#### **ML Models for DDoS Attack Classification**

#### Introduction

This document explains the different Machine Learning (ML) models used for detecting and classifying DDoS attacks in Autonomous Electric Vehicle (EV) networks. The models implemented include Support Vector Machine (SVM), K-Nearest Neighbors (KNN), Logistic Regression (LR), and Naïve Bayes (NB).

### 1. Support Vector Machine (SVM)

SVM was the best-performing model, achieving **95.50% accuracy**. It works by finding an **optimal hyperplane** that maximizes the margin between classes.

### **Python Implementation:**

# 2. K-Nearest Neighbors (KNN)

KNN achieved 94.33% accuracy.

It works by classifying a data point based on the **majority class** of its k nearest neighbors.

#### **Python Implementation:**

```
from sklearn.neighbors import KNeighborsClassifier import joblib def
train_knn(X_train, y_train): knn_model =
KNeighborsClassifier(n_neighbors=5) knn_model.fit(X_train, y_train)
joblib.dump(knn_model, "models/knn_model.pkl") # Save model return
knn_model
```

# 3. Logistic Regression (LR)

Logistic Regression achieved 90.75% accuracy.

It is useful for **binary classification** tasks and predicts probabilities using a **logistic function**.

#### **Python Implementation:**

# 4. Naïve Bayes (NB)

Naïve Bayes achieved 82.06% accuracy.

It is based on **Bayes' theorem** and assumes independence between features, making it efficient for **probabilistic classification**.

### Python Implementation:

#### Conclusion

- SVM was the best-performing model with the highest accuracy.
- KNN performed well but can be computationally expensive for large datasets.
- Logistic Regression is effective for simple classification tasks.

| • | <b>Naïve Bayes</b> works well for probabilistic classification but assumes feature independence. |
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