

Advanced Techniques in Network Traffic Analysis: Utilizing Wireshark for In-Depth Live Data Packet Inspection and Information Capture

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Abstract - The escalating frequency and sophistication of cyber-attacks have placed a spotlight on the importance of Quality of Service (QoS) and robust network security mechanisms. Effective traffic analysis and distribution are critical for maintaining the integrity of network applications and safeguarding data. Our research focuses on the development and assessment of a network intrusion detection system (NIDS) that utilizes the advanced packet analysis capabilities of Wireshark, a renowned network protocol analyzer. The system is designed to enhance network security by enabling more efficient and accurate data collection through network monitoring tools. This serves as the foundation for detecting and thwarting a wide range of malicious activities, including malware and spyware incursions, within network traffic. The innovation of our approach lies in its integration of packet sniffing technology with AI-driven analysis techniques. By applying Machine Learning algorithms to the data captured by Wireshark, our NIDS is capable of identifying patterns indicative of intrusive behavior, which might elude traditional detection methods. We present a comprehensive evaluation of our system, showcasing its efficacy in real-time intrusion detection and its potential for integration into existing network infrastructures. This study contributes to the ongoing discourse in cybersecurity, offering a sophisticated tool in the battle against digital threats and reinforcing the role of intelligent systems in protecting network environments..

Keywords- Network Traffic Analysis, Wireshark for Data Packet Inspection, Real-Time Network Monitoring, packets, Data Packet Analysis Techniques, Network Security and Forensics.

I. INTRODUCTION

In an era where the internet has become the backbone of global communication and data exchange, network security stands as a critical pillar of Computer Science and Information Technology [1]. With daily cyber threats escalating in both volume and complexity, the quality of service (QoS) and the reliability of network applications are under constant threat. This has compelled researchers and network operators to devise sophisticated mechanisms for traffic analysis and distribution to preemptively identify and mitigate these threats [2]. The advent of Network Intrusion Detection Systems (NIDS) marks a significant milestone in our ability to defend against cyber-

attacks. These systems serve as the sentinels of network security, monitoring traffic to detect malicious activities in their incipiency. Our investigation revolves around one such tool at the forefront of network monitoring technology—Wireshark. Known for its comprehensive packet analysis capabilities, Wireshark enables the meticulous scrutiny of network traffic, thereby forming the crux of our intrusion detection methodology. This paper elucidates the development of an NIDS that harnesses Wireshark's potent packet sniffing functionalities, complemented by AI and Machine Learning techniques [3-4]. By integrating these computational methods, the proposed system goes beyond conventional rule-based detections. It evolves into a dynamic, learning, and predictive model, capable of discerning even the stealthiest of cyber threats.

We provide a detailed introduction to the architecture of our NIDS, the rationale behind selecting Wireshark, and the innovative use of AI to analyze network traffic. This paper sets the stage for a discussion on the intersections of network security, data analysis, and AI, and how this synergy can yield robust solutions to the ever-growing challenges of cyber security.

II. LITERATURE REVIEW

The study conducted by [5] involved an examination of multiple novel intrusion detection methodologies, with a subsequent evaluation of their efficacy utilizing KDD Cup99 intrusion data. The models of Support Vector Machine (SVM) and Gaussian Process (GP) were taken into consideration for the purpose of intrusion detection. Subsequently, a hybrid model comprising SVM and GP was developed, alongside an aggregation technique that employed SVM, GP, and SVM-GP models as the principal means of classification. [6] et al. presented a comprehensive summary of the prerequisites and advantages associated with intrusion detection systems. This paper presents a comprehensive examination of intrusion detection systems, including their various types, lifecycles, geographical distributions, and attack modalities. In contemporary times, Intrusion Detection Systems (IDS) have emerged as a crucial component of safeguarding the security of businesses and network users. The implementation of security measures is regulated by IPS. Throughout the course of a life

cycle, distinct phases emerge and subsequently attain greater clarity. There exist additional obstacles that must be addressed. This particular methodology is executed with the express purpose of identifying irregularities and detecting instances of maltreatment, while alternative methodologies may also be employed. The study conducted by [7] highlighted the necessity of enhancing intrusion detection and prevention systems to enhance the security of computer networks. The unreliability of these tools, primarily attributed to the occurrence of false positives and false negatives, renders their usage challenging. Presently, it is evident that these systems are imperative in guaranteeing the security of enterprises. It is advisable to integrate multiple detection mechanisms to guarantee the security of the computing system. The Inertial Particle Separator (IPS) has been found to offer only limited mitigation of the aforementioned issues and is therefore deemed insufficiently robust for application in the construction industry. The authors in [8] explicate the function and influence of intrusion detection and prevention systems within network settings. The Intrusion Prevention System (IPS) is capable of identifying and intercepting harmful packets originating from malware, bots, viruses, and targeted attacks. Additionally, it possesses the ability to proactively mitigate network damage by preventing any further network activity.

III. WIRESHARK INFORMATION

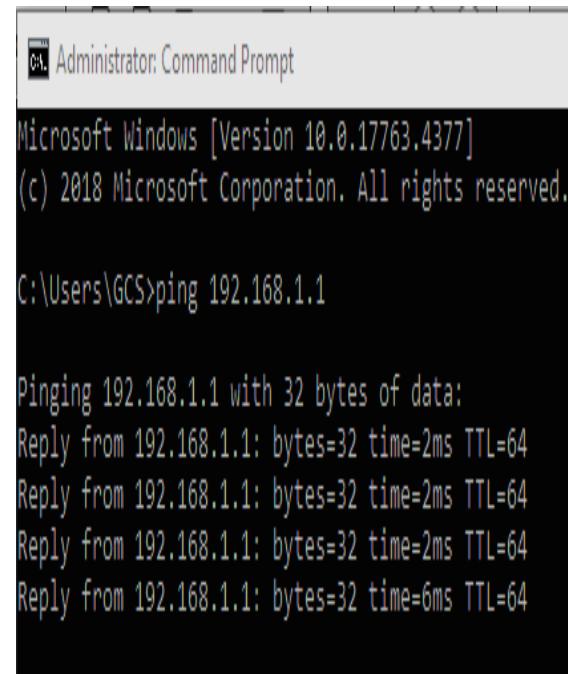
Wireshark comprises advanced functionalities that enable users to identify potential network intrusions [9] and suspicious behaviors. Presented below is an inventory of sophisticated functionalities of Wireshark.

Wireshark offers two distinct filter categories: capture filters and display filters. It is not possible to modify a capture filter. Display filters are employed based on specific criteria for the purpose of presenting packets in Wireshark. Display filters allow users to selectively view packets of interest while temporarily concealing those that are uninteresting. The I/O graph is a feature of Wireshark that provides a summary of packet flow. The I/O Graphs feature of Wireshark presents a visual representation of the aggregate network traffic contained within the captured data, typically quantified in terms of bytes or packets per second. There are colors defined for background of packets indicating suspicious packets in red color.

IV. RESULTS AND DISCUSSION

In order to attain the intended objective of the system, Wireshark and Snort are utilized for the creation of an intrusion detection system. According to [10], Snort is presently the most extensively utilized open source IDS system, and it is compatible with Windows operating systems. The aforementioned task is accomplished through the observation of inbound network traffic, which is commonly transmitted via LANs and Wi-Fi that links the server to the network hub. The traffic is cross-referenced with a repository of established signatures and irregularities associated with attacks, and notifications are generated upon identification of a breach. The utilization of signatures is prevalent in contemporary cybersecurity practices, as they are designed to counteract emerging forms of hacking and keep pace with updates to

exploit databases. The Network Intrusion Detection System (NIDS) [11] is a system designed for monitoring network intrusions. The process of intrusion detection is executed utilizing analogous methodologies to those employed in signature pattern matching and anomaly detection. The technique of error detection involves the monitoring of regular network operations to identify any atypical network behaviors, such as abrupt spikes in the volume of network traffic (measured in IP packets per second). The technique of signature pattern comparison involves the comparison of network data with pre-existing attack techniques stored in a database.



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Administrator: Command Prompt
Microsoft Windows [Version 10.0.17763.4377]
(c) 2018 Microsoft Corporation. All rights reserved.

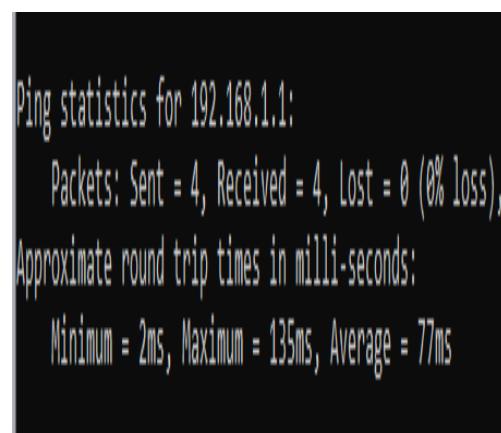
C:\Users\GCS>ping 192.168.1.1

Pinging 192.168.1.1 with 32 bytes of data:
Reply from 192.168.1.1: bytes=32 time=2ms TTL=64
Reply from 192.168.1.1: bytes=32 time=2ms TTL=64
Reply from 192.168.1.1: bytes=32 time=2ms TTL=64
Reply from 192.168.1.1: bytes=32 time=6ms TTL=64

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Fig. 1. Server is being tested for intrusion detection

The testing of server for processing is shown in figure1.



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Ping statistics for 192.168.1.1:
    Packets: Sent = 4, Received = 4, Lost = 0 (0% loss),
    Approximate round trip times in milli-seconds:
        Minimum = 2ms, Maximum = 135ms, Average = 77ms

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Fig. 2. No loss of packets at 192.168.1.1 address

Figure2 shows the sending and receiving of packets for intrusion detection systems. It means no loss of packets is seen at IP address 192.168.1.1

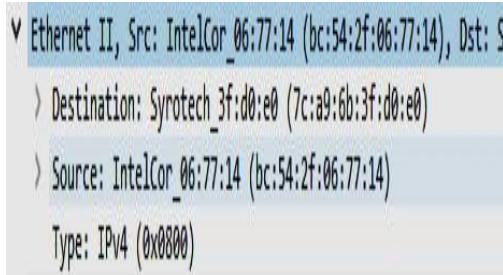


Fig. 3. Information sent on the network

Figure 3 shows the type of information that is being sent on network. It comprises type of protocol used, destination address and source address.

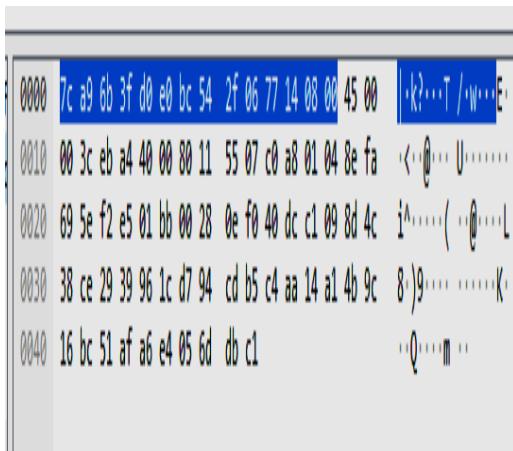


Fig. 4. Information received on the network

Figure4 shows packets and their codes in networking language.

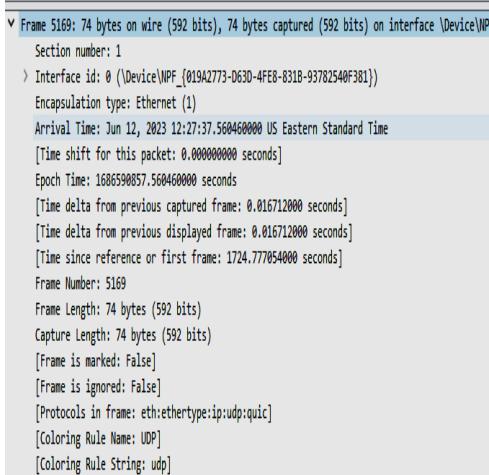


Fig. 5. Intrusion information over LAN

Figure 5 shows brief information over network such as section number, interface id, encapsulation type used, arrival time, epoch time, frame length, capture length and time framework for packets.

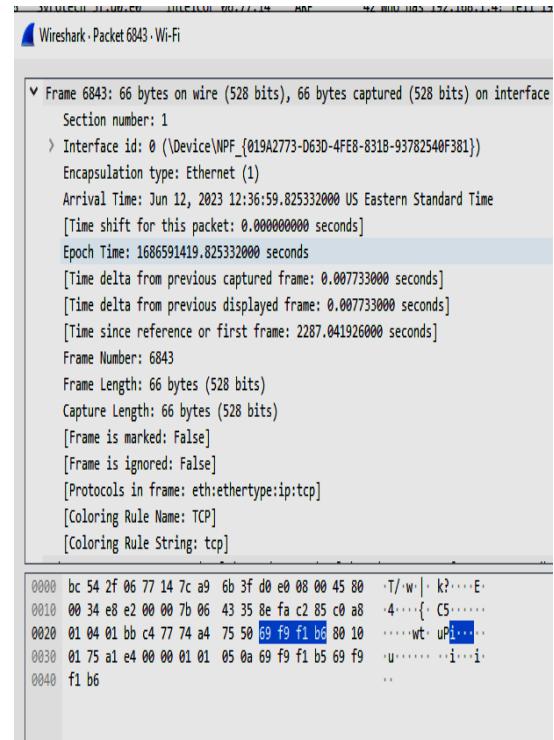


Fig. 6. System captures packets and informs user about suspicious packets

In figure6, the packets are captured live by the system and the user is informed about suspicious or malicious packets while capturing of packets through router over LAN.

| | | | | |
|-------------------|----------------|----------------|------|--|
| 7412 2565.914549 | 192.168.1.1 | 192.168.1.4 | ICMP | 138 Destination unreachable (Port unreachable) |
| 7413 2567.048811 | 192.168.1.4 | 172.217.161.10 | UDP | 71 64157 + 443 Len=29 |
| 7414 2567.056611 | 172.217.161.10 | 192.168.1.4 | UDP | 67 443 + 64157 Len=25 |
| 7415 2567.423178 | 192.168.1.4 | 192.168.1.1 | NBNS | 110 Refresh NB DESKTOP-653LEKG<00> |
| 7416 2567.426540 | 192.168.1.1 | 192.168.1.4 | ICMP | 138 Destination unreachable (Port unreachable) |
| 7417 2568.938208 | 192.168.1.4 | 192.168.1.1 | NBNS | 110 Refresh NB DESKTOP-653LEKG<00> |
| 7418 2568.941410 | 192.168.1.1 | 192.168.1.4 | ICMP | 138 Destination unreachable (Port unreachable) |
| 7419 2570.453523 | 192.168.1.4 | 192.168.1.1 | NBNS | 110 Refresh NB DESKTOP-653LEKG<00> |
| 7420 2570.4590188 | 192.168.1.1 | 192.168.1.4 | ICMP | 138 Destination unreachable (Port unreachable) |
| 7421 2571.968684 | 192.168.1.4 | 192.168.1.1 | NBNS | 110 Refresh NB DESKTOP-653LEKG<00> |
| 7422 2573.483720 | 192.168.1.4 | 192.168.1.1 | NBNS | 110 Refresh NB DESKTOP-653LEKG<00> |
| 7423 2574.785742 | 192.168.1.1 | 192.168.1.4 | ICMP | 138 Destination unreachable (Port unreachable) |

Fig. 7. Suspicious packets as identified during intrusion

The highlighted packets shown in figure 7 depict malicious or suspicious packets where intruders try to access the system.



Fig. 8. Expert analysis by intrusion detection system showing warning for the suspicious packets

Figure 8 shown above uses expert analysis algorithm to predict warning messages regarding suspicious packets in the network.

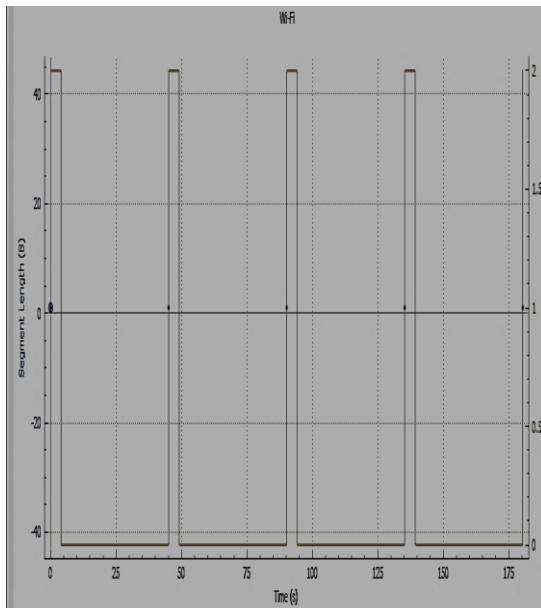


Fig. 9. Segment length vs. Time graph for 192.168.1.1

In figure 9, a graph is shown depicting segment length of packets in y-axis over a particular time in x-axis. It is done for network with address 192.168.1.1

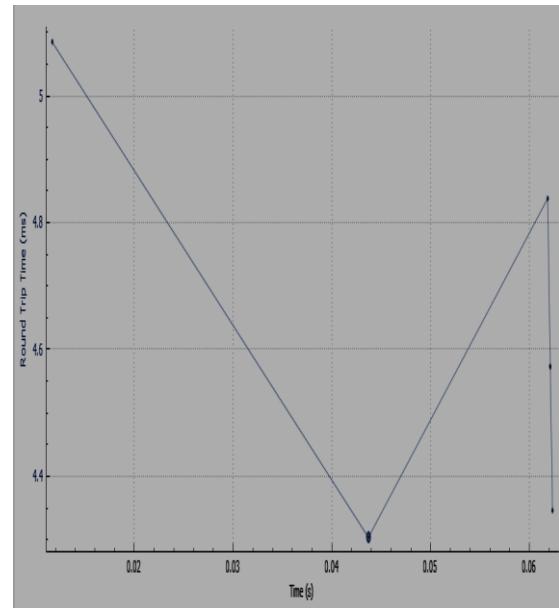


Fig. 10. Round trip time in response to figure 6

In figure10, calculation of round trip time over time intervals is depicted for packets captured in figure6 by the network.

V. CONCLUSION & FUTURE WORK

A very important tool is Wireshark, which offers a wide range of sophisticated capabilities including display filters, I/O graphs, color coding, and expert information. The features stated above may be effectively used to identify different kinds of network intrusions on computer systems. Through the use of the packet analyser program Wireshark [12, 13] , the study developed a novel framework for collecting data packets in real-time and detecting potentially harmful behaviors or packets. With the present approach, real data packets from the network were captured using Wireshark and manually mapped to the attack pattern dictionary. [14-18]

The model is used to identify different types of network intrusions. The findings of the current inquiry imply that Wireshark has a wide range of capabilities, functioning not only as a tool for packet analysis and problem-solving but also as an effective method for spotting unwanted access attempts. Using Wireshark, a person with knowledge of data packets, protocols, and related topics may find illegal access.

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