) = 0.05$

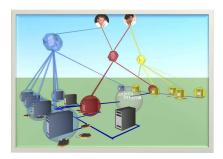
- $p(test_{vampire}) = p(test_{vampire}|vampire)p(vampire) + p(test_{vampire}|normal)p(normal) = 0.05095$
- Therefore $p(vampire|test_{vampire}) = 0.019627$
- A more intuitive explanation, on average for every 1000 people tested
 - 1 person is a vampire, and 100% certain that for that person there is a true positive test result, so there is 1 true positive test result
 - 999 people are not vampires, and among those people there are 5% false positives, so there are 49.95 false positive results
 - Therefore therefore the probability that one of the people among the 1+49.95=50.95 positive test results really is a vampire is $\frac{1}{50.95}\approx 0.019627$

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Problem with IDS

- Very difficult to have good recall and precision
- Malware typically is a small percentage of traffic
 - Looking for one packet in a million (or billion)?
- Insufficiently sensitive, IDS will miss the malware (low recall)
- Too sensitive, IDS will have too many alerts (low precision)

Swarm Intelligence (PNNL Project**)**



- Defense using swarm intelligence and simple software agents
 - Swarm of digital ants, each finds evidence per machine
 - Group of findings will indicate the actual problem
 - Movement based on pheromone, swarm an infected machine
- Better (faster and more robust) than having an IDS per machine?

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Swarm Design

- Actually a hierarchy of agents, lower two levels...
 - Sentinel resident per machine receives information per agent
 - Sensors wander the network, there are several types of Sensors each looking for a certain type of evidence
- General operation is as follows
 - When a Sensor arrives to a computer, it performs a simple test
 - Test results given to Sentinel, determine if system is *healthy*
 - If results are helpful, then reward Sensor which attracts others

What type of IDS is this? (Note, "failed" is not an answer)

What are the advantages?

New Directions

- Combining multiple IDS types
 - Combine signature and anomaly at host and network
 - Detect new attacks with low false alarm rate
 - Intrusion Detection Alert Correlation considers multiple event streams from different IDS
- Specification-based intrusion detection
 - Describe attacks in more general terms (unlike snort)
- Attack prediction
 - Using on-line statistics is it possible to predict an attack?
- Better integrated IDS and firewall system