

Network Intrusion Detection Using AI/ML

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Introduction

- Wireless Sensors Network is a network of tiny nodes to observe the environment.
- Any type of unwanted or unusual activity is called an intrusion.
- These intrusions can steal the sensitive information.
- Intrusions are the significant threat to security of the network and the sensitive data of the users.

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Problem Statement

- Cybersecurity threats are increasing day-by-day, creating challenges to the traditional intrusion detection techniques.
- Machine learning models like Random Forest, Decision Trees, k-Nearest Neighbors can be used to enhance the detection accuracy.
- This project aims to build a model which can detect the intrusions efficiently.

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Literature Review

Algorithm	Year	Reference	Result	Limitation
kNN	2014	W. Li, P. Yi, Y. Wu, L. Pan, J. Li	98.5%	Sensitive to the value of 'k'
SVM, kNN	2017	M. M. U. Chowdhury, F. Hammond, G. Konowicz, C. Xin, H. Wu, and J. Li	94.62%	Computational requirements are very high
Deep Learning	2019	R. Vinay, M. Alazab, K. P. Soman, P. Poorna, A. Al-Nemrat	88.7%	Requires high computational power and large dataset
kNN and sine-cosine algorithm	2021	J.-S. Pan, F. Fan, S.-C. Chu, H.-Q. Zhao, and G.-Y. Liu	99.34%	Not good for higher dimensional data
SVM and Firefly algorithm	2024	M. Karthikeyan, D. Manimegalai, and K. RajaGopal	99.34%	Not scalable, not useful for every intrusion detection

Table 1: Summary of Literature on intrusion detection



Methodology

- Reviewed literature on intrusion detection.
- Loaded and explored the NSL-KDD dataset.
- Data preprocessing, feature exploration and feature extraction using PCA for model development.
- Train the models like kNN, Random Forest, SVM etc.
- Model evaluation using metrics like accuracy, precision and recall.
- Performance comparison of the respective models.

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Result

- Random Forest achieves highest accuracy of 99.87% in detecting intrusions. It is equivalent to the decision tree with max_depth parameter equal to 3.
- k-Nearest Neighbors with the value of n_neighbors parameter equal to 20 achieves 98.94%.
- Support Vector Machine achieves the accuracy of 97.04%.
- Gaussian Naive Bayes achieves the accuracy of 91.61%.
- Logistic Regression achieves the accuracy of 87.17%.

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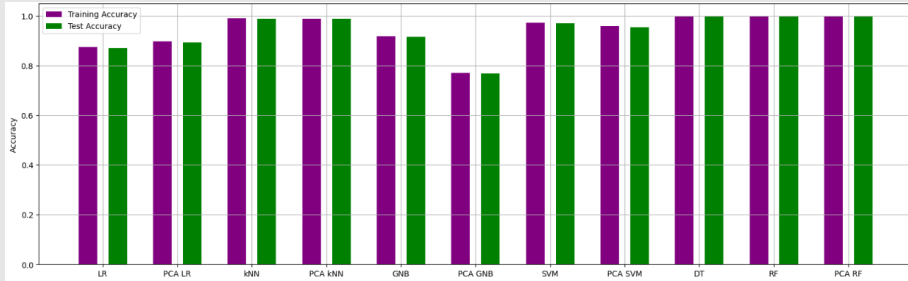
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Result (Continued)

The comparison between the respective results based on the accuracy is as follows:



Conclusion

- Machine learning algorithms like SVM, kNN, Random Forest are very important for intrusion detection for their capabilities of classification and prediction.
- The integration of AI and ML models enhance the ability to detect complex, unknown threats by learning patterns from large datasets, making intrusion detection systems more adaptive and intelligent over time.

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Future Work

- In future, we will try to deploy the model for real-time intrusion detection using Kafka.
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References

-  Panwar, Shailesh Singh, and Y. P. Raiwani. "Data reduction techniques to analyze NSL-KDD Dataset." Int. J. Comput. Eng. Technol 5.10 (2014): 21-31.
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Thank You!

