

Analyzing and Detecting Flash-based Malware

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Christian Wre!(j"\$ADG=(,,#\$....who?

- PhD candidate at the Institute of System Security
 - Established in April 2016 in Brunswick, Germany by Prof. Konrad Rieck
 - Previously at the University of Göttingen
- TU Braunschweig
 - Oldest "institute of technology" in Germany (founded in 1745)
 - 40-year-long history of computer science







Malware

Malicious software (Malware)

- Lasting problem of computer security
- Omnipresence of Trojans, Bots, Adware, ...
- Increase of targeted attacks using Malware
- Flash-based malware
 - Malware targeting the Adobe Flash platform
 - Drive-by-Downloads, malicious redirects, exploits, ...



Adobe Flash

• Flash is dead!

- Deployed on 500 million devices across different platforms
- Used on 25% of the top 1,000 Alexa web sites

Dynamic and multimedia content on web pages

- Advertisement, video streaming, gaming, ...
- 20 years of deployment
- Powerful scripting language: ActionScript



Adobe Flash Vulnerabilities

- Increasing number of CVEs
 - About 550 different vulnerabilities in total
 - Until 2015: 167 new vulnerabilities (80% code execution)

Disclaimer! Effective August 2015





Attack Vectors and Scenarios

1. Structural Exploits against the Flash Player

- Vulnerabilities in the file format parser
- 2. Malicious ActionScript code
 - Launching or preparing exploits (Obfuscation, heap-spraying, ...)
- 3. Environment fingerprinting
 - Selecting targets based on interpreter or OS information

Concrete attacks may fall into more than one of these categories



Obfuscation

- Staged execution
 - Dynamic code-loading in form of another animation loadMovie (ActionScript 2), Loader object (ActionScript 3)
 - Layered encryption/ polymorphism *Runtime-packers (secureSWF, DoSWF)*
 - Exploit legacy code
- Source-code Obfuscation
 - Variable substitution, string assembly, dead code, etc.
- Probing the execution environment
 - Triggering a malware's payload on specific systems only



Probing the environment

- Information about the execution environment
 - System.capabilities (ActionScript 2)
 - flash.system.Capabilities (ActionScript 3)
- LadyBoyle malware exploiting CVE-2015-323

```
switch (this.version) {
    case "win 11,5,502,146": break;
    case "win 11,5,502,135": break;
    case "win 11,5,502,110": break;
    case "win 11,4,402,287": break;
    case "win 11,4,402,278": break;
    case "win 11,4,402,265": break;
    default:
        return this.empty();
```

}

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- **Comprehensive analysis of Flash animations** Support for all versions of ActionScript and Adobe Flash platforms
 - Structural Analysis (static)
 - Guided code-execution (dynamic)
- Learning-based detection of Flash-based malware
 - Detects <u>90–95</u>% of malicious Flash files at <u>0.1% and 1.0%</u> FPs
 - Significantly outperforms related approaches
 - Best learning-based detector for Flash-based Malware
 - No need for manually constructed detection rules



Structural Analysis

- Flash animations are composed out of "tags"
 - Containers to store code, animation specs and data (audio, video, images, fonts, etc.)
 - Future versions may extend on the number of tags
 - Possible occurring nested (*DefineShape, ...*)
- Offering a huge attack surface
 - Many exploits rely on a specific (sequences of) tag
 - Memory corruption exploits such as CVE-2007-0071



Structure Reports

- Exemplary report for a LadyBoyle sample using CVE-2015-323
 - 69 FileAttributes
 - 77 Metadata
 - 9 SetBackgroundColor
 - 2 DefineShape
 - 39 DefineSprite
 - 26 PlaceObject2
 - 86 DefineSceneAndFrameLabelData
 - 43 FrameLabel
 - 87 DefineBinaryData // Payload
 - 87 DefineBinaryData // Payload
 - 82 DoABC // ActionScript 3
 - 76 SymbolClass
 - 1 ShowFrame
- More compact: 69 77 9 2 [39 26] 86 43 87 87 82 76 1

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Analyzing Code

Dynamic code analysis

- Single execution "as-is" is not sufficient
- Covering all execution paths is not feasible
- Heuristics needed!
- Previous approaches
 - Determine which paths to execute based on external input ("Exploring Multiple Execution Paths for Malware Analysis", Moser et al.)
 - Symbolic execution of code ("A Symbolic Execution Framework for JavaScript", Saxena et al.)
 - Multi-execution of branches along the intended path ("Rozzle: De-cloaking Internet Malware", Kolbitsch et al.)



Guided Code-Execution

- **Gordon:** Guide the interpreter towards indicative code regions
 - Branches that contains indicative functions (*loadMovie*, *loadBytes*, *ByteArray*, ...)
 - Paths with many instructions
- Two-step procedure
 - Determine Control-flow statically
 - Use CFG to guide the analyzer
 - Multiple runs possible
 - Force Execution at environment sensitive conditions





Execution Reports

• Excerpt of a report for a sample using CVE-2015-323

```
pushString
                      "fla"
R1 973:
        pushString
R1 975:
                     "sh.uti"
                     "fla" + "sh.uti"
R1 977:
        add
R1 978:
        pushString
                     "ls.Bvt"
                     "flash.uti" + "ls.Byt"
R1 980:
        add
R1 981:
        pushString
                     "eArray"
                     "flash.utils.Byt" + "eArray"
R1 983:
        add
        callProperty [ns:flash.utils] getDefinitionByName 1
R1 984:
R1 >
        Looking for definition of
R1 >
             [ns:flash.utils] ByteArray
        Getting definition for
R1 >
R1 >
             [ns:flash.utils] ByteArray
        getLex: [ns:] Class
R1 987:
```

For automatic processing reports meta data is omitted

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Learning-based Detection

- Preprocessing of reports
 - Structure reports: cf. compact representation

69 77 9 2 **[** 39 26 **]** 86 43 87 87 82 76 1

- Execution reports: Instruction names and parameters only
 - Parameters are replaced with their respective type

pushString	STR	
add	STR + STR	
callProperty	getDefinitionByName	NUM
getLex	ID	

- **Embedding:** *n*-gram models of structure and execution reports
- Learning: Classification using Support Vector Machines (SVMs)



- Used to embed string data into vector space

- Generalization of the Bag-of-Words model
- String represented as bag of features

Different variations:

- Words
- Byte n-grams
- Word n-grams



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- Different variations:
 - Words
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 - Word 3 grams





Embedding of *n*-grams in Vectors

Assign each n-gram a dimension in the vector

$$\Phi: x \longrightarrow (\Phi_s(x))_{s \in S}$$

- Embeddings
 - Counting $\Phi_s = \# n$ -gram S in X

• Binary occurrence
$$\Phi_s = \begin{cases} 1 & n \text{-gram } S \text{ in } X \\ 0 & \sim \end{cases}$$



n-Grams of Tag Identifiers

• Example for the structure report

4-grams of tag identifiers

69	77	9	2	[39	26]	86	43	87	87	82	76	1
69	77 77	9 9 9	2 2 2 2	[[[39 39 39	26 26]	86 86	43 43 43	87 87 87 87	87 87 87 87	82 82 82	76 76	1



n-Grams of Instructions and Parameters

- Example for the execution report
 - 4-grams of instructions/ params

```
pushString STR add STR
STR add STR +
  add STR + STR
  add STR + STR
STR + STR callProperty
  + STR callProperty getDefinitionByName
```

...

- No need for manually constructing detection rules
 - Implicit representation of instruction counts, call frequencies, etc.



Learning the Classifier

- Support Vector Machines (SVMs)
 - Modern supervised learning algorithm for classification
 - Invented by Vapnik (1963) and kernelized by Boser (1992)
 - Well-known for its effectiveness, efficiency and robustness
- Important concepts
 - Hyperplane with maximum margin
 - Regularization by softening the hyperplane
 - Let's you compensate mistakes





Evaluation

Datasets

- 26,600 Flash Animations collected over 12 weeks
- 1,923 malicious and 24,671 benign samples
- How well are we able to detect Flash-based malware?
 - Comparison to the state-of-the-art methods
 - Is Gordon applicable in a continuous setting?
- What's all the fuss about two different analyses?
 - Wouldn't be one of them enough?



Experimental Setting

Temporal split of the data

- Weeks 1-6 for training, weeks 7-9 for validation, and the remainder, weeks 10-12 for testing
- All test data has been collected after training

Related approaches

- FlashDetect (T. van Overveldt et al, RAID 2012)
 - Adjusted to 1% false-positives
 - Not supported version have been excluded (version 8 and below)
- Virus scanners listed at VirusTotal



Comparative Evaluation

- Gordon is on a par with tradition approaches
 - No manual effort needed, though





Analyzing and Detecting Flash-based Malware

Combined Detection Performance

- Gordon benefits from two orthogonal analyses
 - Individual representations only detect 60–65% at 0.1% FPs





Temporal Evaluation

- Applied to 12 consecutive weeks: 80–99% detection rate
 - Clear trend towards Gordon's optimal performance





Summary

- Comprehensive Analysis of Flash-based malware
 - Structural analysis
 - Guided code-execution
 - Directed analysis of indicative code regions
- Effective Detection of a large variaty of Flash-based malware
 - High detection rate: <u>90–95%</u> of malicious samples
 - Low false-positve rates
 - Best learning-based detector for Flash-based Malware
 - Can be used to bootstrap traditional methods



Analyzing and Detecting Flash-based Malware

Thank you. Questions?

