# MUNI C4E

# Towards a Data-Driven Recommender System for Handling Ransomware and Similar Incidents

IEEE International Conference on Intelligence and Security Informatics

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November 2, 2021



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Introduction

### Section 1

### Introduction

#### Introduction

# **Motivation**

#### **Ransomware and similar threat**

- The rising complexity and variety of cyberattacks complicate incident handling.
- IDS and secure perimeter are bypassed by **social engineering attacks**, e.g., phishing.
- The malware further **spreads in the network**, exploiting surrounding computers.
- There is little chance of mitigating the spread of infection.

#### **Incident handling**

- Rapid incident response prevents spread of infection and reduces attack impact.
- Effective triage and prioritization of threats and incidents are of utmost importance.
- The behavior of malware can be **anticipated** to some extent.
- Social engineering is difficult to detect we depend on user reports.

# Approach

#### Anticipating the behavior of the malware

- A typical malware uses a few attack vectors and spreads in close proximity first.
- The lateral movement of an attacker can be observed, traced, and even projected.
- However, that requires detailed knowledge of the local environment and collaboration with users and administrators (complicated in large networks).

#### Recommender system for incident handling

- The incident handlers would appreciate any piece of information that would guide them through the network and pinpoint nodes that are immediately threatened.
- The key question of an incident handler is: if this device is infected, which other devices can be infected or threatened?
- Recommendation of the list of devices at risk can be automated.

**Requirements and Design** 

### Section 2

### **Requirements and Design**

**Requirements and Design** 

# Requirements

#### **Data Collection**

The system should collect or be able to access the data on the network and hosts in it. The required data items include a list of network segments, network topology, hosts' services, software, and vulnerabilities, and contacts on administrators.

#### **Rich Information**

The data shall hold as much information as possible. It is often unfeasible to have all the data available; the system shall work even with incomplete data.

#### Interconnection of Heterogeneous Data

The data shall be stored in a way that allows for the interconnection of heterogeneous data. The data should be accessible at any time and updated at least once a day.

# Architecture

The proposed recommender system consists of three parts:

### Data Collection

- Set of tools to monitor the network and the hosts<sup>1</sup>
- Uses NetFlow network traffic monitoring and Nmap network scanner
- Timed execution to update the data regularly

### Data Representation and Storage

- Data model to represent links in heterogeneous data
- Database to store the data

#### Recommendation Subsystem

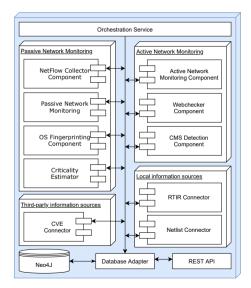
- Calculates similarity and distance of network nodes
- Prioritizes the hosts by their risk score

<sup>&</sup>lt;sup>1</sup>Husák, M., Laštovička, M., & Tovarňák, D. (2021, August). System for Continuous Collection of Contextual Information for Network Security Management and Incident Handling. In The 16th International Conference on Availability, Reliability and Security.

# **Data Collection**

For the whole **network**, the system collects:

- List of the active hosts in the network,
- Network topology (via Nmap from several observation points),
- List of network segments with:
  - location (e.g., department, building, server room),
  - purpose (e.g., workstations, servers, IP pool of VPN, ...),
  - contact on responsible person (e.g., local IT administrator),
- History of security incidents:
  - including a list of involved devices,
  - available via RTIR or similar system.



### **Data Collection**

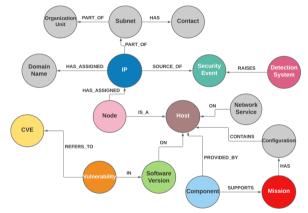
For each **host** in the network, the system collects the following:

- Fingerprint of the operating system (via NetFlow or Nmap),
- List of open ports and services, including the name and version of the underlying software (via NetFlow and NBAR2 signatures or Nmap),
- For web server: name and version of Content Management System (via WhatWeb),
- Name and version of a web browser used on the system (via NetFlow),
- Name of the antivirus software on the system and its latest update (via NetFlow),
- List of vulnerabilities (via vulnerability scanner or estimated from fingerprints),
- Location, purpose, contacts on administrators or main users (if available).

**Requirements and Design** 

### **Data Representation and Storage**

Data are stored in Neo4j graph database and structured using the CRUSOE data model<sup>2</sup>



<sup>&</sup>lt;sup>2</sup> Komárková, J., Husák, M., Laštovička, M., & Tovarňák, D. (2018, August). CRUSOE: Data model for cyber situational awareness. In Proceedings of the 13th International Conference on Availability, Reliability and Security.

### **Recommendation Subsystem**

The recommendation subsystem is a service that:

- 1. receives an identifier of a host in the network (e.g., IP address) on the input,
- 2. looks up the host in the database,
- 3. looks up devices in the proximity of the host,
- 4. calculates their similarity to the host on the input,
- 5. prioritizes the found hosts by their risk score,
- 6. returns a sorted list of similar devices in close proximity as the output.

More details in the following section.

**Calculations and Recommendations** 

### Section 3

### **Calculations and Recommendations**

### **General Idea**

The recommendations are based on the **proximity** and **similarity** of the hosts in the network to the host on the input; similar hosts in close proximity are prioritized.

#### Proximity

Two hosts can be close to each other in physical and logical network topology, e.g., in the same room or in the same IP range. Alternatively, the two machines can be close to each other if they are controlled by the same users or administrators.

#### Similarity

The similarity is based on the similarity in software equipment, role, profile, or shared history of the two hosts. Similarity in software equipment is a prevalent feature due to the fact that the attackers typically exploit certain services or software.

## **Risk Score**

Formally, the hosts are sorted by their risk score (R) calculated as a quotient of the similarity (S) and distance (D) of the two hosts:

$$R = \frac{S}{D} = \frac{s_1 * s_2 * \dots s_n}{\min\{d_1, d_2, \dots, d_n\}}$$

In practice, it would be advantageous to assign weights to *S* and *D* or their elements.
The weights could be extracted from real-world scenarios and tuned in operations.
The weights are left for future work.

### **Distance Calculation**

When the ransomware is reported, we do not know yet how it spreads:

- Malware spreading over the network will typically spread in the same subnet.
- Malware infecting files and drives will spread to machines used by the same user.
- Malware in email attachments will spread in the same department.

### Distance

The distance between the two hosts is the minimal value of various distance metrics.

- Breadth-first graph traversal is used to find hosts with minimal distance in any of the distance metrics (in the implementation using graph database).
- The distance in logical network topology is the length of the path in the graph.
- Arbitrary distance metrics can be added as needed: physical distance, location in the same room, similarity of IP addresses, ...

# **Similarity Calculation**

The malware often uses exploits of specific software or services.

- If malware uses SSH brute-forcing, then Linux machines with SSH servers are at risk.
- We do not know the exact software equipment and may only assume similarities.
  - If the malware exploits Outlook email client, we shall look up all Windows machines.

#### Similarity

The similarity is calculated as a product of partial similarities  $s_1 * s_2 * ... s_n$ . Each partial similarity is a value in the range < 0, 1 >.

- The similarity of software equipment and network services are the main features.
- CPE strings represent pieces of software running on a host.

# **Similarity Calculation**

#### **Examples of similarity metrics**

- CPE string similarity
  - CPE is an array of strings (vendor, product, version, ...) weighted 0.5, 0.25, 0.125, ...
  - The metric is the sum of weights of equal strings from the left to the first difference.
- CPE categories
  - If there is always 1 main CPE for each category, then simple CPE similarity is used.
  - Categories can be OS, browser, antivirus, ...
- Service similarity
  - If a service is provided by one host but not the other, a default value of 0.8 is used.
  - CPE strings are compared if both hosts provide the service.
- Similarities in vulnerabilities or past incidents
  - The number of common CVEs divided by the number of unique CVEs in the network.
  - The number of common past incidents divided by the total number of past incidents.

**Example of Using the System** 

### Section 4

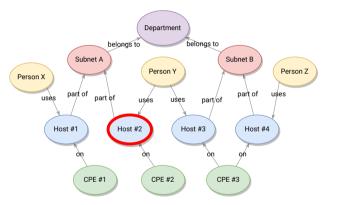
### Example of Using the System

# **Distance calculation**

Host #2 is reported to be infected, it's distance to other hosts is:

- 2 to Host #1 (same subnet)
- 2 to Host #3 (same user)
- 4 to Host #4 (subnets belonging to the same department)

Host #4 is too far – the calculation of similarity between Host #2 and Hosts #1 and #3 follows.



# Similarity calculation

The OS fingerprint of Host #2 is compared to fingerprints of Hosts #1 and #3

Host #1 and Host #2 share the same vendor, product, and version the similarity is 0.5 + 0.25 + 0.125 = 0.875

Host #2 and Host #3 share only the vendor – their similarity is 0.5

CPE format	cpe:part:vendor:product:version:update:edition:language
Weights	0.5, 0.25, 0.125, 0.0625, 0.03125, 0.03125
CPE #1	cpe:2.3:o:microsoft:windows_7:-:sp2:*:*
CPE #2	cpe:2.3:o:microsoft:windows_7:-:sp1:*:*
CPE #3	cpe:2.3:o:microsoft:windows_10:-:*:*:*

The list of similar devices in close proximity for Host #2 goes as follows:

- Host #1, risk score is 0.875/2 = 0.4375
- Host #3, risk score 0.5/2 = 0.25

Conclusion

### Section 5

### Conclusion

# Conclusion

Summary

- We proposed a design of a **recommended system for incident handling**.
- If a compromise of a host in the network is reported, the system instantly provides a prioritized list of other hosts at risk for rapid response.
- The system considers various ways of attack propagation or lateral movement.

Future work

- This paper is merely the first step in the future research.
- The system is under development and will be evaluated in operations.
- The weights of the metrics will be inferred from past incidents.
- Integration with other incident handling tools will follow.

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