



# **ICIMP 2023**

The Eighteenth International Conference on Internet Monitoring and Protection

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**ICIMP 2023 Editors**

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# ICIMP 2023

## Forward

The Eighteenth International Conference on Internet Monitoring and Protection (ICIMP 2023), held between June 26<sup>th</sup> and June 30<sup>th</sup>, 2023, continued a series of events targeting security, performance, vulnerabilities in Internet, as well as disaster prevention and recovery.

The design, implementation and deployment of large distributed systems are subject to conflicting or missing requirements leading to visible and/or hidden vulnerabilities. Vulnerability specification patterns and vulnerability assessment tools are used for discovering, predicting and/or bypassing known vulnerabilities.

Vulnerability self-assessment software tools have been developed to capture and report critical vulnerabilities. Some of vulnerabilities are fixed via patches, others are simply reported, while others are self-fixed by the system itself. Despite the advances in the last years, protocol vulnerabilities, domain-specific vulnerabilities and detection of critical vulnerabilities rely on the art and experience of the operators; sometimes this is fruit of hazard discovery and difficult to be reproduced and repaired.

System diagnosis represents a series of pre-deployment or post-deployment activities to identify feature interactions, service interactions, behavior that is not captured by the specifications, or abnormal behavior with respect to system specification. As systems grow in complexity, the need for reliable testing and diagnosis grows accordingly. The design of complex systems has been facilitated by CAD/CAE tools. Unfortunately, test engineering tools have not kept pace with design tools, and test engineers are having difficulty developing reliable procedures to satisfy the test requirements of modern systems. Therefore, rather than maintaining a single candidate system diagnosis, or a small set of possible diagnoses, anticipative and proactive mechanisms have been developed and experimented with. In dealing with system diagnosis data overload is a generic and tremendously difficult problem that has only grown. Cognitive system diagnosis methods have been proposed to cope with volume and complexity.

Attacks against private and public networks have had a significant spread in the last years. With simple or sophisticated behavior, the attacks tend to damage user confidence, cause huge privacy violations and enormous economic losses.

The CYBER-FRAUD track focuses on specific aspects related to attacks and counterattacks, public information, privacy, and safety on cyber-attacks information. It also targets secure mechanisms to record, retrieve, share, interpret, prevent and post-analyze cyber-crime attacks.

Current practice for engineering carrier grade IP networks suggests n-redundancy schema. From the operational perspective, complications are involved with multiple n-box PoP. It is not guaranteed that this n-redundancy provides the desired 99.999% uptime. Two complementary solutions promote (i) high availability, which enables network-wide protection by providing fast recovery from faults that may occur in any part of the network, and (ii) non-stop routing. Theory on robustness stays behind the attempts for improving system reliability with regard to emergency services and containing the damage through disaster prevention, diagnosis and recovery.

We take here the opportunity to warmly thank all the members of the ICIMP 2023 technical program committee, as well as all the reviewers. The creation of such a high-quality conference program would not have been possible without their involvement. We also kindly thank all the authors who dedicated much of their time and effort to contribute to ICIMP 2023. We truly believe that, thanks to all these

efforts, the final conference program consisted of top-quality contributions. We also thank the members of the ICIMP 2023 organizing committee for their help in handling the logistics of this event.

We hope that ICIMP 2023 was a successful international forum for the exchange of ideas and results between academia and industry and for the promotion of progress in the field of Internet monitoring and protection.

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Performance Evaluation of an Authentication Scheme for IoT Networks  
*Chi Ho Lau and Sammy Chan*

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# Performance Evaluation of an Authentication Scheme for IoT Networks

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**Abstract**—With the wide application of Internet of Things (IoT), the security of IoT systems has attracted significant research interest. In particular, since many devices can be connected to an IoT network, they face an authentication issue, which may be exploited by attackers to break into them. Recently, we have proposed an authentication scheme based on the blockchain technology to authenticate IoT devices before they can join an IoT network. In this paper, we develop a stochastic model to evaluate the efficacy of the authentication scheme. Numerical results indicate that the proposed scheme significantly increases the probability that a device stays in a healthy state.

**Index Terms**—IoT, blockchain, authentication.

## I. INTRODUCTION

The Internet of Things (IoT) has brought tremendous improvement to our quality of life. Machines, devices, sensors can connect and communicate with each other via networks. Together with existing Internet standards, IoT devices, such as wireless cameras and innumerable sensors, provide services for information transfer, analytics, and applications [1].

The security of IoT has already attracted the attention of many researchers [2]. It is not a trivial issue that may only affect an individual household or company. For example, the pitfall of the IoT network may be exploited by the attackers to break into the smart city infrastructure [3]. Among all security concerns of IoT, the authentication of IoT devices is a well-known issue. Since many IoT devices are welcomed to join the network, how to make sure all of them are legitimate is an important issue. A straightforward solution is to scrutinize all IoT devices. One may suggest a system like a vehicle registration system or mobile phone registration system to keep a registry of the owners of these IoT devices. However, the number of IoT devices is far more than the number of vehicles or mobile phones. This solution may cost tremendous administration overhead and discourage users to use IoT devices.

In [4], we proposed a solution to solve the problem of authenticating devices in IoT networks. This solution utilizes blockchain technology to store the identity information of authenticated devices. Based on its characteristics, blockchain is used to create the digital identification of IoT devices and authenticate IoT devices. A private blockchain is generated in each IoT network to isolate the network from outside access. It

highly increases the security level of the IoT network and the integrity of information collected by IoT devices. This paper evaluates the performance of our proposed authentication scheme by considering a stochastic threat model. Numerical results indicate that the proposed scheme significantly increases the probability that a device stays in a healthy state.

## II. OVERVIEW OF AUTHENTICATION SCHEME

The distributed property of blockchain makes malicious tampering or forgery difficult. Also, every transaction within the network is signed by a private key that provides strong protection against forgery. Therefore, blockchain technology is suitable to store identity information.

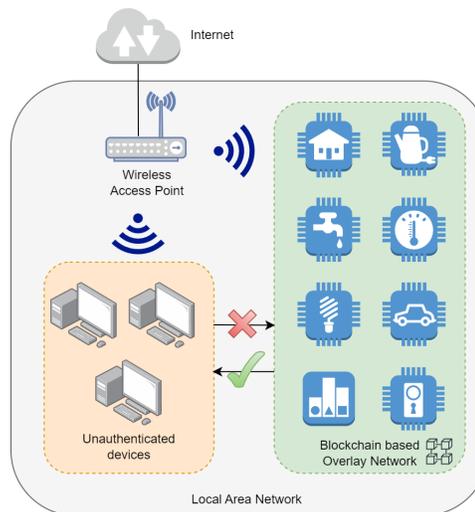


Fig. 1. Overview of the overlay network.

The final target of our scheme is to construct a secure overlay network that includes authenticated devices only. The overlay network separates authenticated devices and unauthenticated devices within the same Local Area Network (LAN). All authenticated devices discard traffic outside the overlay network, which protects them against internal and external attacks. Figure 1 shows the resulting overlay network after using our proposed scheme. More than one overlay network can be established within a LAN.

The authentication within the overlay network is supported by a coin-based blockchain system. The blockchain database stores transactions which can be used to determine the balance of an account characterized by an account number. The account number is a public key of an IoT device which also acts as the device identifier ( $Id$ ). A device identifier will be used to encrypt the communication among the authenticated devices in the overlay networks. To join the overlay network, a device needs to be authenticated by a Hardware Authenticator ( $HA$ ).  $HA$  is an offline device kept by the system administrator. This device has three functions, namely, generating genesis block, signing Authentication Transaction (AT), and generating new blocks.

The communication within the overlay network is encrypted based on the identity information provided by the blockchain system. The encryption and decryption are carried out by a firewall module within the IoT device. All traffic will be encrypted automatically without any modification to the working programs. The firewall provides network-layer encryption. For details of the authentication mechanism, readers are referred to [4].

### III. PERFORMANCE EVALUATION

It is assumed that an attacker targets an IoT network and is interested in salvaging all protected data on each device. These data are protected by the account management module of the OS. It is assumed that physical access to those IoT devices is not available for the attacker. The attacker is unauthenticated and is not in the same network as the IoT. It is also assumed that no internal attack from another device within the IoT network is possible. Therefore, the attacker can only attack IoT systems through the Internet. Exploits on software packages could sometimes expose protected data, such as the exploits on database software that could grant access to the protected folder. However, this access is limited to the workspace of the software package because a properly implemented OS contains an application within a sandbox. Unfortunately, the attacker can still inject malicious programs using these software exploits to perform privilege escalation and gain full control of the system.

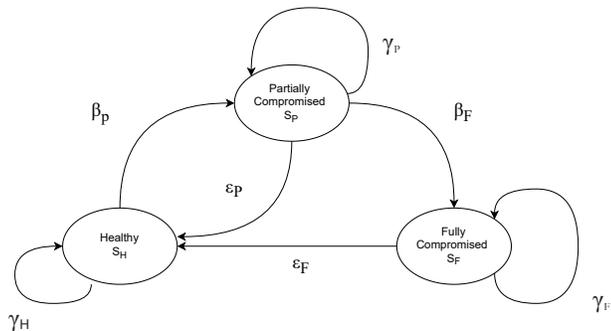


Fig. 2. Security model based on Markov chain.

Figure 2 describes our security model. Consider an IoT device is running at a healthy state ( $S_H$ ) and is not compromised

by any attackers. With probability  $\gamma_H$ , this device stays in this healthy state. The attacker keeps attacking this device which could change the state to others, including partially compromised ( $S_P$ ) and fully compromised ( $S_F$ ). The fully compromised state must be transitioned from the partially compromised state. Such a process cannot be reversed because it is irrational to turn a fully compromised system into a partially compromised one. Both compromised statuses can be recovered and return to a healthy state (e.g., by patching). A partially compromised state recovers with probability  $\epsilon_P$  while the fully compromised state has probability  $\epsilon_F$ . The probability for a healthy device to transition to a partially compromised state is  $\beta_P$  and to transition from partially compromised to fully compromised is  $\beta_F$ . However, both compromised states can remain in the same state. The partially compromised state has a probability  $\gamma_P$  to stay that way while a fully compromised state recovers probability  $\gamma_F$ . Assuming the state can only have a one-step transition, the transition probability matrix of the above Markov chain is:

$$P = \begin{bmatrix} \gamma_H & \beta_P & 0 \\ \epsilon_P & \gamma_P & \beta_F \\ \epsilon_F & 0 & \gamma_F \end{bmatrix}$$

If we put a Markov chain  $\{X_n\}$  in the long run such that  $n \rightarrow \infty$ , the probability for each state  $j$  will converge to a limiting probability ( $\pi_j$ ). These converged probabilities are considered as the steady-state probabilities. The limiting probabilities are not affected by the initial condition  $X_0$ . It can be shown that this model is a regular Markov chain when  $\beta_P$ ,  $\beta_F$  and  $\epsilon_F$  are  $> 0$ . To estimate the long-term behavior for the IoT device, we must assume the probability of being compromised  $> 0$ .

A set of system equations can be set to determine the limiting distribution ( $\pi_0, \pi_1, \pi_2$ ):

$$\gamma_H \pi_0 + \epsilon_P \pi_1 + \epsilon_F \pi_2 = \pi_0 \quad (1)$$

$$\beta_P \pi_0 + \gamma_P \pi_1 = \pi_1 \quad (2)$$

$$\pi_0 + \pi_1 + \pi_2 = 1 \quad (3)$$

Equations (2) and (3) can be rewritten as:

$$\pi_1 = \frac{\beta_P \pi_0}{1 - \gamma_P}.$$

$$\pi_2 = 1 - \pi_0 - \frac{\beta_P \pi_0}{1 - \gamma_P}.$$

By substituting  $\pi_1, \pi_2$  into Equation (1)

$$\pi_0 = \frac{\epsilon_F (1 - \gamma_P)}{(1 - \gamma_P)(1 - \gamma_H + \epsilon_F) - \beta_P (\epsilon_P - \epsilon_F)}$$

Since for each state  $i \sum_{j=0}^{\infty} P_{ij} = 1$ , therefore,  $\beta_P = 1 - \gamma_H$ . By solving equations (2) & (3):

$$\pi_0 = \frac{\epsilon_F (1 - \gamma_P)}{\epsilon_F (1 - \gamma_P) + \beta_P (\epsilon_F + \beta_F)}$$

$$\pi_1 = \frac{\beta_P \epsilon_F}{\epsilon_F (1 - \gamma_P) + \beta_P (\epsilon_F + \beta_F)}$$

$$\pi_2 = \frac{\beta_P \beta_F}{\varepsilon_F(1 - \gamma_P) + \beta_P(\varepsilon_F + \beta_F)}$$

#### IV. NUMERICAL RESULTS

Kuhn *et al.* categorize computer vulnerabilities into 18 groups [5] based on the US National Vulnerability Database (NVD) [6]. To estimate the security enhancement of the proposed scheme, this paper further groups them into categories based on the potential impact on the system: partially compromise vulnerability ( $V_P$ ) and fully compromise vulnerability ( $V_F$ ). Table I shows these two groups of vulnerabilities.

By using the previous number of vulnerabilities, the possible values of  $\beta_P$  and  $\beta_F$  can be estimated. Assume the average exploitation rate for  $V_P$  and  $V_F$  be  $E_P$  and  $E_F$ , respectively. Therefore,  $\beta_P = E_P V_P$  and  $\beta_F = E_F V_F$ . The limiting distribution is computed using the above information.  $\pi_j$  consists of four parameters:  $\beta_P$ ,  $\beta_F$ ,  $\gamma_P$ ,  $\varepsilon_F$ .

$$p[n] = 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 0.99$$

Let  $p[n]$  be a series of probabilities that are used to substitute into the model as the above parameters respectively to determine the possible outcome of  $\pi_j$ . To compute the value of  $\pi_j$ , we substitute  $p[n]$  to one of the parameters one a time while setting the other three parameters as 0.5. The last element (0.99) is added to demonstrate the behavior when the parameters are approaching the limits.

The proposed scheme can use the software firewall to filter unauthenticated communication. However, some of the vulnerabilities cannot be resolved because the proposed scheme relies on them. Three partially compromise vulnerabilities including configuration, cryptographic issues, authentication issues, and all fully compromise vulnerabilities cannot be stopped by the proposed scheme.

Misconfiguration can still paralyze the proposed scheme because the privileged user can deactivate the software or avoid starting it at the beginning. The proposed scheme relies on cryptography to authenticate and secure communication. If the problem resides in cryptographic issues, the proposed scheme will not work properly. Authentication is the key to the proposed scheme; it will fail if it cannot do authentication properly. The proposed scheme would use a different  $V_P$  based on this property. Table II shows the values of  $V_P$  for the cases of using and without using the proposed scheme, respectively.

The calculation of the probability for the proposed scheme will be using the new  $V_P$ . Therefore, the model should produce a higher value for  $\pi_0$ , which is the probability for a healthy state while reducing the probability for a partially and fully compromised state.

Figure 3 shows the probability of a healthy state at the steady-state of the Markov process without applying the proposed scheme. All parameters are displaying a decaying behavior when  $p[n]$  increases except  $\varepsilon_F$ .  $\gamma_P$  has the fastest decaying rate which is polynomial decay.  $\beta_P$  and  $\beta_F$  are linear decay. The range of  $\pi_0$  when replacing  $\beta_P$  with  $p[n]$  is larger

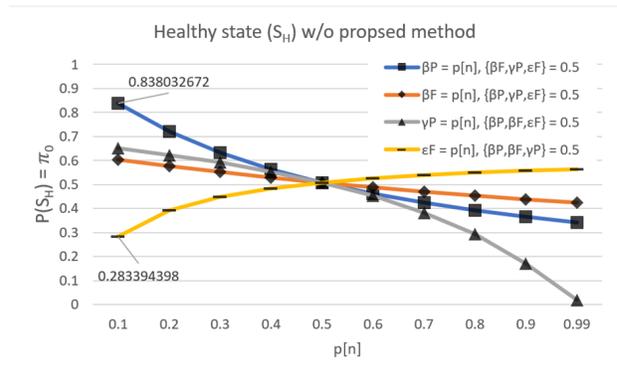


Fig. 3. Estimated probability of healthy state ( $S_H$ ) without the proposed scheme.

than  $\beta_F$ , which also decays faster. This phenomenon indicates that the healthy state is more sensitive to the probability of partially compromise vulnerability than fully compromise vulnerability. Parameters including  $\beta_P$ ,  $\beta_F$ ,  $\varepsilon_F$  yield 0.28 to 0.83 and 0.51 on average. The overall average probability for staying in a healthy state without using the proposed scheme is 0.48.

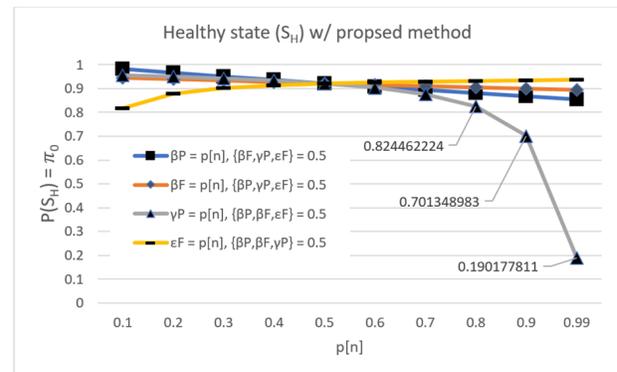


Fig. 4. Estimated probability of healthy state ( $S_H$ ) with the proposed scheme.

Figure 4 shows the probability of a healthy state at the steady-state of the Markov process by applying the proposed scheme. Compared to Figure 3, the probability of staying in a healthy state is greatly increased. Parameters including  $\beta_P$ ,  $\beta_F$ ,  $\varepsilon_F$  yield 0.81 to 0.95 and 0.91 on average in the result. The decaying behavior is similar except  $\gamma_P$ . Between 0.9 and 0.99, the decay is greater than the previous data point. This further indicates the healthy state is more sensitive to the probability of partially compromise vulnerability than fully compromise vulnerability. The proposed scheme increased the overall probability of staying in a healthy state from 0.48 to 0.89, which is an 85% improvement.

A similar analysis can be carried out for the partially compromised state and fully compromised state. Here, we summarize the results in Table III, which shows the comparison of the average probability for a healthy state, partially compromised, and fully compromised state for an IoT system

TABLE I  
VULNERABILITY COUNTS AND CATEGORIES FROM 2008-2016

| Type of vulnerability               | Count | Percentage | Level                          | Possible attack                                       |
|-------------------------------------|-------|------------|--------------------------------|---|
| Format String Vulnerability         | 110   | 0.294709   | $V_F$                          | - Execute arbitrary code                              |
| Configuration                       | 195   | 0.522438   | $V_P/V_F$                      | - Exposure of config file<br>- Execute arbitrary code |
| OS Command Injections               | 208   | 0.557267   | $V_F$                          | - Execute arbitrary code                              |
| Race Conditions                     | 377   | 1.010047   | $V_F$                          | - Privilege escalation                                |
| Link Following                      | 389   | 1.042197   | $V_F$                          | - Privilege escalation                                |
| Credentials Management              | 589   | 1.578031   | $V_F$                          | - Privilege escalation                                |
| Cryptographic Issues                | 779   | 2.087073   | $V_P / V_F$                    | - Information leakage<br>- Password leakage           |
| Authentication Issues               | 920   | 2.464836   | $V_P / V_F$                    | - Information leakage<br>- Privilege escalation       |
| Cross-Site Request Forgery (CSRF)   | 1161  | 3.110516   | $V_P$                          | - Information leakage                                 |
| Numeric Errors                      | 1199  | 3.212324   | $V_F$                          | - Privilege escalation                                |
| Code Injection                      | 1545  | 4.139317   | $V_F$                          | - Execute arbitrary code                              |
| Path Traversal                      | 1686  | 4.51708    | $V_P$                          | - Information leakage                                 |
| Information Leak / Disclosure       | 2939  | 7.874079   | $V_P$                          | - Information leakage                                 |
| Input Validation                    | 3763  | 10.08171   | $V_P / V_F$                    | - Information leakage<br>- Execute arbitrary code     |
| SQL Injection                       | 3828  | 10.25586   | $V_P / V_F$                    | - Information leakage<br>- Execute arbitrary code     |
| Permissions, Privileges, and Access | 4661  | 12.48761   | $V_F$                          | - Privilege escalation                                |
| Cross-Site Scripting (XSS)          | 6220  | 16.66443   | $V_P$                          | - Information leakage                                 |
| Buffer Errors                       | 6756  | 18.10047   | $V_F$                          | - Privilege escalation                                |
| Total                               | 37325 | 100        | $V_P = 57.57$<br>$V_F = 67.83$ |   |

TABLE II  
COMPARISON OF  $V_P$ .

|       | Without proposed scheme | With proposed scheme |
|-------|-------------------------|----------------------|
| $V_P$ | 0.5757                  | 0.0507               |
|       | Average improvement     | 80.43%               |

with and without our proposed authentication scheme. The proposed scheme increases the probability of a healthy state for the IoT device to 0.89. The probability of being partially compromised or fully compromised is reduced to 0.10.

TABLE III  
COMPARISON OF STATE PROBABILITIES.

|   | Without proposed scheme | With proposed scheme | Difference |
|---|-------------------------|----------------------|------------|
| Average $\pi_0$ (Healthy state)               | 0.4852                  | 0.8911               | +83.66%    |
| Average $\pi_1$ (Partially compromised state) | 0.2989                  | 0.0625               | -79.09%    |
| Average $\pi_2$ (Fully compromised state)     | 0.2158                  | 0.0463               | -78.53%    |
|   | Average improvement     |                      | 80.43%     |

## V. CONCLUSION

In this paper, we have developed a stochastic threat model for IoT systems, which is used to evaluate the efficacy of our earlier developed authentication scheme. Numerical results have demonstrated that when the authentication scheme is deployed, the security level of IoT systems is significantly increased.

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