

# Week 2 Model Comparison Report

CyberIDS Research Team

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## 1 Objective

The goal of this study is to evaluate multiple binary classification models for network intrusion detection.

## 2 Models Evaluated

The following models were trained and compared:

- Logistic Regression
- Decision Tree (Pruned)
- Random Forest
- K-Nearest Neighbors (KNN)
- Support Vector Machine (SVM)

## 3 Evaluation Metrics

The following metrics were used to assess model performance:

- Accuracy
- Precision
- Recall
- F1-Score
- Cross-Validation Score

Model	Accuracy	Precision	Recall	F1-Score	CV Score
Logistic Regression	0.9997	0.9994	0.99999	0.9997	0.9996
Decision Tree (Pruned)	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>
Random Forest	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>
K-Nearest Neighbors	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>	<b>1.0</b>
SVM	0.9998	0.9997	0.99999	0.9998	0.9998

Table 1: Model Evaluation Metrics

## 4 Key Insights

- **High accuracy across all models** indicates robust classification performance.
- **Possible overfitting** in Decision Tree and Random Forest (achieving perfect scores).
- **Cross-validation confirms generalization** but requires further unseen data validation.

## 5 Future Improvements

- Investigate **overfitting risks** in Decision Tree and Random Forest.
- Perform **hyperparameter tuning** for KNN and SVM to optimize performance.
- Validate models on a **completely fresh dataset** for real-world effectiveness.

## 6 Conclusion

This comparison highlights the performance of different classification models for intrusion detection. Tree-based models achieve high accuracy, but their real-world reliability must be further tested. Logistic Regression and SVM provide strong recall, making them effective for detecting malicious activity.

### Next Steps:

- Fine-tune hyperparameters for KNN and SVM.

- Validate models with unseen datasets.
- Investigate feature importance for model optimization.

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