

## **TASK 2**

### **Paper Title:**

Intrusion Detection System Based on Pattern Recognition

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## **1. Summary**

The paper presents an innovative approach to network security, focusing on the development of a sophisticated Intrusion Detection System (IDS) using machine learning and pattern recognition methods. Addressing the escalating challenges in network security due to the explosion of Internet of Things (IoT) and changes brought about by the COVID-19 pandemic, the paper introduces a two-layered IDS designed to enhance the detection and prevention of network intrusions. The first layer classifies network connections based on the service used, while the second layer identifies malicious activities using a minimal yet optimized set of features. This novel approach in this paper was that it introduced a new method for “distinguishing between normal and malicious network activities different from the traditional IDS models. The methodology involves the creation of two multivariate normal statistical models representing normal and attack behaviors, which are employed during the testing phase to classify network activities using a maximum likelihood estimation function. The paper's experimental results demonstrate the proposed IDS's superior performance over existing systems, achieving high detection rates and accuracy with minimal false positives, using only a limited set of features.

### **1.1 Motivation/Purpose/Aims/Hypothesis**

With the progress in network security and increasing internet usage, the paper aims to address the challenges in intrusion detection through a novel IDS leveraging machine learning and pattern recognition techniques. The hypothesis is that a two-layered IDS can more effectively identify network intrusions compared to traditional systems.

### **1.2 Contribution**

The paper's contribution as mentioned earlier is the introduction of a two-layered IDS that first classifies network connections based on the service used and then identifies malicious activities using optimized feature detection. This approach is expected to enhance the accuracy and efficiency of intrusion detection.

### **1.3 Methodology**

The methodology involves creating two multivariate normal statistical models during the training phase: one for normal behavior and another for attack behavior. These models are then used in the testing phase to classify network activities into either normal or attack categories. This classification is achieved using a maximum likelihood estimation function.

### **1.4 Conclusion**

The paper concludes that the proposed IDS exhibits superior performance in network intrusion detection compared to related systems. Using just four features, it achieves a detection rate (DR) of 97.5%, false alarm rate (FAR) of 0.001, Matthews correlation coefficient (MCC) of 95.7%, and an overall accuracy of 99.8%.

## 2. Limitations

### 2.1 First Limitation/Critique

A notable limitation of the study is the potential overfitting of the IDS model to specific network scenarios. This could limit the system's effectiveness in diverse or evolving network environments.

### 2.2 Second Limitation/Critique

The study may also face challenges in real-world applications due to the dynamic nature of cyber threats. The rapid evolution of malware and attack methodologies might render the two-layered approach less effective over time without continuous updates and learning.

## 3. Synthesis

The ideas presented in this paper have significant implications for the future of network security, particularly in the realms of IoT and remote work environments. The two-layered IDS approach, emphasizing machine learning and pattern recognition, could pave the way for more intelligent and adaptive security systems. These systems could be designed to learn from ongoing network activities, thereby staying ahead of evolving cyber threats. Moreover, the model's high accuracy and low false alarm rate highlight its potential for integration into existing security frameworks, offering enhanced protection without overwhelming network administrators with false positives. However, due to eccentric nature of threat and malware attacks, adapting and updating this model to keep pace with the ever-changing landscape of cyber threats will be crucial, potentially involving real-time data analysis and incorporating new machine learning algorithms to detect novel attack vectors.