



# Binary classification of malware by analyzing its behavior in the network using machine learning



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## Introduction

- The International Telecommunications Union (ITU) develops the global cybersecurity index (GCI). In the last GCI published in 2020, Honduras performs last place in the America region and is placed in the 178th position out of 182 countries with a punctuation of 2.2 out of 100 (Index, 2020).
- This research work is based on the premise that any new variant of malware behaves similarly to that of its predecessor [1], together with the fact that most malware communicates with external hosts [2].
- The proposed model bases its operation on the behavior-based approach at the data network level [2]. This model parses files containing frames and packets captured from the network, known as PCAP files.

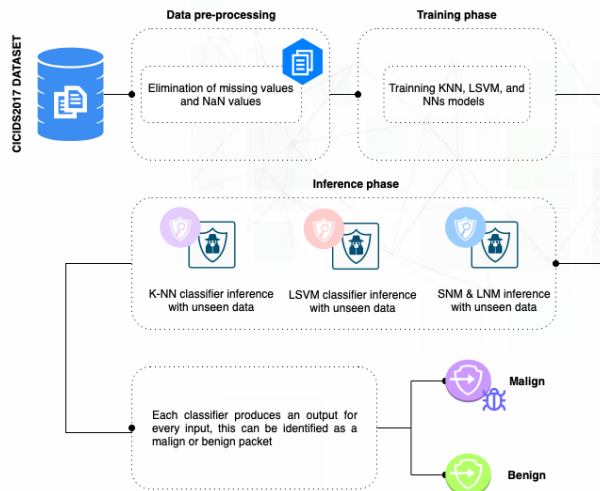


Figure 1 - Methodology diagram

## Objective

Develop a model to analyze the information in PCAP files and classify network connections as either benign or malicious (malware generated).

## Methodology

- The process uses two methods: (a) traditional machine learning algorithms (K-NN and SVM) and (b) various deep neural networks models. Both methods are trained and tested using the CICIDS2017 dataset (Figure 1).
- Neuronal network (NN) model architectures (Figure 2) used the cross-entropy loss function BCEWithLogitsLoss and SGD as the optimization algorithm. Both models use sigmoid as the activation function.

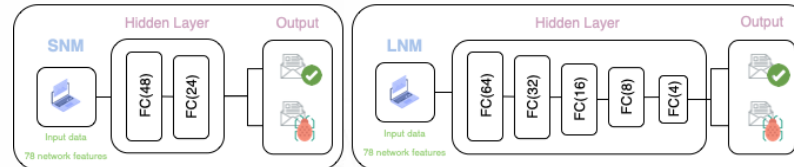


Figure 2 – Neuronal network models architecture

## Conclusions

- A process capable of classifying network connections as either benign or malicious (malware-generated) was successfully developed achieving above 95% in the recall, and f1 metric, attained using only traffic information from the data network.
- The method with the best performance without taking execution time into consideration was K-NN of method (a) with non-standardized and standardized data.
- The choice of the best performance was made based on the recall and f1 due to the classifier's context.
- The impact of data standardization had a positive impact on the evaluation metrics. In consequence, the use of data standardization is strongly recommended.

## Conflict of interests

The author makes known that there are not conflict of interest.

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Figure 3 - Best classifiers comparison using standardized data.



Figure 4 - Best classifiers comparison using non-standardized data.

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## Bibliography

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