AI Revolutionizing Automotive Cybersecurity

With a focus on Intrusion Detection Systems

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Speaker Introduction



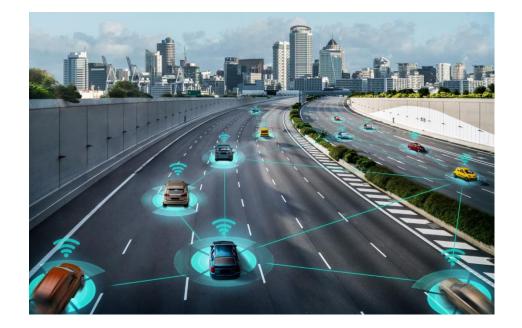
Natasha Alkhatib Cybersecurity Consultant at ETAS Bosch

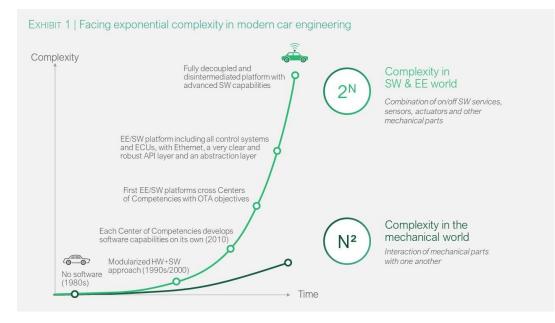
- Holds a Ph.D. in AI for Automotive Cybersecurity from Institut Polytechnique.
- Specializes in AI-powered automotive cybersecurity solutions.
- Holds a PhD in the field and extensive experience in developing in-vehicle intrusion detection systems.
- Consults for clients, implementing security concepts, integrating ETAS products like IDSs, firewalls, and fuzzing systems.



Motivation Double-edged sword: Automotive Connectivity & Complexity







Connectivity

Complexity

 \sim 80M vehicles with high or full automation by 2030

~ 367M Globally by 2027

Increased potential for safety-relevant attacks

Motivation UNECE WP.29 requirements

- The automotive sector is undergoing a profound transformation with the digitalization of in-car systems that are necessary to deliver vehicle automation, connectivity and shared mobility. This comes with significant cybersecurity risks.
- 2. The two UN regulations require that measures be implemented across 4 distinct disciplines to tackle these risks by establishing clear performance and audit requirements for car manufacturers:
 - Managing vehicle cyber security
 - Securing vehicles by design to mitigate risks along the value chain
 - Detecting and responding to security incidents across vehicle fleet
 - Providing safe and secure software updates and ensuring vehicle safety is not compromised, introducing a legal basis for O.T.A updated to on-board vehicle software.







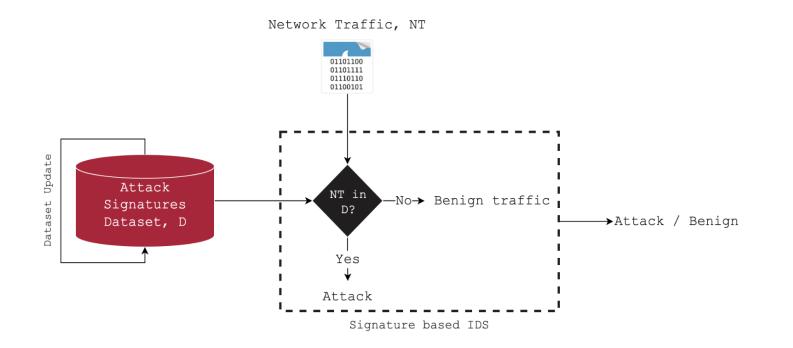
What is an Automotive IDS ?

- ISO/SAE 21434 standard mandates a defined incident response process.
- This process is bolstered by the automotive Intrusion Detection System (IDS), which functions as an in-vehicle sensor and connects to a backend system.
- The onboard IDS monitors the Electronic Control Units (ECUs) and communication networks for external attacks.
- Upon detection, it collects the data and transmits it to the manufacturer's Security Operations Center (SOC) for analysis.
- Based on the data collected by the automotive IDS, the OEM makes informed decisions on how to respond to these attempted attacks.



Automotive Intrusion Detection principles

Signature-based technique

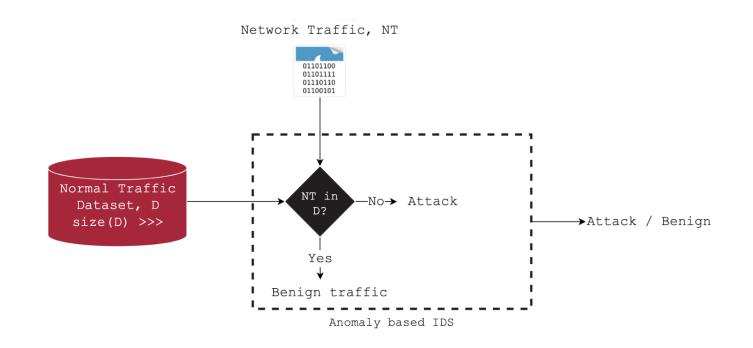


- S False positive rate
- Identify previously known attacks
- Fail to identify novel or previously unseen attacks



Automotive Intrusion Detection principles

Anomaly-based technique



 Capable of identifying novel or previously unseen attacks



Challenges for automotive IDS



Real-time constraints

Continuous Monitoring High Detection Rate Low False-Alarm Rate

Robustness



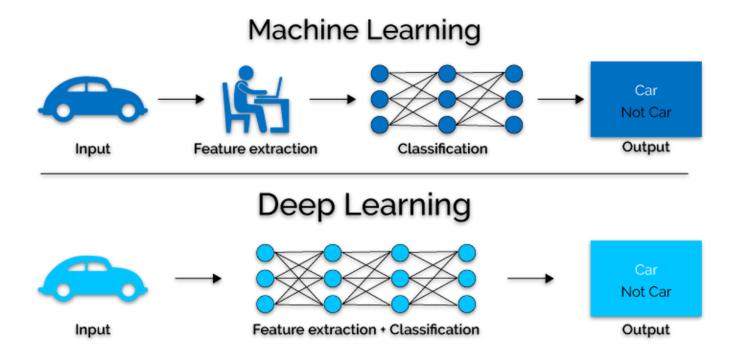
The game changer in automotive Intrusion Detection

Deep learning techniques for Intrusion Detection



Deep learning (DL) is a machine learning subfield that uses multiple layers for learning data from representations

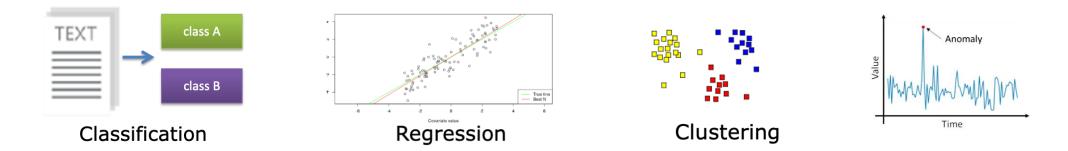
- DL is exceptionally effective at learning patterns



Picture from: https://www.xenonstack.com/blog/static/public/uploads/media/machine-learning-vs-deep-learning.png

Deep learning techniques for Intrusion Detection - learning types

- **1. Supervised:** learning with labeled data
 - Signature-based IDS
 - Example: regression for predicting real-valued outputs
 - Example email classification, image classification
- 2. Unsupervised: discover patterns in unlabeled data
 - Example: cluster similar data points
 - Example: Anomaly-based IDS
- 3. Semi-supervised/self-supervised: learn data representations with labeled data
 - Example: anomaly-based IDS





Intrusion Detection for Controller Area Network (CAN)

Broadcast data frames (current state of the vehicle)



J	(11 hits)	I R OLC	Data Field (0 to 8 Bytes)	I D E	CRC	АСК	E O F	l F S
0.05s	ID 0X102							
0.01s	ID 0X45D							

The Controller area network **Protocol definition**

Message-based protocol standard

_

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Node

X

120 Ω

Node

А

120 Ω

Node B

CAN Controller

CAN Transceiver

CAN H CAN L

Rx

CAN_H

CAN L

Tx

The Controller area network Vulnerabilities







No authentication Broadcast domain

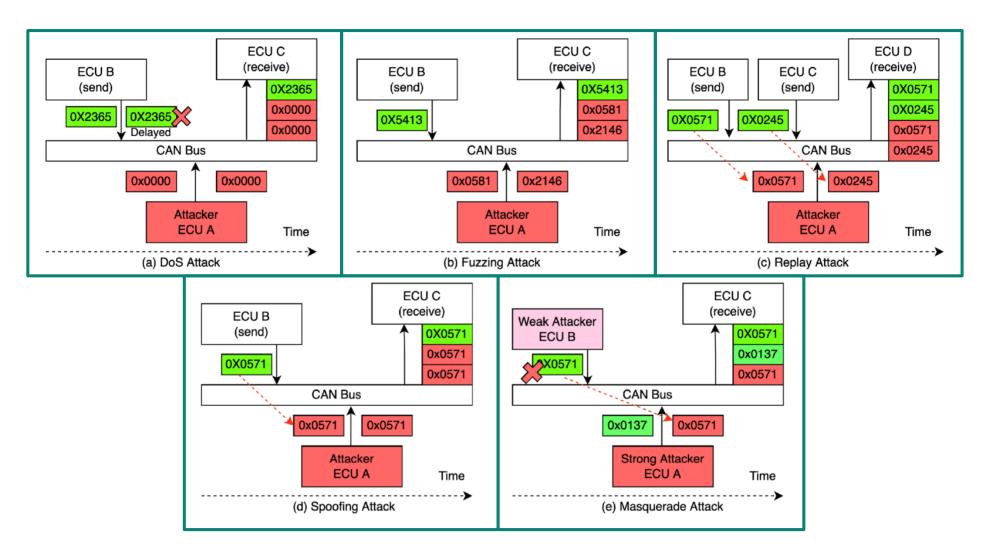


No encryption



ID-based priority

The Controller area network Attacks





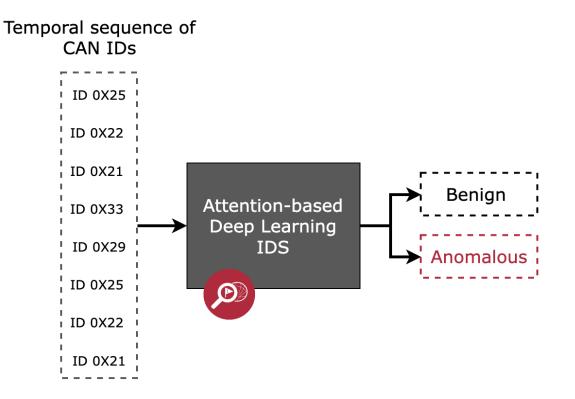


Research question

Can we exploit the structure of Attention networks for intrusion detection on CAN?

Motivations

- Model sequential CAN data
- Make each ID in the sequence encode the context information from both *left* and *right*
- Objective function for capturing normal sequence behavior:
 - Prediction of next ID message isn't enough
 - Self-supervised learning for whole sequence contextual encoding





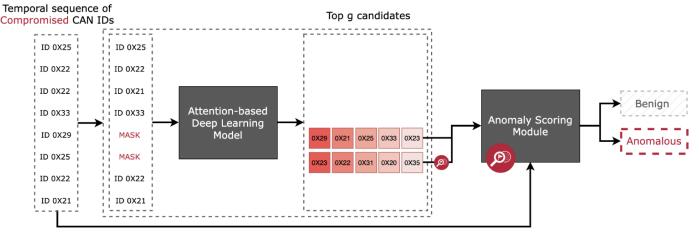
Methodology

How to exploit the structure of Attention networks for intrusion detection on CAN?

- Model trained on normal data → accuracy on predicting masked CAN IDs for normal test sample.
- Randomly **replace** a ratio of CAN IDs in a sequence with a specific **MASK** token
- Predict the **masked CAN IDs** in CAN sequence
- Anomalous CAN ID

 \rightarrow observed CAN ID is **not in the top-g candidate** set predicted by attention-based model

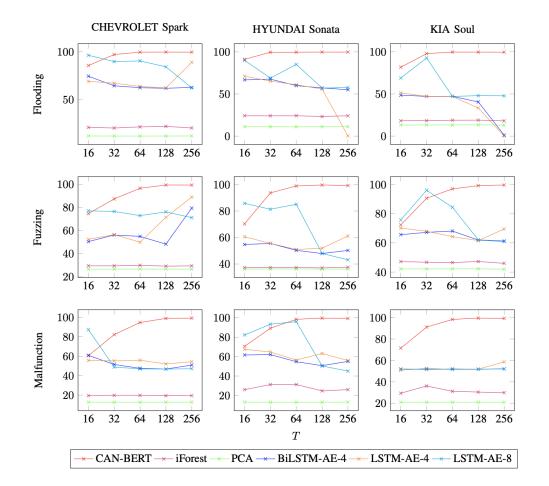
- Anomalous CAN ID sequence
 - \rightarrow > *r* anomalous CAN IDs



Anomaly detection stage



Comparison of the attention-based model with other baselines



- Sequence length variation: {16,32,64,128,256}
- **Target:** Identify a message injection attack as soon as possible
- Takeaways:
 - ML algorithms perform poorly and maintain the same F1- score metric w.r.t sequence length.
 - DL based models outperformed the traditional anomaly detection models w.r.t sequence length.
 - Attention-based obtains respectable F1 scores
 ∈ [0.85, 0.99]
 - BERT-based models are better at capturing the patterns of CAN ID sequences

Al as a force multiplier in automotive cybersecurity

etAs (infineon

Other use cases

- Al technologies for:
 - automotive threat intelligence in the VSOC
 - fuzzing and penetration testing
 - investigation
 - research
 - report

Conclusion

- Toward explainability and interpretability of deep learning-based IDS
- Quantization of deep learning-based IDS
- End-to-end framework for overall intrusion detection

Key Takeaways

- AI as a force multiplier for automotive cybersecurity
- Revolutionization in the automotive Intrusion Detection
 Field
- Benefits: Improved detection rates, faster response times, proactive defense
- Considerations: Data security & privacy, explainability of AI models



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Upcoming Webinar: Securing the automotive industry in the quantum computing eraDate & Time: 23. October 2024, 9am (EST)

