

A Review on Malware Threat Analysis Approaches for IoT

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Abstract: Internet of things (IoT) is a concept that has been widely used to improve business efficiency and customer's experience. It involves resource constrained devices connecting to each other with a capability of sending data, and some with receiving data at the same time. The IoT environment enhances user experience by giving room to a large number of smart devices to connect and share information. However, with the sophistication of technology has resulted in IoT applications facing with malware threat. Therefore, it becomes highly imperative to give an understanding of existing state-of-the-art techniques developed to address malware threat in IoT applications. In this paper, we studied extensively the adoption of static, dynamic and hybrid malware analyses in proffering solution to the security problems plaguing different IoT applications.

Keywords: Data analytics, Internet of things, Malware threat, Mobile threat

1. Introduction:

Over the years, the internet of things (IoT) concept has exhibited great potential for actuating various domains (personal and enterprise environments); with examples and likely applications but not restricted, smart health for cashless and easy admission into major hospitals, smart cities for energy cost and pollution reduction, smart transportation for developing alternative means to solve road traffic issues as well as smart homes whereby energy industries are developing systems for increasing energy preservation and security among others [1], [2].

With the advent of IoT, computing platforms for general purpose that runs on conventional desktops are now been substituted by platforms like tablets and smartphones. High functionality applications that were once restrained for usage on highly efficient desktops and laptops are currently accessible on the existing mobile platforms as their computational power rises. The ripple effects of the growing trend in usage and popularity of the smartphones have been evidenced in several internet applications where products, accessibility and applications have been migrated to the platform for productivity and interoperability enhancement [3], [4]. IoT technologies are being utilized as foundational technologies for the cooperative-intelligent transport system (C-ITS), that is regarded as next generation intelligent transportation system [5]. Hence, IoT has become a bridge that

unites the physical and digital world through the inclusion of smart objects that relate with physical ambience with no direct human interference [6]. Highly essential in IoT applications are confidentiality, authentication, access control and integrity through the implementation of accurate security and privacy protocols [7]. Whilst the novel functionality derived from the IoT can be utilized in improving the lives of humans, the possibility of conventional cyber-attacks on IoT system is rife [8]. More so, it has been well established that a lot of IoT gadgets are susceptible to simple intrusion trials such as utilizing weak or at times default passwords. For instance, in spite of the high sensitivity exhibited by mobile platforms and the tendencies for abuse, various security threats not so different from those that are currently affecting their desktop counterparts have begun to emerge. Similarly with the capacity of smartphones to have several sensors and connection with a lot of IoT gadgets, it becomes a main target for malwares [9].

Furthermore, their nature to host other integrated components such as microphones, inbuilt cameras, accelerometer etc. makes them prone to hijack by malware as well as eavesdrop on their enclosure [10]. Recently, significant numbers of malware including worms, viruses, Trojan horses and rootkits have begun to emerge which are targets at exfiltrating confidential data in mobile platform or leverages on the compromised asset so as to access confidential networks through which the gadget has genuine access [11], [12]. Atamli and Martin [13] has identified the three primary groups of malicious entities threatening IoT as: (1) external attackers, (2) malicious users and (3) bad manufacturers [13]. For the purpose of assisting academics and developers to easily comprehend the insight of several kinds of IoT security attacks as well as to create relevant security measures in their IoT developments, Nawir et al. [14] designed a well-organized taxonomy as presented in Figure 1 that outlines eight different categories under which attackers can attack the IoT systems.

Meanwhile, traditional computers bring a lot of attacks in IoT environment and uses these computers to infect other connected devices in IoT environment. Having seen these trends, IoT applications are imminently a new area of security research. In this paper, we comprehensively explored the analysis techniques and approaches of current IoT applications regarding series of threats from malware. Several points of interest that an attacker can manipulate either by gaining access to unauthorized information or by causing a denial of

service have been identified alongside with the appropriate security architecture to prevent the occurrence.

Malware detection methods can be static or dynamic [35]. In dynamic malware detection approaches, the program is executed in a controlled environment (e.g., a virtual machine or a sandbox) to collect its behavioral attributes such as required resources, execution path, and requested privilege, in order to classify a program as malware or benign [36], [37], [38]. Static approaches (e.g., signature-based detection, byte-sequence n-gram analysis, opcode sequence identification and control flow graph traversal) statically inspect a program code to detect suspicious applications. David et al [39] proposed DeepSign to automatically detect malware using a signature generation method. The latter creates a dataset based on behaviour logs of API calls, registry entries, web searches, port accesses, etc, in a sandbox and then converts logs to a binary vector. They used deep belief network for classification and reportedly achieved 98.6% accuracy. In another study, Pascanu et al. [40] proposed a method to model malware execution using natural language modeling. They extracted relevant features using recurrent neural network to predict the next API calls. Then, both logistic regression and multi-layer perceptions were applied as the classification module on next API call prediction and using history of past events as features. It was reported that 98.3% true positive rate and 0.1% false positive rate were achieved. Demme et al. [41] examined the feasibility of building a malware detector in IoT nodes' hardware using performance counters as a learning feature and K-Nearest Neighbor, Decision Tree and Random Forest as classifiers. The reported accuracy rate for different malware family ranges from 25% to 100%. Alam et al. [42] applied Random Forest on a dataset of Internet-connected smartphone devices to recognize malicious codes. They executed APKs in an Android emulator and recorded different features such as memory information, permission and network for classification, and evaluated their approach using different tree sizes. Their findings showed that the optimal classifier contains 40 trees, and 0.0171 of mean square root was achieved. In order to detect crypto-ransomware on Android devices as management nodes of an IoT networks, Azmoodeh et al. [43] recorded the power usage of running processes and identified distinguishable local energy consumption patterns for benign applications and ransomware. They broke down the power usage pattern into sub-samples and classified them, as well as aggregating sub-samples' labels to determine the final label. The proposed approach reportedly achieved 92.75% accuracy. The need to secure IoT backbone against malware attacks motivated Haddad Pajouh et al. [44] to propose a two-layer dimension reduction and two-tier classification module to detect malicious activities. Specifically, the authors used Principle Component Analysis and Linear Discrimination Analysis to reduce the dataset and then used Naïve Bayes and K-Nearest Neighbor to classify samples. They achieved

detection and false alarm rates of 84.86% and 4.86%, respectively. While OpCodes are considered an efficient feature for malware detection, there does not appear to have been any attempt to use OpCodes for IoT and IoBT malware detection. In addition, using deep learning for robust malware detection in IoT networks appears to be another understudied topic. Thus, in this paper, we seek to contribute to this gap by exploring the potential of using OpCodes as features for malware detection with deep Eigenspace learning.

E. Bertino, K.-K. R. Choo, D. Georgakopoulos, and S. Nepal, 2016, The Internet of Things (IoT) is the latest Internet evolution that incorporates a diverse range of things such as sensors, actuators, and services deployed by different organizations and individuals to support a variety of applications. The information captured by IoT presents an unprecedented opportunity to solve large-scale problems in those application domains to deliver services; example applications include precision agriculture, environment monitoring, smart health, smart manufacturing, and smart cities. Like all other Internet based services in the past, IoT-based services are also being developed and deployed without security consideration. By nature, IoT devices and services are vulnerable to malicious cyber threats as they cannot be given the same protection that is received by enterprise services within an enterprise perimeter. While IoT services will play an important role in our daily life resulting in improved productivity and quality of life, the trend has also "encouraged" cyber-exploitation and evolution and diversification of malicious cyber threats. Hence, there is a need for coordinated efforts from the research community to address resulting concerns, such as those presented in this special section. Several potential research topics are also identified in this special section.

X. Li, J. Niu, S. Kumari, F. Wu, A. K. Sangaiah, and K.-K. R. Choo, 2017, Internet of Things (IoT) is an emerging technology, which makes the remote sensing and control across heterogeneous network a reality, and has good prospects in industrial applications. As an important infrastructure, Wireless Sensor Networks (WSNs) play a crucial role in industrial IoT. Due to the resource constrained feature of sensor nodes, the design of security and efficiency balanced authentication scheme for WSNs becomes a big challenge in IoT applications. First, a two-factor authentication scheme for WSNs proposed by Jiang et al. is reviewed, and the functional and security flaws of their scheme are analyzed. Then, we proposed a three-factor anonymous authentication scheme for WSNs in Internet of Things environments, where fuzzy commitment scheme is adopted to handle the user's biometric information. Analysis and comparison results show that the proposed scheme keeps computational efficiency, and also achieves more security and functional features. Compared with other related work, the proposed scheme is more suitable for Internet of Things environments.

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J. Gubbi, R. Buyya, S. Marusic, and M. Palaniswami, 2013, Ubiquitous sensing enabled by Wireless Sensor Network (WSN) technologies cuts across many areas of modern day living. This offers the ability to measure, infer and understand environmental indicators, from delicate ecologies and natural resources to urban environments. The proliferation of these devices in a communicating-actuating network creates the Internet of Things (IoT), wherein, sensors and actuators blend seamlessly with the environment around us, and the information is shared across platforms in order to develop a common operating picture (COP). Fuelled by the recent adaptation of a variety of enabling device technologies such as RFID tags and readers, near field communication (NFC) devices and embedded sensor and actuator nodes, the IoT has stepped out of its infancy and is the the next revolutionary technology in transforming the Internet into a fully integrated Future Internet. As we move from www (static pages web) to web2 (social networking web) to web3 (ubiquitous computing web), the need for data-on-demand using sophisticated intuitive queries increases significantly. This paper presents a cloud centric vision for worldwide implementation of Internet of Things. The key enabling technologies and application domains that are likely to drive IoT research in the near future are discussed. A cloud implementation using Aneka, which is based on interaction of private and public clouds is presented. We conclude our IoT vision by expanding on the need for convergence of WSN, the Internet and distributed computing directed at technological research community.

F. Leu, C. Ko, I. You, K.-K. R. Choo, and C.-L. Ho, 2017, Recently, Wireless Body Sensor Networks (WBSNs) have been popularly employed to measure people's physiological parameters, particularly for disease monitoring, prevention, and treatment. In this study, we propose a smartphone-based WBSN, named Mobile Physiological Sensor System (MoPSS), which collects users' physiological data with body sensors embedded in a smart shirt. A patient's vital signs are continuously gathered and sent to a smart phone in a real-time manner. The data are then delivered to a remote healthcare cloud via WiFi. After performing necessary classification and analysis, the health information of individual patients is also stored in the cloud, from which authorized medical staffs can retrieve required data to monitor patients' health conditions so that when necessary, caregivers are able to reach the patients as soon as possible and provide required assistance. Our simulations demonstrate that the presented healthcare system provides a better solution for health management.

M. Roopaei, P. Rad, and K.-K. R. Choo, 2017, Irrigation Is Crucial For Agriculture Production To Ensure That Farmers Are Able To Meet Crop Water Demands Even In Situations Where There Is Inadequate Rainfall. However, poor irrigation scheduling and inefficient utilization of water resources are two of several ubiquitous parameters restricting production in many agricultural regions. Cultivators can use information

such as light, humidity and temperature levels to modify irrigation schedules and avoid the risk of damaging crops.² For example, soil sensors can be used to collect information on how water flows through the land and can be used to track changes in soil moisture, temperature, and levels of nitrogen and carbon. These sensors can work in conjunction with drip irrigation methods and fertigation to avoid unnecessary waste of water and fertilizer, thus, increasing fruit and leaf quality. Real-time data of weather predictions, soil conditions, crop features, etc. can support farmers in making informed decisions on which crops to plant where and when as well as when to plough, etc. This allows the monitoring, optimization, and precise control of high-yielding (wheat, corn, etc.) and sensitive crops (vineyards, tropical fruits, etc.), whether cultivated outdoors or in greenhouses. This permits farmers to help reach maximum crop production with optimal quality.

X. Li, J. Niu, S. Kumari, F. Wu, and K.-K. R. Choo, 2017, Smart city is a development tendency of future city, which improves almost all aspects of quality of urban residents' life by adopting Information and Communication Technology. In smart city, people can interact directly with the community and the infrastructure at anytime and anywhere, where GLObal MObility NETwork (GLOMONET) is an important network infrastructure for smart city. Recently, Gope and Hwang proposed an efficient authentication scheme for GLOMONET. However, we find their scheme lacks session key update and wrong password detection mechanisms, and vulnerable to denial-of-service attack. Besides, the session key can be known by HA (home agent), and perfect forward secrecy cannot be ensured. Furthermore, in their scheme, HA has to take heavy secret key management work. Based on previous work, this paper first summarizes the security and function requirements of authentication for GLOMONET in smart city environment. Later, this paper proposed a robust biometrics based three-factor authentication scheme for GLOMONET in smart city. Security features of the proposed scheme are analyzed in detail, and comparisons of our scheme with other related schemes are illustrated. Analysis and comparison results show that our scheme meets the preconcerted security requirements of authentication for GLOMONET in smart city environment, and it is robust for GLOMONET in smart city environments with higher security requirements.

D. Miorandi, S. Sicari, F. De Pellegrini, and I. Chlamtac, 2012, The term "Internet-of-Things" is used as an umbrella keyword for covering various aspects related to the extension of the Internet and the Web into the physical realm, by means of the widespread deployment of spatially distributed devices with embedded identification, sensing and/or actuation capabilities. Internet-of-Things envisions a future in which digital and physical entities can be linked, by means of appropriate information and communication technologies, to enable a whole new class of applications and services. In this

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article, we present a survey of technologies, applications and research challenges for Internet-of-Things.

Kott, A. Swami, and B. J. West, 2016, The rapid emergence of Internet of Things is propelled by the logic of two irresistible technological arguments: machine intelligence and networked communications. Things are more useful and effective when they are smarter, and even more so when they can talk to each other. Exactly the same logic applies to things that populate the world of military battles. They too can serve the human warfighters better when they possess more intelligence and more ways to coordinate their actions among themselves. We call this the Internet of Battle Things, IoBT. In some ways, IoBT is already becoming a reality¹, but 20-30 years from now it is likely to become a dominant presence in warfare. The battlefield of the future will be densely populated by a variety of entities (“things”) – some intelligent and some only marginally so – performing a broad range of tasks: sensing, communicating, acting, and collaborating with each other and human warfighters². They will include sensors, munitions, weapons, vehicles, robots, and human-wearable devices. Their capabilities will include selectively collecting and processing information, acting as agents to support sensemaking, undertaking coordinated defensive actions, and unleashing a variety of effects on the adversary. They will do all this collaboratively, continually communicating, coordinating, negotiating and jointly planning and executing their activities. In other words, they will be the Internet of Battle Things.

C. Tankard, 2015, Internet of Things (IoT) is playing a more and more important role after its showing up, it covers from traditional equipment to general household objects such as WSNs and RFID. With the great potential of IoT, there come all kinds of challenges. This paper focuses on the security problems among all other challenges. As IoT is built on the basis of the Internet, security problems of the Internet will also show up in IoT. And as IoT contains three layers: perception layer, transportation layer and application layer, this paper will analyze the security problems of each layer separately and try to find new problems and solutions. This paper also analyzes the cross-layer heterogeneous integration issues and security issues in detail and discusses the security issues of IoT as a whole and tries to find solutions to them. In the end, this paper compares security issues between IoT and traditional network, and discusses opening security issues of IoT.

C. J. D’Orazio, K. K. R. Choo, and L. T. Yang, 2017, Increasingly, big data (including sensitive and commercial-in-confidence data) is being accessible and stored on a range of Internet of Things (IoT) devices, such as our mobile devices. Therefore, any vulnerability in IoT devices, operating system or software can be exploited by cybercriminals seeking to exfiltrate our data. In this paper, we use iOS devices as case studies and highlight the potential for pairing mode in iOS devices (which allows the establishment of a trusted

relationship between an iOS device and a personal computer) to be exploited for covert data exfiltration. In our three case studies, we demonstrate how an attacker could exfiltrate data from a paired iOS device by abusing a library and a command line tool distributed with iTunes. With the aim of avoiding similar attacks in the future, we present two recommendations. **M. Conti, A. Dehghantanha, K. Franke, and S. Watson, 2018**, The Internet of Things (IoT) envisions pervasive, connected, and smart nodes interacting autonomously while offering all sorts of services. Wide distribution, openness and relatively high processing power of IoT objects made them an ideal target for cyber-attacks. Moreover, as many of IoT nodes are collecting and processing private information, they are becoming a goldmine of data for malicious actors. Therefore, security and specifically the ability to detect compromised nodes, together with collecting and preserving evidences of an attack or malicious activities emerge as a priority in successful deployment of IoT networks. In this paper, we first introduce existing major security and forensics challenges within IoT domain and then briefly discuss about papers published in this special issue targeting identified challenges.

3. Conclusion:

In this study, we have documented comprehensive analysis techniques with cutting-edge solutions to address malware threat in IoT applications. In particular, static, dynamic as well as hybrid analyses have been adopted by researchers to confront security issues plaguing several IoT applications. The effectiveness of the documented analysis techniques was demonstrated using case studies including smart home systems, smart factories, smart gadgets and IoT application protocols. Systems such as 6thSense, a comprehensive context-aware architecture for sensors security in IoT devices offers approximately 97% accuracy and F-score. Similarly the location-privacy preserving mechanisms (LPPMs) with integrated targeted maneuvers offers a robust defense against white-box attacks by reducing the probability success of white-box attacks to 3%.

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