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# The Effectiveness of Snort in DDoS Attack Scenarios

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## Abstract

Secure communication in the digital era has become important, while on the other side, the attackers are consistently moving into more sophisticated ways to acquisition their target. DDoS was one of the common techniques used with the aim to overwhelm and make their target fail to function. To detect and limit this threat, many organizations have begun to employ various solutions. However, it was discovered that several solutions are ineffective while others require high cost for the implementation, and this has become a challenge for medium to low-sized organizations to meet their business strategy. For that, in this study we aim to introduce Snort as an open-source solution for the network detection and prevention system to determine how it undertakes and performs analysis especially in terms of accuracy and speed in relation to DDoS attacks. This study will involve a few steps: first, to establish the simulation environment; secondly, to perform the simulation with a DDoS attack to allow Snort to capture the traffic and respond according to the pre-established rules; and finally, to measure and evaluate the result. Snort as an open-source product, in nature allows the public to contribute, and that becomes the advantage compared with other commercial products especially in detecting anomalies. This will help the administrators to react more quickly by having an accurate information with an earlier warning system. Additionally, Snort is affordable, making it good choice for the organization.

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## 1. Introduction

In this digital era, the internet has become an essential part of communities. With the rapid development of the internet, make it suitable to be accessed from everywhere through various platforms and devices, and this creates opportunity for the perpetrators to gain advantage through the network for their criminal activities, making it very likely to become insecure, and this is not only an effect on the individual users but even more on a bigger scale, such as an organization or business, thus it becomes necessary to have an adequate protection on the network.

Today network security is essential for safeguarding data and systems, preserving data availability and integrity, and guaranteeing privacy and data protection (Iftikhar et al., 2023; Nandy et al., 2024a). In order to safeguard their data and systems from cyberattacks and other security risks, businesses must pay close attention to network security as technology advances and the number of cyberthreats consistently increasing.

Various types of attacks have been demonstrated by the attacker that resulted in huge losses to their target. Distributed Denial of Service (DDoS) attacks is one of the common methods that are widely used with the intention to create a serious disruption to the services or make a network system become a malfunction. On September 6 and 7, 2019, Wikipedia had a noteworthy DDoS attack that made the website unavailable in Germany and numerous other European countries (Cavanagh, 2019). Users reported having trouble accessing Wikipedia pages as a result of the severe service outages produced by this attack. This incident demonstrates how susceptible international information platforms are to hacks and how they affect users all over the world.

Derived from such situation, it has become essential to have an appropriate defense to detect DDoS using methods like Intrusion Detection Systems (IDS) or Intrusion Prevention Systems (IPS) (C. M. et al., 2024; Javanmardi et al., 2024). These systems will act as a major barrier prior to successfully entering the target network. IDS as a native solution has the main function of identifying threats within a network. This system monitors the traffic and forwards the alerts to the respective authorities for any suspicious activities detected. However, IDS have the drawback that they could not take immediate action to block the threats, therefore, manual intervention is still required to address the detected issues (Nandy et al., 2024). In contrast, IPS not only identifies the threats but is also able to automatically block them. IPS actively works to stop attacks before they can cause damage, thereby reducing the need for user intervention during detection. There are various solutions available in the market offering network protection with anti-DDoS as one of their key features, where this solution often requires a high cost of investment, which may not be suitable particularly for medium-low organization due to their resource's limitation (Iftikhar et al., 2023).

Snort, then introduced into the industry as an open-source software which is very well-known among cybersecurity professionals (Adiwal et al., 2023). It offers free tools with solid capability to analyze the network traffic and identify any suspicious packet according to the pre-defined signature, including DDoS. Snort is the world's leading open-source Intrusion Prevention System (IPS). Snort IPS uses a series of rules that help define harmful network activities and employs these rules to find matching packets and generate alerts for users.

With the above background, it leads this study to assess and analyze the accuracy of Snort, including the response time against such events, with the aim to offer alternative solutions that can be easily adopted not only for the network protection but also can be enhanced to provide a new experience towards the notification mechanism.

### **1.1 Literature Review**

Distributed Denial of Service (DDoS) assaults have been shown to corrupt a variety of internet systems by establishing botnets and infecting them with malicious agents or zombies. DDoS assaults have the potential to cause monetary losses, reduced productivity, and strained ties with clients and staff. Furthermore, extensive DDoS attacks might target IoT technology. In the previous study, the researchers tried to detect DDoS attacks using Deep Q-Network (DQN) (Purba et al., 2022). Consequently, DQN prioritized detection accuracy above in-depth analysis. Because Snort has advantages over Deep Q-Network (DQN) in DDoS detection, we used it to test a simulation. Snort employs rule-based detection, which enables it to rapidly evaluate attack patterns, whereas DQN necessitates lengthy and time-consuming training (Ndukwe & Nneka, 2024). Snort can instantly ascertain how long it will take to flag suspicious traffic because it is built to respond quickly to threats. On the other hand, because DQN depends on model evaluation, it could not be as accurate. Additionally, Snort implementation is simpler, so users can utilize it efficiently without requiring a deep understanding of machine learning. When taking these things into account, Snort provides a more efficient DDoS detection solution than the DQN-based method (Purba et al., 2022).

In another research, the implementation of an Intrusion Detection System (IDS) or Intrusion Prevention System (IPS) is typically embedded with features for real-time notifications to administrators, which can be delivered through emails, calls, or other communication channels such as a Telegram bot when specific events are identified. Snort sending SMS notifications has been covered in earlier studies, but there are a number of drawbacks. First off, sending a lot of notifications can get costly because SMS is frequently paid per message. Furthermore, the 160 character limit on SMS limits the quantity of information that may be sent. Because SMS communications are not encrypted and can be intercepted, security concerns also surface. Additionally, not all customers have limitless SMS plans, and mobile providers may mark certain SMS messages as spam (Komang et al., 2020).

Derived from the above references, this study is also intended to enhance the capability of Snort beyond DDoS protection, by integrating the event notification using Telegram as the communication platform. This service has a variety of benefits, such as being cost-free and enabling the delivery of several notifications without incurring extra charges. Telegram allows managers to obtain more detailed information because it has a high delivery rate and allows messages to be up to 4,096 characters long. The Telegram bot's real-time automated notifications improve danger response even further (Erlansari et al., 2020). Telegram also offers end-to-end encryption for enhanced security and convenient access from a variety of devices. Therefore, it has been shown that employing Telegram as a notification channel in the intrusion detection system is more successful and efficient in preserving network security.

## 2. Research Methods

In this study, we will conduct a simulation on the isolated network to minimize the impact. We employed Kali Linux to deliver the attack, Ubuntu with Snort installed to act as middleman to capture and analyze the logs, and Windows Server 2019 as the target. In this simulation we enabled the notifications by embedding the Telegram - API to send the notification during the events.

### 2.1 Building Environment

In the context of this study, we developed a simple network diagram as displayed in "Fig. 1,". To visualize the system architecture and data flow as well as the relationships between components.

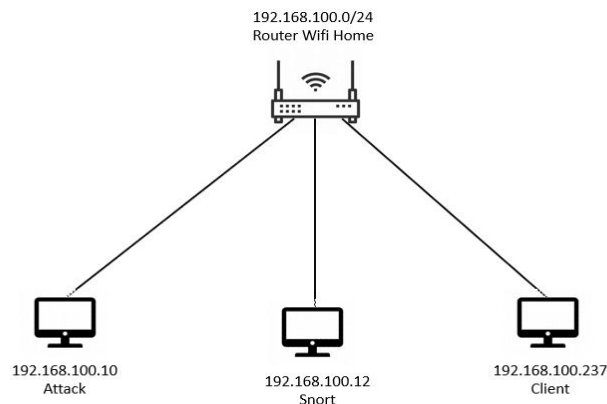


Fig 1. Diagram Network

As our simulation basis, various rules have been applied to maximize and gain a proper throughput from Snort. These rules are designed to specifically identify various patterns to match the flag defined and respond it accordingly. The rule header is a key component of every Snort rule, serving to define the basic characteristics of the traffic to be analyzed. Within this header, there are several important elements,

including the rule action, which determines the action to be taken when certain conditions are met, as well as the protocol that indicates the type of communication being monitored, such as TCP (Turner & Joseph, 2017). Furthermore, the source and destination sections in the rule header are also crucial as they indicate the source and destination IP addresses of the target analysis. This helps Snort determine where an attack may originate from and where it is targeted. After the header, there are rule options that provide further details on how data packets should be examined. These options allow users to set specific criteria, such as matching content in the payload or configuring other parameters relevant to DDoS attack detection. This explanation will provide deeper insights into how network monitoring systems can be optimized to protect infrastructure from cyber threats. Below is an example of a rule header using an alert “Fig. 2.”

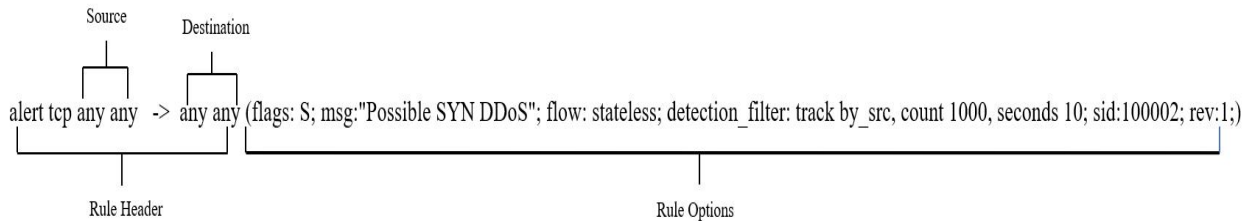


Fig. 2. Snort Rule for SYN DDoS

Alert rules are another important instrument used by Snort. These rules are designed to quickly and effectively identify and respond to abnormal conditions in network traffic “Fig. 3.”

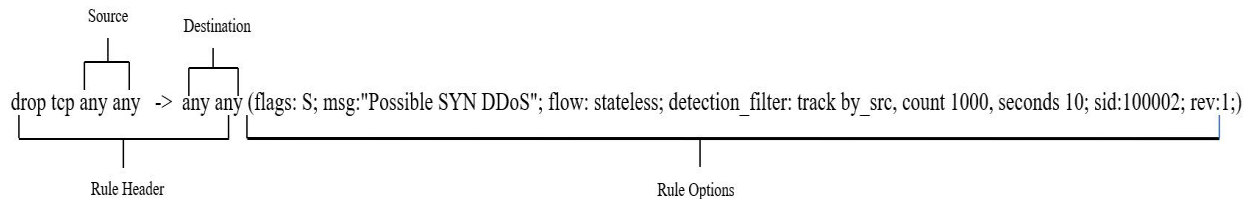


Fig. 3 Snort Rule for Drop SYN DDoS

In opposed to “Alert”, the implementation of “Drop” in Snort makes them behave as IPS, which can effectively prevent malicious packets from reaching their destination. This enables organizations not only to respond quickly to the threats but also to maintain the integrity and availability of their services. There are other rules that exist in addition to “Alert” and “Drop”, such as “Block”, which is used to stop the current packet and all subsequent unwanted packets in the stream. There is also a log for recording incoming packets or terminating the session with a specific protocol such as TCP reset or ICMP unreachable message (Pradipta, 2017).

## 2.2 Simulation

For the simulation, Snort will first be configured as an IDS, before switching to IPS to analyze the differentiation on how effectively it captures and react according to the predefined rules. When these rules are met, Snort will generate alerts and sending of messages using the Telegram bot. The notification messages will include important information such as the type of attack detected, the IP address of the attacker, and the time of the incident, allowing administrators to have a comprehensive information before the action. As visualized in “Fig. 4,” explaining how the data flow works, it begins when a packet enters the network, passes

through Snort for analysis to determine whether the traffic is legitimate or not. If not, it will block and send an alarm to the administrator.

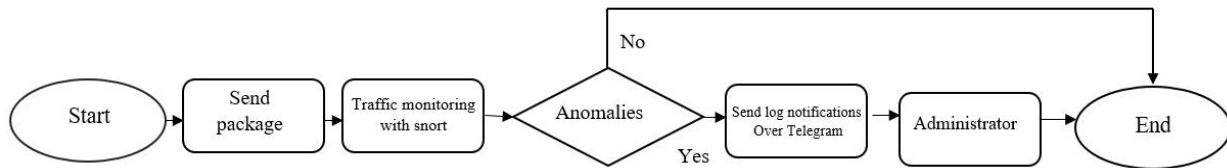


Fig. 4 Data flow from Snort to Telegram

### 2.3 Data Collection and Measurement

Snort is not only capable of logging traffic based on predefined rules but also can detect various network activities and anomalies. In this context, we present a traffic log diagram after running Snort in duration for one hour. This diagram visually illustrates the types of traffic detected and shows how Snort analyzes and categorizes network packets over time “Fig. 5.”.

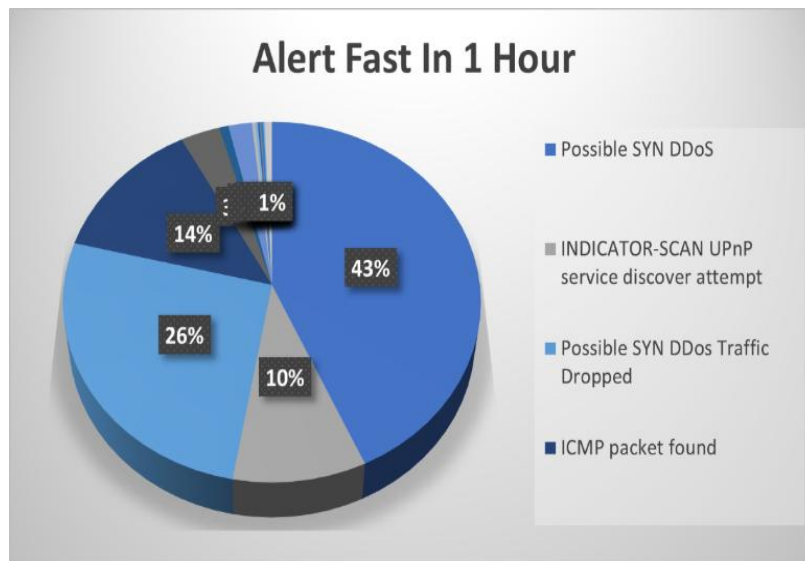


Fig. 5 Network Traffic

From the simulation result, the largest portion (43%) indicates a potential SYN DDoS attack, where the attacker sends numerous SYN packets in an attempt to open connections to the server without completing them, which can lead to resource exhaustion on the server. A total of 26% of the diagram shows SYN DDoS traffic that may have been blocked, meaning the detection system successfully identified and removed some suspicious SYN packets. Additionally, 14% indicates detected ICMP packets, which are often used for network diagnostics, such as ping commands. Suspicious ICMP packets may indicate network scanning activity or attacks. Finally, 10% represents attempts at UPnP service discovery through indicator scanning. UPnP allows network devices to communicate with each other without manual configuration, but it can become an attack vector if not properly secured.

We obtained this data while running Snort with the command `hping3 -S -p 80 -i u10 192.168.100.237` using the following rules “Fig. 6.”. After simulating numerous attacks and monitoring the log results, Snort indicates that it successfully achieved a 100% success rate in detecting and sending notifications, with an average time

of 2 seconds after the attack was detected. It is proven that Snort is capable of act within a short time, this can be influenced based on the characteristic of the attack performed and network architecture.

```
alert tcp any any -> any any (flags: S; msg:"Possible SYN DDoS"; flow: stateless; detection_filter: track by_src, count 1000, seconds 3; sid:100001; rev:1;)
```

```
alert tcp any any -> any any (flags: A; msg:"Possible ACK DDoS"; flow: stateless; detection_filter: track by_dst, count 1000, seconds 3; sid:100002; rev:1;)
```

```
alert tcp any any -> any any (flags: R; msg:"Possible RST DDoS"; flow: stateless; detection_filter: track by_dst, count 1000, seconds 3; sid:100003; rev:1;)
```

```
alert tcp any any -> any any (flags: F; msg:"Possible FIN DDoS"; flow: stateless; detection_filter: track by_dst, count 1000, seconds 3; sid:100004; rev:1;)
```

```
alert udp any any -> any any (msg:"Possible UDP DDoS"; flow: stateless; detection_filter: track by_dst, count 1000, seconds 3; sid:100005; rev:1;)
```

```
alert icmp any any -> any any (msg:"Possible ICMP DDoS"; detection_filter: track by_dst, count 1000, seconds 3; sid:100006; rev:1;)
```

Fig. 6 Snort Protocol Rules

### 3. Result and Discussion

The results of the DDoS simulation are visualized in this section, along with the abnormalities that were discovered by Snort that serving as an IDS and IPS. Overall, Snort has been demonstrated and proven effective in identifying various kinds of anomalies, particularly on DDoS. It also shows how the integration with Telegram can help to improve the network administration’s responsiveness to the threat.

Following “Fig. 7,” showing the alerts triggered from Snort as an IDS to generate the notification without dropping the traffic, along with the throughput display on Task Manager from the target machine before and during a DDoS attack. This is showing a clear result of changes in network activity by comparing the data collected before and during the attack as displayed on “Fig. 8.”

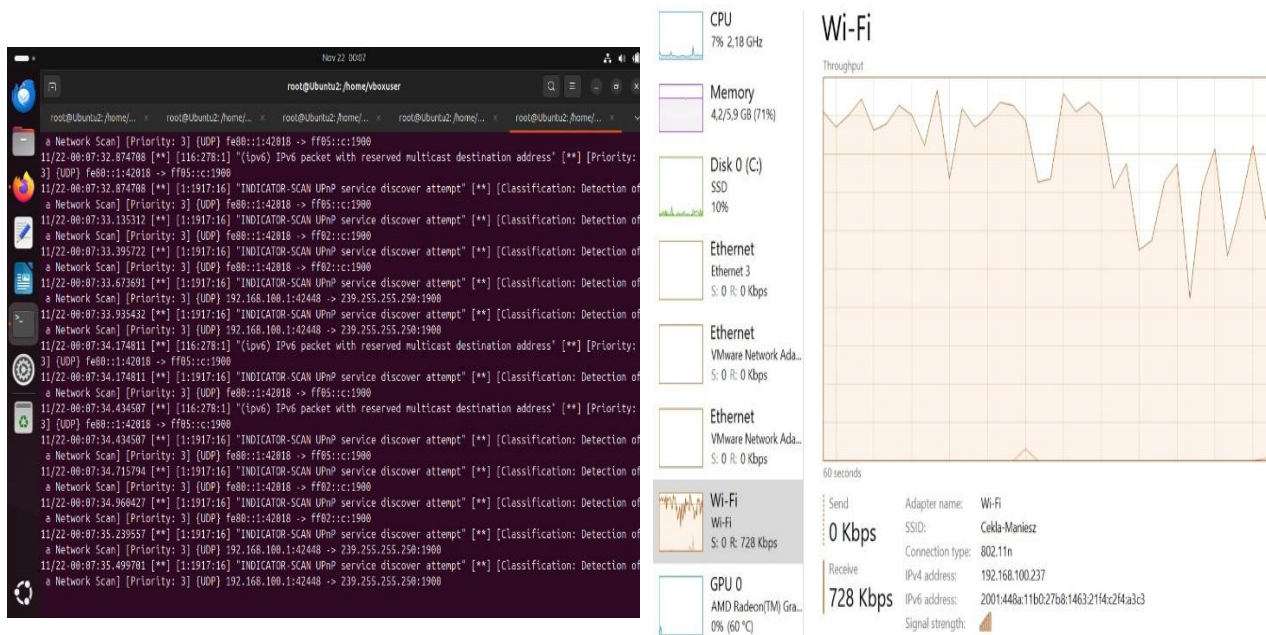


Fig. 7 Before SYN DDoS Attack

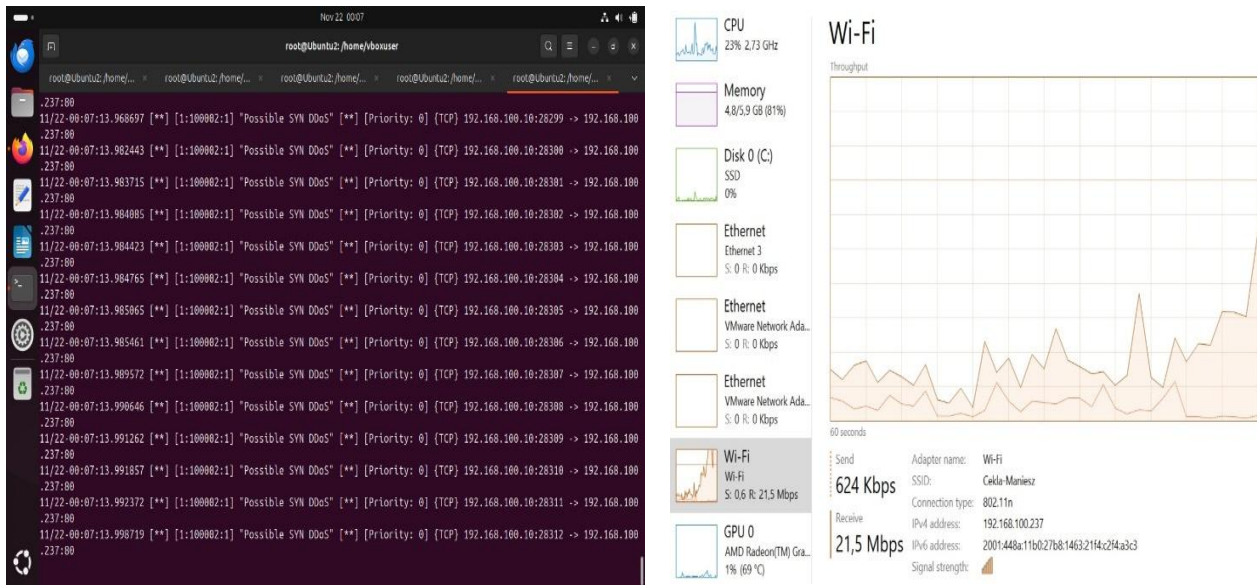


Fig. 8 After SYN DDoS Attack

In the second scenario, Snort is setup to function as an IPS. It shows the fact that in “Fig. 9” and “Fig. 10,” the graphic of the network performance on the target remains consistent without experiencing any spikes this indicates the capability of Snort to stop the malicious packets before reaching the target machine.

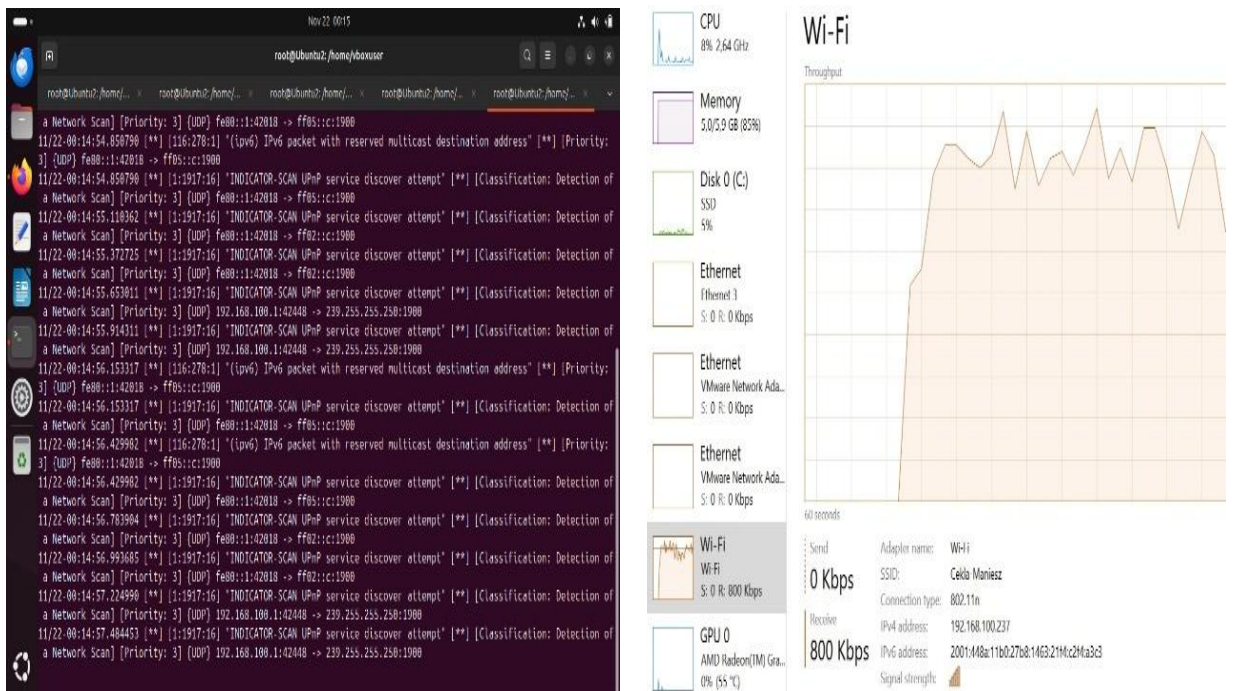


Fig. 9 Before SYN DDoS Drop

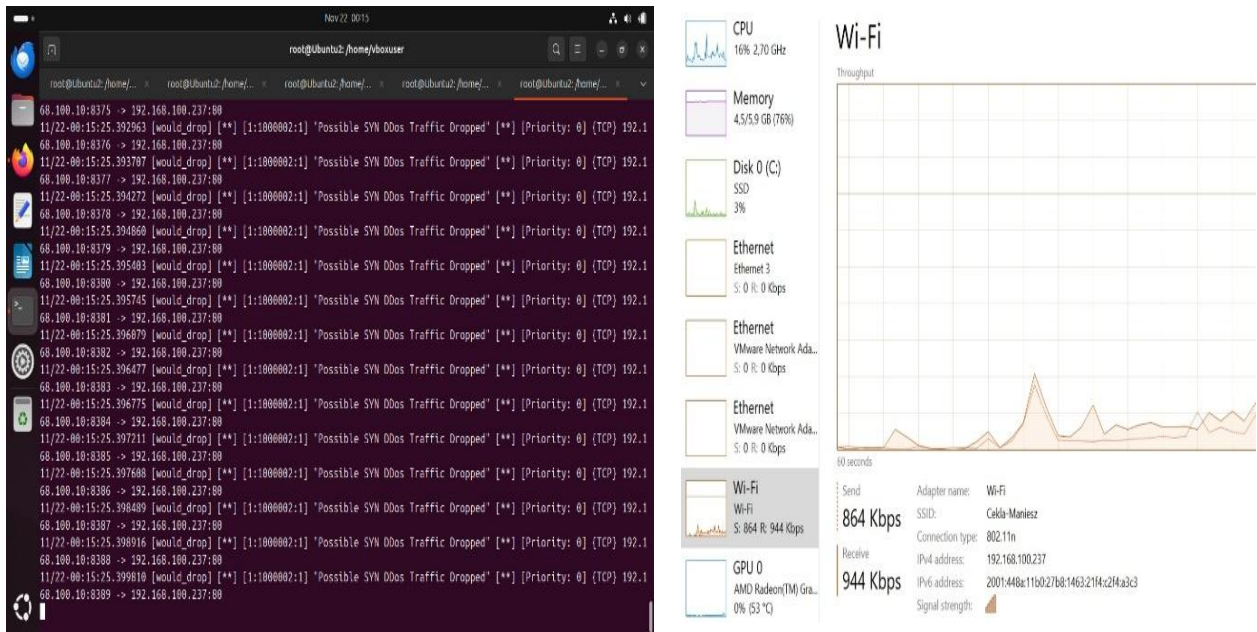


Fig. 10 After SYN DDoS Drop

Additionally, as illustrated in "Fig. 11," the system's notification allows to instantly alerting network administrators for suspicious activity or possible threats is demonstrated by the integration of Snort with a Telegram bot, for notification system.

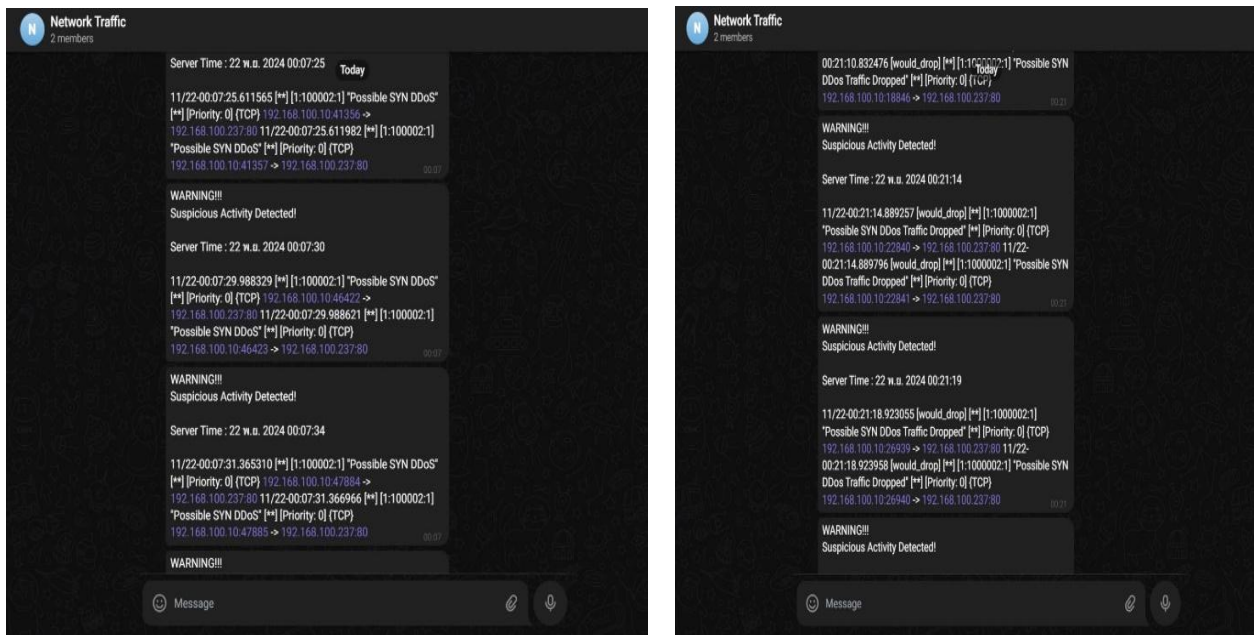


Fig. 11 Notification Result on Telegram



#### 4. Conclusions

This study, provide a simulation of using Snort to detect the suspicious packet over the network, mainly for DDoS. Based on the result obtained, it highlights Snort's capability not only to detect various types of attacks and its effective response, but also demonstrates how the Telegram-based notification system can be integrated to improve the network administrator's responsiveness to the potential threats. Thus, this conclusion emphasizes the importance of combining detection technology and communication to strengthen overall network security.

Due to the limited resources, the simulation only involved a limited number of servers and DDoS attacks, suggesting that larger-scale trials could be developed to ensure the validity of this result. Performance analysis, such as response time and overhead, could also be an area for further development to ensure the efficiency of this system.

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