

WIRELESS MONITORING AND CONTROLLING SYSTEM USING CYBER SENSOR NETWORK FOR INDUSTRIAL APPLICATIONS

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Abstract— The expanding reach and wide deployment of digital systems into our most fundamental physical infrastructures pose both a complexity and security concern. The complexity of modern industrial networks primarily stems from three areas: presence of heterogeneous hardware and software, dynamic network composition and usage patterns, and decentralization of control. Dealing with these complexities requires a solution that is flexible and manageable. In existing system, industrial data's are only monitoring through WAN using SOAP and IF-MAP protocol in software manner. But data monitoring is not enough because in critical time admin should have rights to control the industrial data's to avoid the accidents and production wastage. In proposed system, we are using both hardware and software manner. Here SOAP protocol is used in hardware part detects the sensor values through WSN system. IF-MAP is used in software part for data sharing and security purpose for WAN. D-bus is used to link the hardware and software parts. Using this system we can monitor and control the entire industrial system. Here mainly industrial parameters are controlled through WAN.

I. INTRODUCTION

The expanding reach and wide deployment of digital systems into our most fundamental physical infrastructures pose both a complexity and security concern [1], [2]. The complexity of modern industrial networks primarily stems from three areas: presence of heterogeneous hardware and software, dynamic network composition and usage patterns, and decentralization of control [3]. Dealing with these complexities requires a solution that is flexible and manageable.

An autonomic digital ecosystem (DE) is a model for future production systems that builds on the notion of autonomic self-management by embedding exploitable control features within modules [4]. Cyber-physical ecosystems (CPEs) include interconnected subsystems that sense and act upon the physical world to form a complex system of systems. These CPEs are a foundational layer of a wider more encompassing digital ecosystem. CPEs typically bridge the cyber-world of computing and communications with the physical world by embedding wireless sensor nodes into the

physical world [5]. Therefore, wireless sensor networks (WSNs) are a fundamental building block of digital ecosystems.

B. IF-MAP Introduction

As was mentioned in the previous section, the communication interface is a critical component of AICS. A messaging interface based on the interface to metadata access points (IF-MAP) version 2.0 standard was implemented. IF-MAP is an open protocol standard published by the Trusted Computing Group [15]. Originally made available in April 2008, version 2.0 rev. 47 was published in November 2011. It utilizes a publish, subscribe, messaging paradigm implemented with the simple object access protocol (SOAP). The design goal of the IF-MAP working group was to enable the sharing of data between network devices and systems. IF-MAP has multiple defined parts: a base protocol and metadata for a specific usage. Currently, the only metadata defined is for network security, which is sufficient for use in AICS [16]. One of the benefits provided by this definition is a common reference standard of network security data. In addition to the currently defined metadata, it is possible to create a new metadata specification if the base protocol is followed.

1) Base Protocol and Metadata: The base protocol specifies possible actions for clients. IF-MAP clients have access to three actions for metadata: publish, search, and subscribe. Clients store or publish metadata into a metadata access point (MAP) for others to consume. A search allows clients to obtain published metadata from the MAP based on a flexible criterion. Subscribe requests asynchronous results from a predefined search whenever a client publishes new metadata. In this way, clients can monitor for specific relevant events and be alerted when one occurs. It should be noted that all IF-MAP operations and data types are expressed utilizing XML. This helps ensure a consistent format of the data when it is exchanged between multiple diverse parties.

The protocol also defines two species of data: metadata and identifiers. Metadata in this context is any shared data about network devices, policies, status, behavior, or observed relationships. The five defined identifiers are identity, IP Address (v4 and v6), MAC address, access request, and device. Identifiers are used to refer to specific metadata items. The metadata of interest to this project are as follows.

- **ip-mac:** a binding between an IP address and a mac address valid for a time frame;
- **device-characteristic:** an inherent characteristic of an entity such as its manufacturer or OS;

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- **device-ip:** the IP address of an associated device;
- **discovered-by:** a link identifying which sensor device has discovered a new IP or MAC address;
- **event:** activity of interest on a network such as a virus attack or abnormal behavior.

2) *SOAP:* All IF-MAP actions are transmitted using SOAP. The SOAP version 1.2 protocol is a specification utilizing XML for exchanging structured and typed information in a distributed environment. Typically, hypertext transfer protocol (HTTP) or SMTP is used for message handling. It provides an extensible framework for transferring application specific information without specifically detailing the semantics of the data carried. The framework provides required actions to perform on a SOAP message in order to send and receive it. SOAP is typically used as a messaging framework layer of a Web-based SOA. SOA is an enabling technology for development of large systems in industry.

C. D-Bus

In this paper, the use of D-Bus is described for internal communication between sensor components and the external messaging service. D-Bus is a system created for interprocess communication (IPC) that has been used in reconfigurable mobile network devices [18]. It was designed to avoid round trips and allow for asynchronous operation. Conceptually, it works in terms of messages and handles many of the difficult IPC synchronization issues. Several high-level language bindings are available, which provides a flexible implementation solution. The following describes the D-Bus system and the terminology commonly used with it.

A *bus* is a virtual path that messages are passed over. Multiple instances of a bus may exist and are managed by a software daemon. Several Linux-based operating systems offer two buses, system and session, for use without requiring user configuration. Fine-grained user access to buses and actions on the appropriate bus service is available via configuration files. Applications that utilize a bus take on the role of either a server or client. Server processes listen for incoming method requests from clients.

A *method* is made available by publishing it on a message bus. Methods are remotely invoked operations of an object. As is found in object-oriented programming languages, objects contain methods. The input and output parameters need to be defined in advance of publication. A method is registered with the D-Bus daemon under an *interface* and available through an *object path*. Interfaces are a collective group of methods that help define the type of an object.

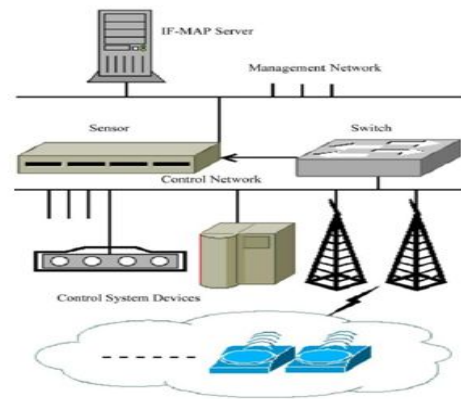
Fuzzy logic provides a framework for system modeling in linguistic form capable of coping with imprecise and vague meanings of words. The Type-2 Fuzzy Logic System (T2 FLS) is an extension of the Type-1 Fuzzy Logic System (T1 FLS) that has been successful in modeling and minimizing the effects of various kinds of dynamic uncertainties. Here, the Interval Type-2 Fuzzy Logic System (IT2 FLS), as a specific case of T1 FLS, is used to encode the linguistic domain knowledge about the specific network system and dynamically adjust the sensitivity of the anomaly detection algorithms.

The IAA component is based on an existing low-cost online rule extraction technique [21]. The model is composed of a set of fuzzy rules that are constructed based on a

window-based feature vector using an online version of the adapted nearest neighbor clustering (NNC) algorithm. The anomaly detection algorithm was specifically designed to allow for both fast learning and fast classification on the constrained computational resources of an embedded device. The overall architecture of the proposed anomaly detection system is depicted in Fig. 3. The network traffic is processed by an IT2 FLS that uses a fuzzy logic rule base with encoded linguistic domain knowledge to calculate the cyber-security context. This cyber-security context expresses the belief that an intruder is currently present in the system.

In the next stage, the network traffic is separated into individual communication streams. Each stream is delineated by an IP address retrieved from the internal message bus. Other features, such as port numbers or protocol types, can also be used. Packets assigned to individual communication streams are then passed into dedicated anomaly detection algorithms. Each anomaly detection algorithm maintains its own buffer of incoming packets, which is used to extract a set of window-based features.

AICS was developed as a deployable modular framework on a commonly available hardware platform with components that provide network host information, identify anomalous network traffic behaviors and deploy dynamic virtual honeypots. In the test scenario, the sensor was deployed on a self-contained hardware platform. This effectively hides the communication between components from other network entities, thereby adding another layer of defense. When considering the possible addition of information from external hosts on the IF-MAP interface, a balance between security and information access would need to be evaluated.



II. COMPARISON TO EXISTING WORK AND EVALUATION

A related anomaly detection project by Dussel *et al.* relies on the computation of similarity between transport-layer packet payloads embedded in a geometric space [28]. They performed experiments on traffic from a SCADA test bed containing various application-layer protocols. A primary difference from AICS includes the use of packet data, as opposed to header derived information, to produce *n*-gram populated feature vectors. A simple distance measure is then compared with stored “normal” vectors to determine if an anomaly exists in the traffic. The AICS IAA algorithm maintains two models of system behavior: a global threat model that considers all traffic and a set of individual models for a given stream. Their solution appears to broadly apply itself to all traffic. It is difficult to compare results with

different test data sets, but their average detection rate of 88%–92% at a false positive level of 0.2% may be due to the differences just mentioned.

An additional beneficial outcome of the presented solution is removal of the configuration burden from the human operator. This capability stems from the self-configuring aspects of an Autonomic design. For example, the IAA anomaly detection routine doesn't require human created rules. It learns normal behavior from observations of the system. In addition, the dynamic honeypots are configured from system observations. This functionality reduces dependence on network expertise and level of human configuration effort.

The CPU and RAM utilization was monitored during the test scenario. As the hardware platform is a fanless system, temperature data was recorded as well. The 200 000 anomaly evaluation packets contained 19 627 063 bytes of payload information. Approximately 97 packets per second were processed normal network transmission speed. It was observed that each anomaly alert sent from IAA on the IF-MAP mechanism incurred an overhead of .7 ms of communication processing time.

The CPU thermal sensors were polled every 5 s. Room temperature at the start of the test was 23 °C. Core 0 temperature after a 5-min warm-up period was 44 °C. Values during the test ranged from 42 °C to 47 °C. Core 1 temperature during the warm-up period was 45 °C. The test values ranged from 43 °C to 48 °C. It is speculated that the difference in the range low value and the initial start might be due to variations in room temperature during the test.

The at-rest CPU utilization, as measured by the Linux top command on 1-s intervals, was 1.1%. This was due mainly to the Xorg process providing HMI services. The utilization maximum during the test was 47.4% with an average of .65% and standard deviation of 2.29%. The maximum occurred when the virtual honeypots were probed with Nmap. The graph in Fig. 7 shows the sampled CPU utilization before and after the maximum that occurred between time index 100 and 181. Another key performance indicator for Linux machines is the amount of time the CPU has been waiting for I/O to complete. The maximum value was 10% with an average of 0.59% and standard deviation of 0.62%. In other words, the machine was not overwhelmed with the test data when running in real time.

CONCLUSION

Inspired by self-configuring aspects of autonomic computing, the work presented here supports the important goal of securing industrial ecosystems by providing network security awareness in a heterogeneous control system network. Contributions of this paper include 1) a flexible two-level communication layer based on autonomic computing and service-oriented architecture and 2) descriptions of three dynamic modules that respond to a changing environment.

A novel working sensor prototype was developed based on a unique implementation of the Trusted Network Group's IF-MAP Web-services-based communication protocol. Sensor communication between self-contained modules is accomplished with D-Bus. At multiple steps of a test scenario, the communication layer was utilized to provide

information in a well-defined format between components and to external entities.

As can be seen in the architecture and scenario sections, the self-configuration capability of autonomic systems is exemplified by the IAA and DHP components. The DHP host entities are created automatically based upon passive monitoring of network activity. Anomaly behavior in IAA is detected by clustering of information retrieved from a moving window of traffic.

The only source of outside information explicitly provided is the IP addresses of devices deemed to be critical in the functioning system. This information is asynchronously delivered by a de-coupled two-part communication system.

Multiple intelligent modules were deployed on a test system to monitor for anomalous behavior and create deceptive emulated hosts. The modules are exemplary implementations of self-learning components. In a test scenario, 45 of the 46 network attached devices were recognized, and 10 of the 12 emulated devices were created with representative characteristics. In addition, 99.9% of anomalous packets were recognized. The modules utilized the communication layer to provide notifications and retrieve information.

Future work includes integration of AICS with external components utilizing the IF-MAP standard. These components include functionality to correlate information from the sensor with production data. Multiple sensors will be deployed necessitating further use Web standards such as WS-Management.

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