AECR-ML FOR IOT ATTACK DETECTION IN EVS

Nguyễn Hữu Tài

¹ Trường ĐH Công Nghệ Thông Tin. ĐHQG TP.HCM

What?

We propose AECR-ML, a hybrid framework to detect cyberattacks on electric vehicle (EV) IoT systems.

- AECR extracts technical attack intelligence from CTI reports using a fine-tuned language model.
- ML models (e.g., Random Forest, Deep Learning) detect real-time anomalies from sensor data.
- Fusion improves detection accuracy and responsiveness.

Why?

- EVs are getting smarter but more exposed. As IoT grows in EVs, key systems like sensors and comms become prime targets for attacks (spoofing, DoS, MiTM), risking operation and safety.
- Existing solutions are reactive or narrow: Rule-based methods lack adaptability, and data-only models may miss new threats. Our model combines real-time sensor data and CTI insights for smarter, proactive defense.

Data Collection AECR-ML Performance Evaluation Hybrid AECR-ML Attack Detection Model Preprocessing & Normalization AECR: Threat Intelligence Extraction Real-time Anomaly Detection CTI Reports

Description

1. Data Collection

• Integrate IoT sensor data from EVs and intelligence from CTI reports to clean and normalize inputs for the model.

Details:

- Collect data from sensors (GPS, speed, environment,...) and CTI reports.
- Use AECR to extract attack knowledge from CTI reports.
- Normalize and synchronize data from both sources.



Figure 1 . Sensor data and collection

2. AECR-ML Model

This step focuses on designing and training a hybrid machine learning model that detects cyberattacks by leveraging both sensor data and intelligence from AECR.

- **Model training:** Apply machine learning algorithms (e.g., Random Forest, SVM, Deep Learning) to identify anomalies in EV sensor data.
- AECR feature integration: Enrich the detection model with behavioral features extracted from CTI reports, improving detection of unknown or zero-day attacks.
- Multi-source fusion: Merge sensor-based features and AECR-derived insights using feature fusion strategies.
- Optimization:
 hyperparameter
 cross-validation
 accuracy,
 precision,
 and
 generalization.

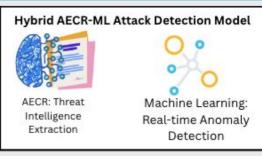


Figure 2. AECR-ML Model

3. Performance Evaluation

- Simulated Testbed: Build an EV IoT environment using Raspberry Pi and Arduino to emulate networks and sensors (e.g., CAN, MQTT).
- Attack Scenarios & Monitoring: Simulate spoofing, DoS, MiTM attacks and use the model for real-time anomaly detection.
- Model Evaluation: Measure detection accuracy, processing efficiency, and scalability across sensor types.