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Title: Detecting Code Injection Attacks in Internet Explorer

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Detecting Code Injection Attacks in Internet Explorer

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Abstract

Code injection vulnerabilities are a major threat to Internet security. The ability for a malicious website to install malware on a host using these vulnerabilities without its knowledge is particularly menacing. In this paper, we approach this problem from a new perspective by constructing a Markov chain graph from the system calls Internet Explorer executes and then modeling this graph over time. We apply a Gaussian process change-point algorithm to detect code injection attacks. To show the efficacy of this approach, we collect a novel dataset of system call traces of 6 code injection attacks using 3 distinct exploits against the Internet Explorer browser. Our algorithm was able to detect all of the code injection attacks with a limited number of false positives.

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Introduction

- The Internet is a dangerous place
- Malicious downloads, Drive-by downloads
- Collect system call traces of Internet Explorer
- Use change-point detection algorithms to find changes in the application's behavior over time
- Real-world example of concept drift (with ground truth)



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Code Injection Attacks

- Just visiting a website can lead to infection
- Javascript is used to host the shellcode
- Exploits typically use vulnerabilities in Internet Explorer's associated plug-ins
- The plug-ins are called within javascript, and the exploit is initiated



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Exploits

- Aurora
 - Uses an invalid pointer reference within Internet Explorer
 - Creates a backdoor connection to a command and control server
- Dllloader
 - IE calls DLLs with relative paths
 - Icon is presented to the user, user downloads dll, IE then calls the dll
- Unsafe Scripting
 - Takes advantage of a buffer overflow exploit in the ActiveX controls of IE



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Data Collection

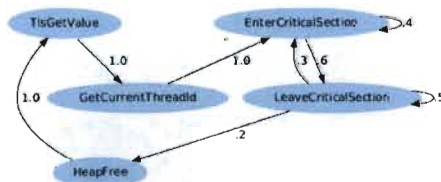
- **Xen Virtual Machine**
 - Provides a safe environment to perform exploits
- **Metasploit**
 - Many exploits are already written
 - Easy to use interface
- **StraceNT**
 - Collects the system calls made by the process

Data Collection: Part II

- Collect normal IE usage for 10 or 15 minutes
- Initiate exploits, then let IE run for 5 or 10 minutes (20 minutes total)
- This process results in files with ~1-1.4 million system calls
- Each time bin consists of 10,000 system calls

Data Representation

- **Markov chain**
 - Vertices are the system calls performed by the process
 - Transition probabilities (edges) are estimated by the traces



Change-Point Detection Algorithm

- Goal: find $p(r_t|x_{1:t})$

$$p(r_t|x_{1:t}) = \frac{p(r_t, x_{1:t})}{\sum_{r_t} p(r_t, x_{1:t})}$$

- Use recursive message passing scheme to find $p(r_t, x_{1:t})$

$$p(r_t, x_{1:t}) = \sum_{r_{t-1}} p(x_t|x_{t-1}) p(r_{t-1}, x_{1:t-1})$$

- $p(x_t|x_{t-1})$ is given by a Gaussian process time-series model

Change-Point Detection Algorithm: Part II

- $p(x_t | x_{t-1}) = N(m_t, v_t)$ where

$$m_t = K_s^T (K + \sigma^2 I)^{-1} x$$

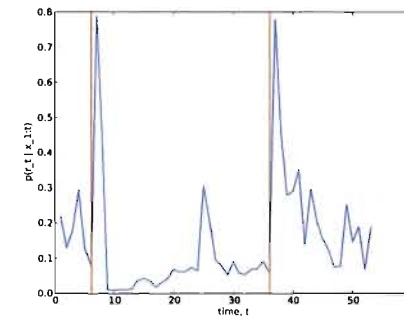
$$v_t = K_{ss} - K_s^T (K + \sigma^2 I)^{-1} K_s$$

- We use the Gaussian kernel

$$K(x, y) = \sigma^2 e^{-\frac{1}{2\lambda^2} \sum_{i,j} (x_{ij} - y_{ij})^2}$$

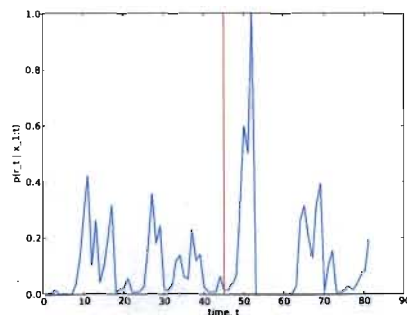
Synthetic Results

- Benign and malicious instruction traces
- 5,000 instructions per bin
- Inserted malicious instruction bins in two places within the benign trace



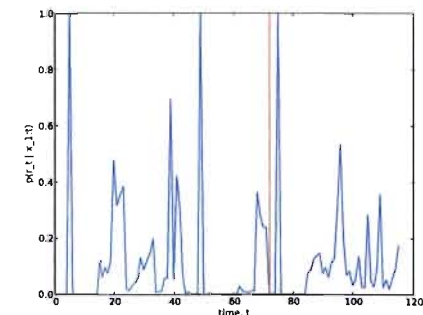
Real-World Results: Aurora

- Bins of 10,000 system calls
- Ran for 10 minutes, exploit, ran another 10 minutes



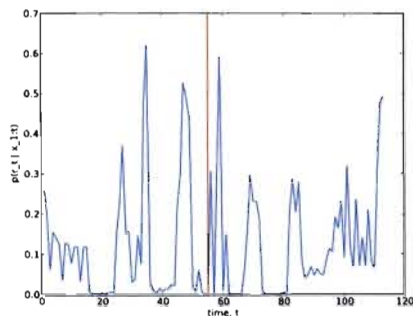
Real-World Results: dllloader

- Bins of 10,000 system calls
- Ran for 15 minutes, exploit, ran another 5 minutes



Real-World Results: unsafescripting

- Bins of 10,000 system calls
- Ran for 10 minutes, exploit, ran another 10 minutes



Conclusions

- We are able to detect when attacks happen
- The number of false positives will be prohibitive in a production setting
- Ideas to lower false positive rate:
 - Use instruction traces instead of system call traces
 - More advanced kernels in the GPTS model
 - Using a simpler change-point model