Effective and Efficient Malware Detection at the End Host

Clemens KOLBITSCH, Paolo MILANI COMPARETTI, Engin KIRDA, Christopher KRUEGEL, Xiaoyong ZHOU, XiaoFeng WANG

ck@iseclab.org

Secure Systems Lab [TU Vienna, Institute Eurecom Sophia Antipolis, UC Santa Barbara]
Indiana University at Bloomington



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Why do we propose yet another malware detection scheme (yamds)?

- Binary signature based detection inherently ineffective
 - We all know the problems...
 - Arms-race, pretty much a lost battle
- Network based approaches evadable
 - Systems scan for communication artifacts
 - Encryption / blending thwart detection



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Why do we propose yet another malware detection scheme (yamds)?

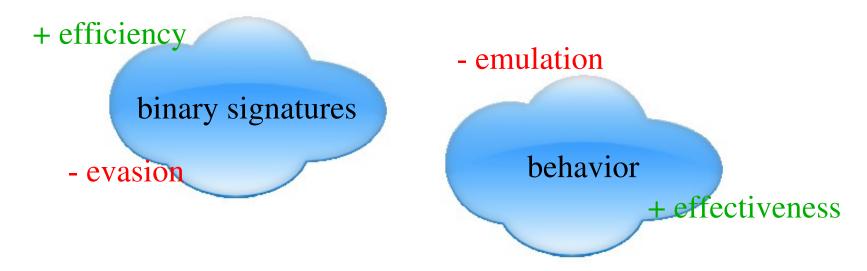
- Don't rely on artifacts of malware instances
 - Instead focus on generic patterns
- Proposed solution:
 - Detection based on malware's behavior
 - Behavior is hard to obfuscate
 - Behavior is hard to randomize
 - Behavior is often stable across various malware version



- Behavior-based detection received some attention over last couple of years
- Despite promising detection results, binary signatures remain the method of choice



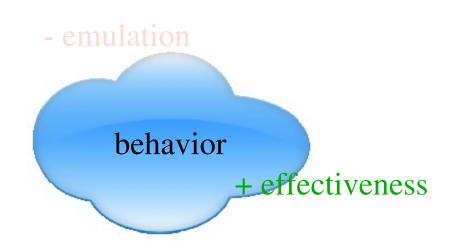
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Outline

- Motivation
- Detecting Behavior
 - Motivating example (Agent)
- Matching Behavior Graphs
- Extracting Behavior Graphs
- Evaluation



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Detecting Behavior



- Characteristic malware behavior
 - Manifest on system (i.e., survive reboot)
 - (Over-) write system executables, dlls, files
 - Create registry entries (autorun)
 - Register as Windows (startup) service
 - Conceal from being detected
 - Restart under some stealthy name (e.g., svchost.exe)
 - Inject into legitimate processes
 - Replicate
 - Send eMails ('check out this picture I found: pic.jpg.exe')
 - Copy to Samba shares, USB drives, etc.
 - Scan and exploit services on LAN or WAN



System Overview

- Detection based on execution characteristics
 - Execute malware in full system emulator (Anubis)
 - Monitor interaction with the operating system
 - Perform detailed (taint-) analysis
 - Generate detection graphs
 - Describe sequence of required system calls leading to security relevant system activity
 - Include dependencies to related, previous calls (using taint dependencies)
- Detect described behavior on end host
 - Log system call activity of unknown executable
 - Match against behavior graph

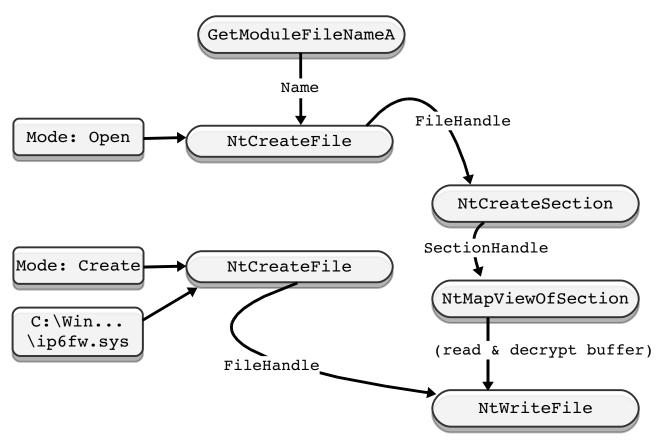


Developer Perspective

- Example: Agent (trojan horse)
- As part of its system manifestation, it
 - Reads content from binary image
 - Decrypts binary content
 - Proprietary decryption routine
 - · Simple, XOR based algorithm
 - Stores binary in system file (c:\Wind...\drivers\ip6fw.sys)
 - Later, restarts IPv6 firewall
 - Turns itself into a system service

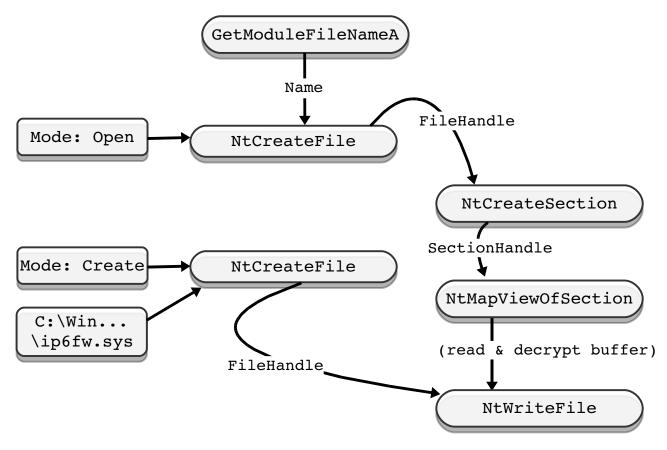


Taint-Trace Perspective



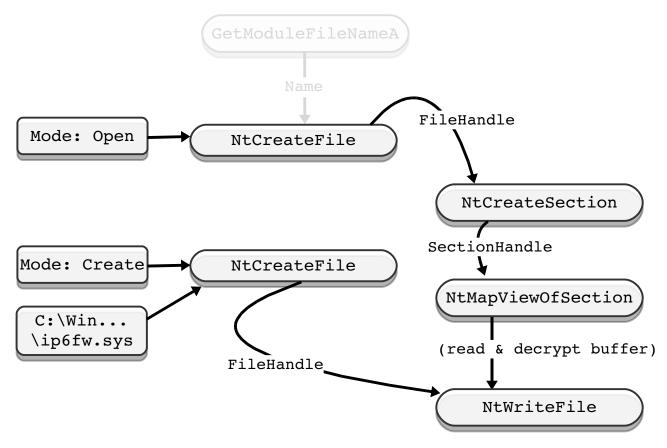


System Perspective



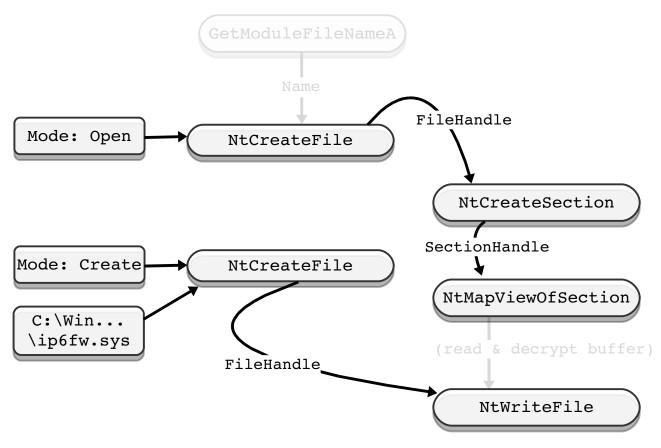


System Perspective



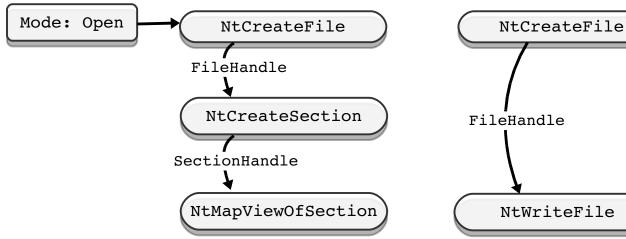


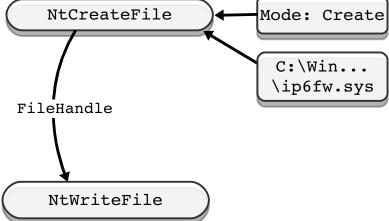
System Perspective





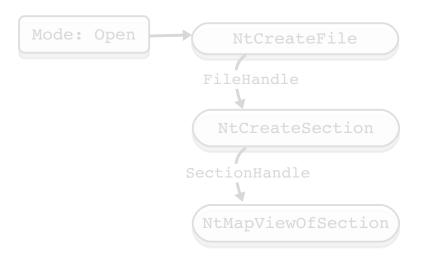
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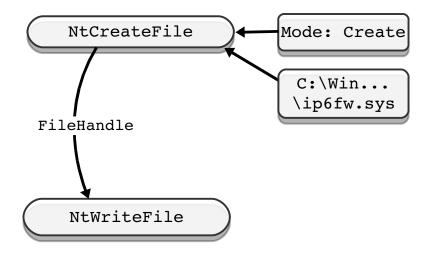






System Perspective







- Detection based on execution characteristics
 - Works well as long as we can see all types of dependencies between system calls
 - Handle dependencies
 - Insufficient for detection
 - Behavior graphs break into trivial subgraphs
 - Data dependencies
 - Convenient for behavior graph generation
 - Necessary for behavior detection



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Matching Behavior Graphs



- Maintaining dependencies using taint propagation
 - Performance overhead: Extended emulation engine
 - Memory overhead: Shadow memory
 - Not applicable to production systems / end hosts
- Maintaining dependencies without taint propagation
 - Handle dependencies
 - Direct value propagation
 - System provided identifiers
 - File, section, process, thread handles
 - Registry keys
 - Socket identifiers
 - Must be constant between call invocations



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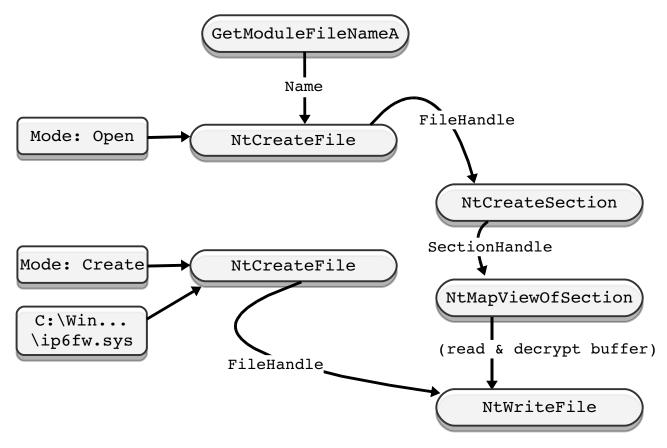
- Maintaining dependencies without taint propagation
 - Data dependencies
 - Arbitrary data (& control) dependency between system calls
 - Might modify values between system calls

Our proposal: Anticipate precise call arguments

- Use recorded execution semantics
- Extract data propagation/manipulation formulas
- Emulate taint dependency between system call A and B
 - Log outgoing parameters of call A
 - Use as input to propagation formula
 - Predicted incoming parameters for system call B
 - Compare predicted and monitored input parameters
 - Assume dependency between A and B if prediction holds

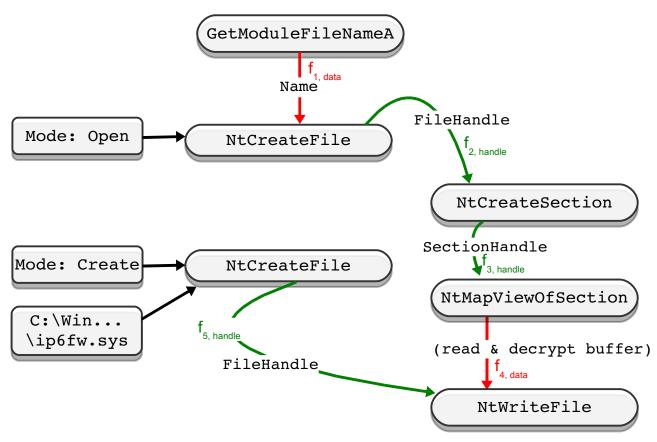


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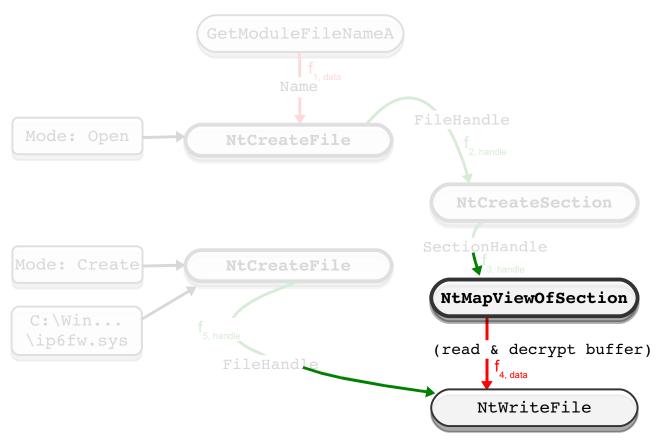


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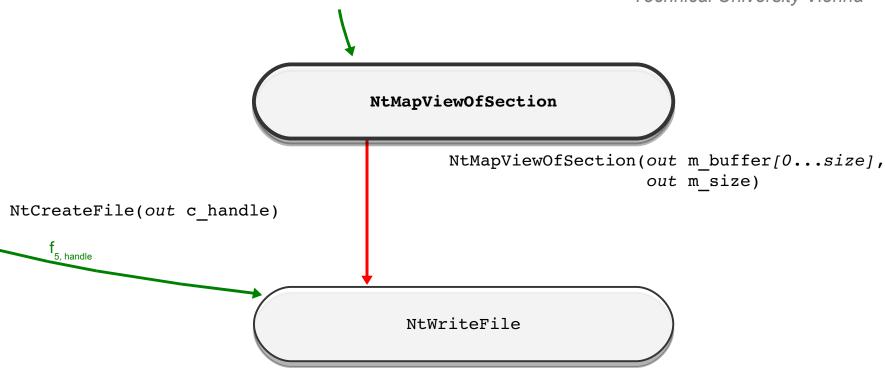


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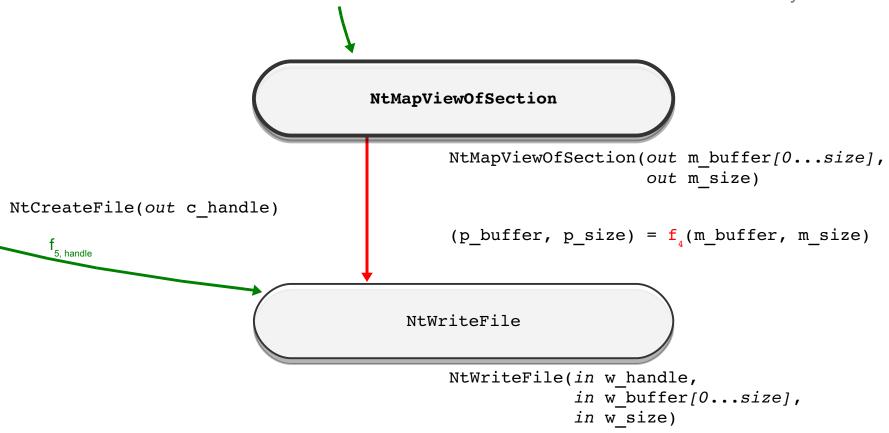


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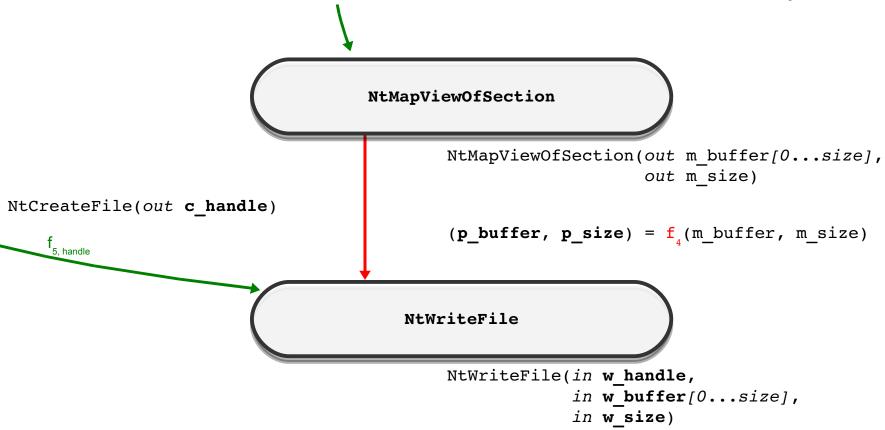


System Perspective





System Perspective





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Extracting Behavior Graphs



- Analyze executable in Anubis sandbox
 - Obtain instruction level log
 - Defeats packers
 - Obtain program flow log
 - Obtain memory access log
 - Generate precise taint propagation trees
 - Data/control dependencies
 - Instructions that access/generate tainted data
 - Link system calls consuming data (sinks) with all taint generating calls (sources)

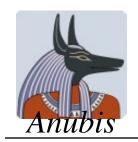


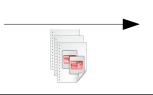






- Scan logs for security relevant behavior
 - Provided with a list of interesting system calls
- Extract graphs matching behavior
 - Include triggering system call X
 - Link in system calls providing tainted data to X
 - Analyze dependencies:
 - Label edges with handle dependencies
 - Call slicer for all data dependencies





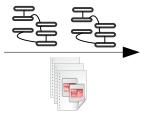






- Find encoding formula for each data dependency
- Binary program slicing
 - Resolve def-use chains
 - Starting at selected call invocation
 - Iterate backwards (using program flow logs)
 - Aided by taint information and memory access logs
 - Optional:
 - Symbolic execution to simplify encoding function
 - Embed into dynamically loadable library (dll)
 - Label graph edges with appropriate function (dll)





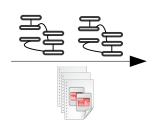




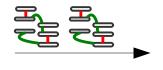


- Resolving def-use chains
 - Three possible sources
 - 1) Previous system call invocation
 - Replaced with stub
 - Provides input values to slice (i.e., recorded, outgoing system call parameters)
 - 2) Immediate values
 - Implicitly encoded in binary slice (e.g., push \$0x3)





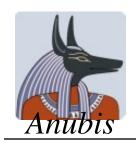


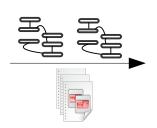




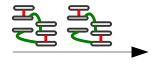


- Resolving def-use chains
 - Three possible sources
 - 3) Preinitialized data segments
 - BSS section
 - Constants
 - Static strings
 - Two-sided approach:
 - Use static values from Anubis analysis
 - Dynamically inspect running process





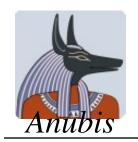


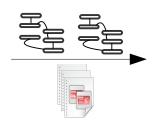




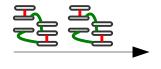


- Fully automated process
 - Analyze binary
 - Generate behavior graph(s)
 - Extract propagation formulas
 - Verify graph on binary
 - Run binary & scanner on real host
 - Verify behavior graph matches (only) on intended executable













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Evaluation



Evaluation

- Effectiveness of behavior graphs
 - Applicable to polymorphic variants of a malware sample?
 - General enough for whole malware families?
- Efficiency of behavior graph matching
 - Overhead through system call logging
 - Additional system load through dependency verification



Effectiveness

- Six current threats / threat families
- Identified using AV (binary) signature
- Encountered 0 false positives

Name	Туре	Samples	Vari AV	ants Our	Samples detected	Eff.
Allaple	Exploit-based worm	50	2	1	50	1.00
Beagle	Mass-mailing worm	50	20	14	46	0.92
Mydoom	Mass-mailing worm	50	32	12	47	0.94
Mytob	Mass-mailing worm	50	20	2	41	0.82
Netsky	Mass-mailing worm	50	22	12	46	0.92
Agent	Trojan horse	50	6	3	49	0.98
Total		300	102	44	279	0.93



Effectiveness

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• Experiment:

Can the system detect malware instances never seen by the graph generator?

Name	Samples	AV va New	ariants Known	Samples detected	Eff.
Allaple	50	0	50	45	0.90
Beagle	50	24	26	30	0.60
Mydoom	50	24	26	36	0.72
Mytob	50	46	4	5	0.10
Netsky	13	8	5	7	0.54
Agent	50	6	44	45	0.90
Total	263	108	155	168	0.63



Effectiveness

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Netsky	13	8	5	7	0.54	
Agent	50	6	44	45	0.90	
Total	263	108	155	168	0.23 0.92	



Efficiency

- I-O bound activity
 - Compressing, archiving
- CPU bound computation
 - Compilation, rendering

Test	Baseline	Log		Full scanner		
7-zip (benchmark)	114 sec	117 sec	2.3 %	118 sec	2.4 %	
7-zip (compress)	318 sec	328 sec	3.1 %	333 sec	4.7 %	
7-zip (archive)	213 sec	225 sec	6.2 %	231 sec	8.4 %	
IE (rendering)	0.41 pages/s	0.39 pages/s	4.4 %	0.39 pages/s	4.4 %	
VC++ (compile)	104 sec	117 sec	12.2 %	146 sec	39.8 %	



Summary

- Behavior can be detected
 - Monitor from system perspective
 - Match against behavior graphs
 - Link graph nodes through argument dependencies
- Handle dependencies
 - Vital for checking
 - BUT not specific enough for doing detection
- Data dependencies
 - Anticipate future call arguments
 - Efficient replacement for taint dependencies
 - Provided through slicing malware semantics



Summary

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Evaluation

- Behavior detection is fast enough for end hosts
- Approach intrinsically robust against polymorphism and metamorphism
- To some extent, behavior graphs are usable across malware variants



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Thanks for your attention!

