

Self-Mutating Malware Detection¹

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Currently mainly based on **signature matching**

- Search for peculiar byte sequences
- What to do if malware is able to change its own byte stream?

Polymorphic worms



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In case of a worm which spreads with different byte stream a possible approach is the one advocated by Hamsa IDS [Li et al., 2006]

- Analyze several worm instances
- Use statistics to find a signature with good False neg. vs. False pos.

Experiments show that this approach can be quite effective for network worms that have an invariant part (typically, the exploit). (BTW, Hamsa suffers *poisoning* attacks)

Metamorphic viruses



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What if malware can mutate itself arbitrarily?

[Chess and White, 2000]

Theoretical studies demonstrated that perfect detection of a self-mutating malware is an undecidable problem

However, real world mutation has to be performed by simple transformations. Current self-mutating prototypes use code obfuscation techniques.

Code obfuscation and self-mutation



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- Code obfuscation is a **semantic-preserving program transformation** that can be used to make a program harder to understand
- Self-mutation is a particular form of code obfuscation, which is performed automatically by the code on itself
- Self-mutation is applied during malicious code replication to generate completely new different instances

- Most of the pieces of malware are not standalone
- Their code is *inserted* in a **host program**
- The (guest) malicious code is *activated* by executing the host program
- Malware has to be detected intertwined with *goodware*

Signature matching becomes useless

Common techniques adopted for malicious code insertion:

- Cavity insertion
- Jump tables manipulation
- Data segment expansion

The malicious code is seamless integrated into the host code

Substitutions



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Instruction substitution

A sequence of instructions is substituted by a semantically equivalent sequence

Variable substitution

Registers or memory address are substituted with available candidates

Instruction permutation



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Data flow independent instructions can be permuted arbitrarily.

Example

```
1: a = b * c;  
2: d = b + e;  
3: c = b & c;
```

can be executed in any order in which the use of c precedes its new definition (1 always before 3)

Control flow alteration



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- Introduction of useless conditional and unconditional branches
- Direct jumps and function calls can be translated into indirect ones
- Destination addresses computed by inserted instructions

Garbage insertion



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- Dead code
- Alive code that does useless things
 - Killed definitions
 - Neutral operations ($a = (b + 0) * 1$)

Cavity insertion



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- Generally, executables contain several portions that are not used to store data or code (they can be introduced by compilers to align code and data structure).
- These **cavities** can be used to insert small pieces of the malicious code that can be forced to be executed with minor modifications of the host code.

Jump table manipulation



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- High-level control flow transfer constructs (e.g., switch) are implemented using **jump tables**.
- Entries of such tables can be modified in order to get the execution to be redirected anywhere.

Data-segment expansion



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- Some of the host's segments can be expanded.
- Not all segments are suited for expansions because that would imply a relocation of most of the code.
- The one storing uninitialized data (bss) is the more appropriate as the expansion allows for the insertion of malicious code without requiring further modification of the host code.

Challenges for the detection



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Conventional detection techniques:

- *Pattern matching* fails since fragmentation and mutation make hard to find signature patterns
- *Emulation* would require a complete tracing of analyzed programs as the entry point of the guest is not known; moreover every execution should be traced until the malicious payload is not executed
- *Heuristics* based on ad-hoc predictable and observable alterations of executables become useless when insertion is performed producing almost no alteration on any of the static properties of the original binary

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Our approach



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- ① Use optimization techniques to obtain **a normal form** of the malware [Perriot, 2003]
- ② Use program analysis abstract models to perform comparison [Cristodorescu et al, 2005]

- Analysis of the transformations adopted to implement self-mutation and experimental observations highlighted some weakness:
 - Transformations led to the generation of useless computations
 - Most transformations are invertible
- Different instances of the same malware can be viewed as under-optimized version of the archetype; the archetype is consequently the normal form of the malicious code

Code normalization

A program is transformed into a canonical form which is simpler in term of structure or syntax while preserving the original semantic and that is more suitable for comparison

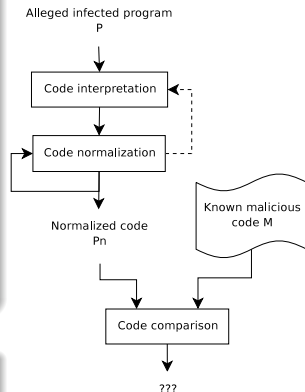
Devised strategy

Code interpretation and normalization

- Given a piece of code P which represents (or contains) an instance of a self-mutating malware we automatically revert all the mutations performed on it
- P is consequently reduced into a form, P_N , which is pretty close to its ideal *archetype* M and which can be recognized more easily

Code comparison

- Detection is performed by looking for known abstract patterns into the transformed program P_N



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Code normalization

Some details



- Executable code is disassembled and translated into an intermediate form to explicit the semantic of each machine instruction
- *Control-flow analysis* and *data-flow analysis* are performed on the code to collect information that will be used by the next step

Machine instruction	Interpreted instruction
pop %eax	r10 = [r11] r11 = r11 + 4
lea %edi, [%ebp]	r06 = r12
dec %ebx	tmp = r08 r08 = r08 - 1 NF = r08@[31:31] ZF = [r08 = 0?1:0] CF = (~ (tmp@[31:31]))

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Code normalization



- Identify all the instructions that do not contribute to the computation (**dead and unreachable code elimination**)
- Rewrite and simplify algebraic expressions in order to statically evaluate most of their sub-expressions (**algebraic simplification**)
- Propagate values computed by intermediate instructions to the appropriate use sites (**expressions propagation**)
- Analyze and try to evaluate control-flow transition conditions to identify tautologies and to rearrange the control to reduce the number of flow transitions (**control-flow normalization**)
- Analyze indirect control flow transitions to discover the smallest set of valid targets and the paths originating (**indirections resolution**)

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An example



```
xor    %ebx,%ebx
mov     $0x1000400c,%eax
mov     %eax,0x10004014
add     %ebx,%eax
test    %ebx,%ebx
jne     <T>
push    %ebx
mov     $0x0,%ebx
```

T:

```
jmp     *%eax
leave
ret
nop
```

T:

```
r11 := r11 ^ r11
r10 := 0x10004014
[0x1000400c] := r10
r10 := r10 + r11
tmp = r11 - r11
ZF = [tmp = 0?1:0]
jump (ZF = 1) T
[r15] := r11
r15 := r15 - 4
r11 := 0
```

```
jump r10
r15 := r16
r16 := m[r15]
r15 := r15 + 4
return
```

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```
r11 := 0
r10 := 0x10004014
[0x1000400c] := 0x10004014
r10 := 0x10004014 + 0
tmp = 0 - 0
ZF = [0 = 0?1:0]
jump (ZF = 1) T
[r15] := 0
r15 := r15 - 4
r11 := 0
```

T:

```
jump 0x10004014 + 0
r15 := r16
r16 := m[r16]
r15 := r15 + 4
return
```

T:

```
jump 0x10004014
r15 := r16
r16 := m[r16]
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```

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Given the normalized program we need to answer the question:

“Is the program P_N hosting the malware M ?”

- We cannot expect to find a perfect matching of M in P_N even if most of the transformations have been reverted
- The code comparator must be able to cope with some impurities left by normalization (we observed that these impurities are always local to basic blocks)
- The normalized control-flow of the malware is constant

Code comparison

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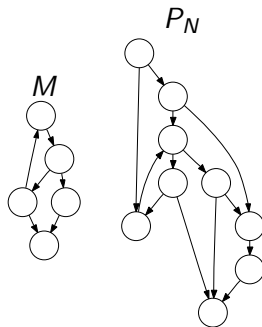
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- P_N is represented through its *interprocedural-control flow graph* (ICFG) and M through its control-flow graph
- The malicious code detection can be formulated as a *subgraph isomorphism decision problem*:
“given two graphs G_1 and G_2 , is G_1 isomorphic to a subgraph of G_2 ?”
(G_1 is M and G_2 is P_N)

Code comparison

Some details

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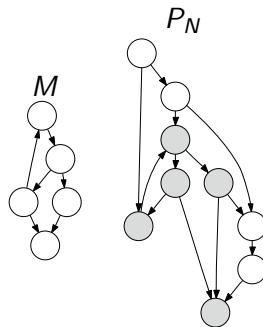
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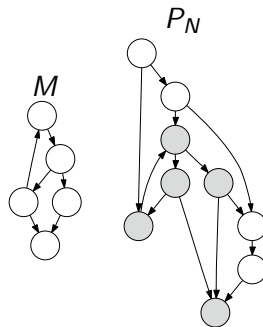
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“given two graphs G_1 and G_2 , is G_1 isomorphic to a subgraph of G_2 ?”
(G_1 is M and G_2 is P_N)
- The graphs are augmented with labels to achieve the necessary trade-off between precision and abstraction (to handle possible impurities)



Instruction classes

Integer arithmetic
Float arithmetic
Logic
Comparison
Function call

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Prototype implementation



- The code normalizer is built on top of BOOMERANG, an open-source decompiler:
 - Translate machine code into the intermediate form through a recursive disassembler
 - Performs data-flow analysis on the intermediate form
 - Performs the normalization steps previously described (some of the transformation have been extended to suit our needs)
 - Able to solve known patterns of indirection
- The prototype receives an executable file and emits its normalized $ICFG_{P_N}$
- The $ICFG_{P_N}$ of the normalized program and the CFG_M of the searched malware are then fed to the VF-LIB2 library which is used to identify possible matches
- In case of match the comparison routine returns the set of $ICFG_{P_N}$ nodes that match the ones of the CFG_M

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Two independent tests were performed:

- ① Evaluation of code normalization effectiveness:
 - Several instances of the same self-mutating malicious code (the virus METAPHOR) were collected and normalized (57% code reduction)
 - The normalized control-flow graphs were all isomorphic, they were not before
- ② Evaluation of code comparison precision:
 - Different executables were collected and their ICFGs were built
 - Each procedure CFG was used to simulate malicious code and searched inside the ICFGs
 - The results of the subgraph isomorphism detection procedure were compared with the results obtained through code fingerprinting
 - A random set of alleged false-positives and false-negatives were selected and inspected by hand

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Type	#		# nodes	Average load	Worst detection
Executables	572				
Functions (# nodes > 5)	25145		(~)	time (secs.)	time (secs.)
Unique functions (# nodes > 5)	15429		100	0.00	0.00
Positive results	#	%	1000	0.09	0.00
Equivalent code	35	70	5000	1.40	0.05
Equivalent code			10000	5.15	0.14
(negligible differences)	9	18	15000	11.50	0.32
Different code			20000	28.38	0.72
(small number of nodes)	3	6	25000	40.07	0.95
Unknown	1	2	50000	215.10	5.85
Bug	2	4			
Negative results	#	%			
Different code	50	100			

Summary



- We proposed a general strategy, based on static analysis, that can be used to pragmatically fight malicious codes that adopt self-mutation to circumvent detectors
- We developed a prototype tool and used it to show that a malware that suffers a cycle of mutations in most cases can be brought back to a canonical shape that is shared among all instances
- We showed that augmented control-flow graphs are well suited to describe a peculiar piece of code and that reliable code identification can be formulated as a subgraph isomorphism decision problem
- Although the subgraph isomorphism is a NP-complete problem, our particular instance seems to be tractable (the graphs we are dealing with are very sparse)

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- Extend our prototype to perform normalization on real world executables and increase the effectiveness of normalization by extending the quality of the analysis performed
- Evaluate algorithms for partial subgraph isomorphism matching and the benefits they could give in our context
- Perform more exhaustive experiments using new malicious code
- Investigate attacks and countermeasures to defeat static analysis