

Scission: Signal Characteristic-Based Sender Identification and Intrusion Detection in Automotive Networks

1st ITG Workshop on IT Security (ITSec)

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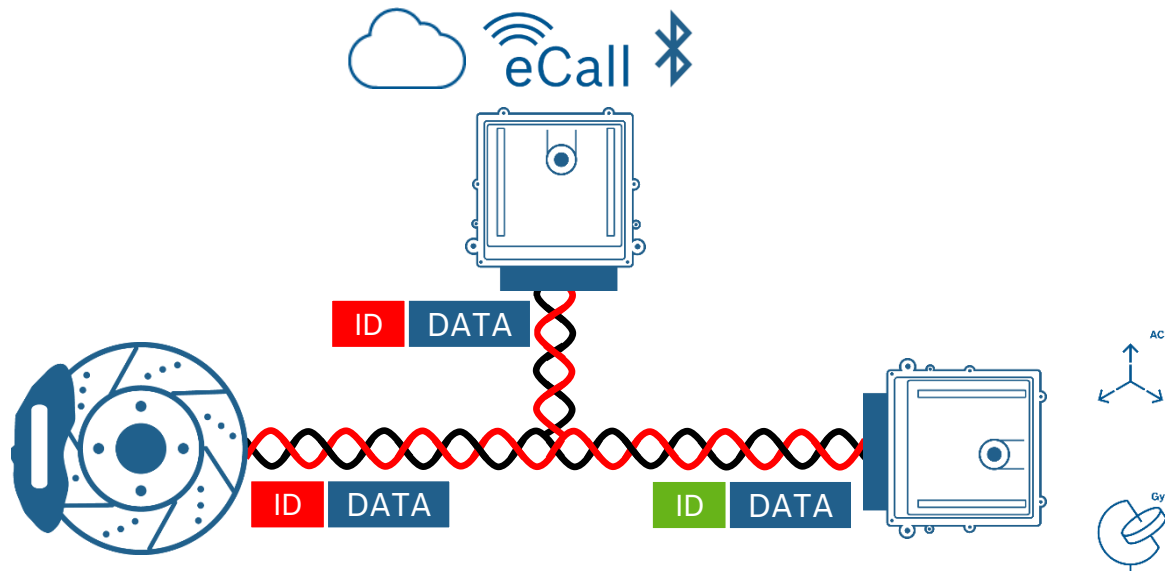
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⁽²⁾ Robert Bosch GmbH



Introduction

- ▶ Attacks on vehicles...
 - ▶ on the rise due to increased connectivity features
 - ▶ may be highly scalable
 - ▶ result in threats for humans and the environment
- ▶ Demonstrated by Miller and Valasek [31]

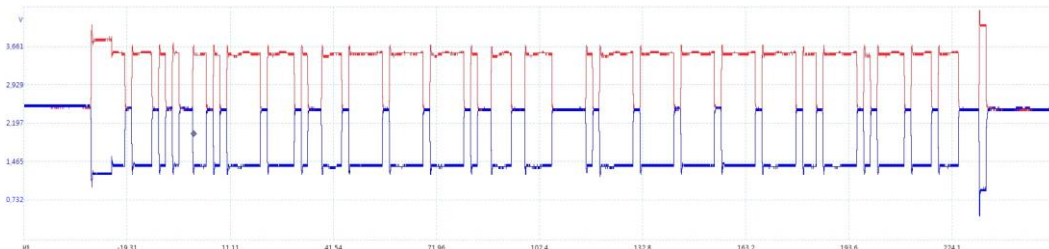


[31] Charlie Miller and Chris Valasek. 2015. Remote exploitation of an unaltered passenger vehicle. Black Hat USA 2015

Introduction

▶ Controller Area Network widely used for in-vehicle communication

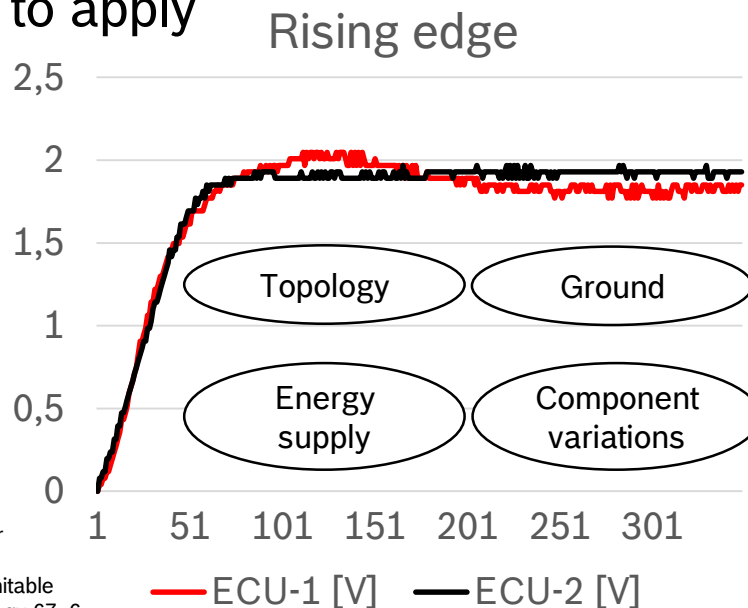
- ▶ 500 kb/s bandwidth
- ▶ 64 bit payload
- ▶ No sender authenticity



▶ Message Authentication Codes hard to apply

▶ Intrusion Detection Systems

- ▶ Signatures
- ▶ Anomalies
- ▶ Physical properties
 - Clock drifts [4]
 - **Variations in the analog signal [33, 6]**

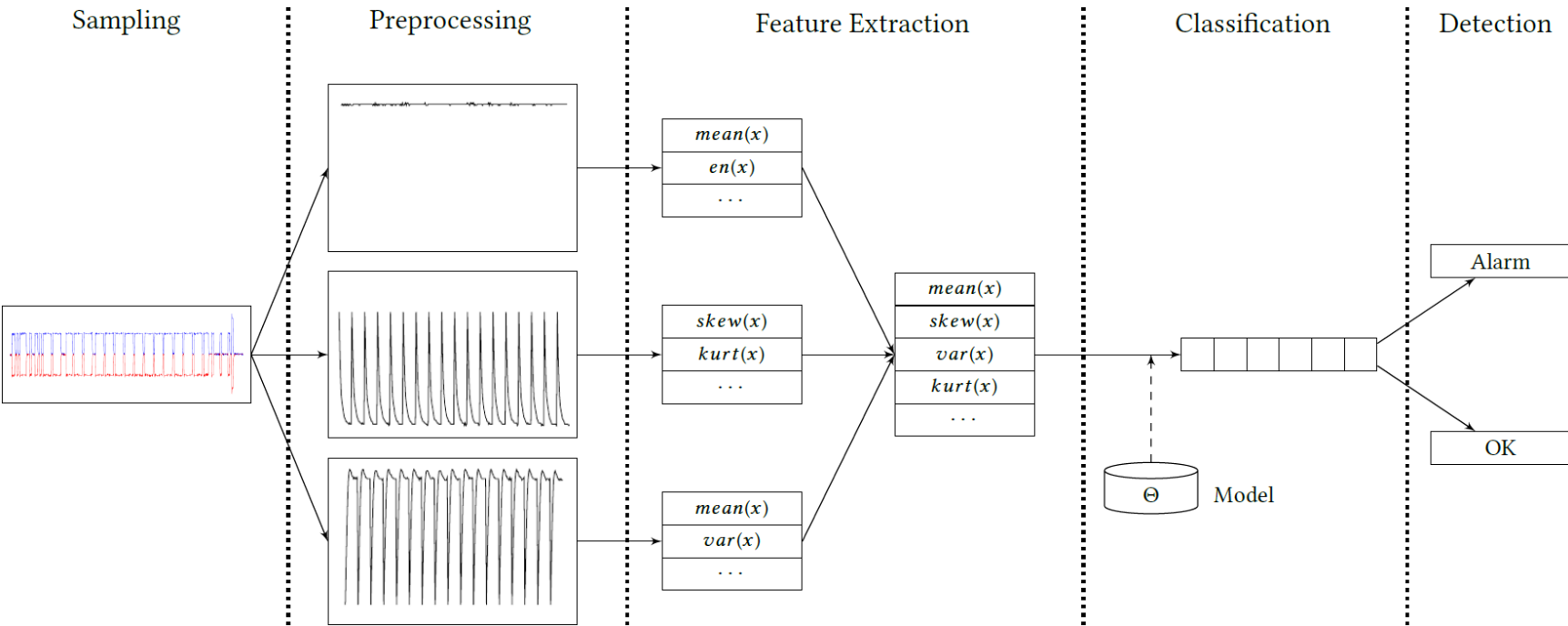


[4] Kyong-Tak Cho and Kang G. Shin. 2016. Fingerprinting Electronic Control Units for Vehicle Intrusion Detection. In 25th USENIX Security Symposium.

[33] P. S. Murvay and B. Groza. 2014. Source Identification Using Signal Characteristics in Controller Area Networks. IEEE Signal Processing Letters 21.

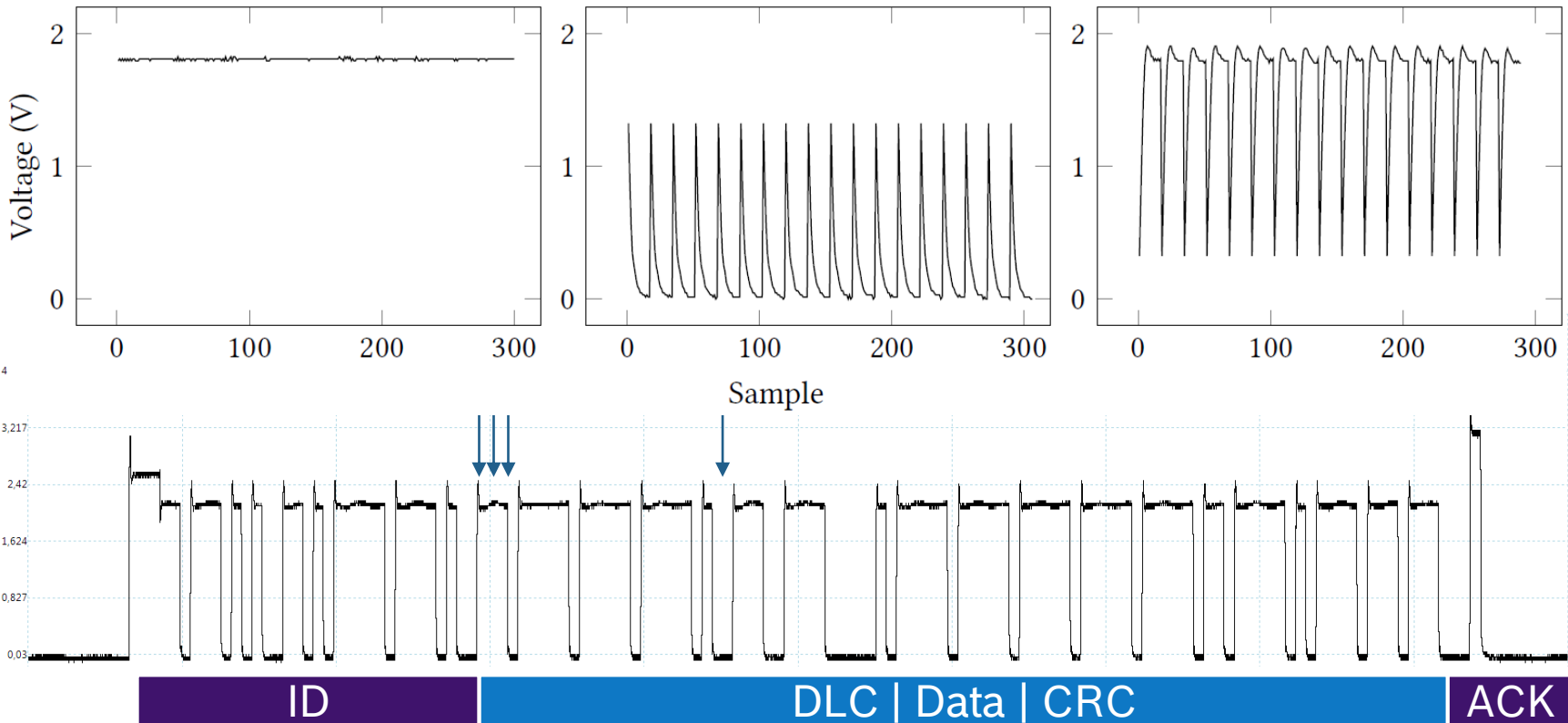
[6] W. Choi, H. J. Jo, S. Woo, J. Y. Chun, J. Park, and D. H. Lee. 2018. Identifying ECUs Using Inimitable Characteristics of Signals in Controller Area Networks. IEEE Transactions on Vehicular Technology 67, 6.

Scission Overview



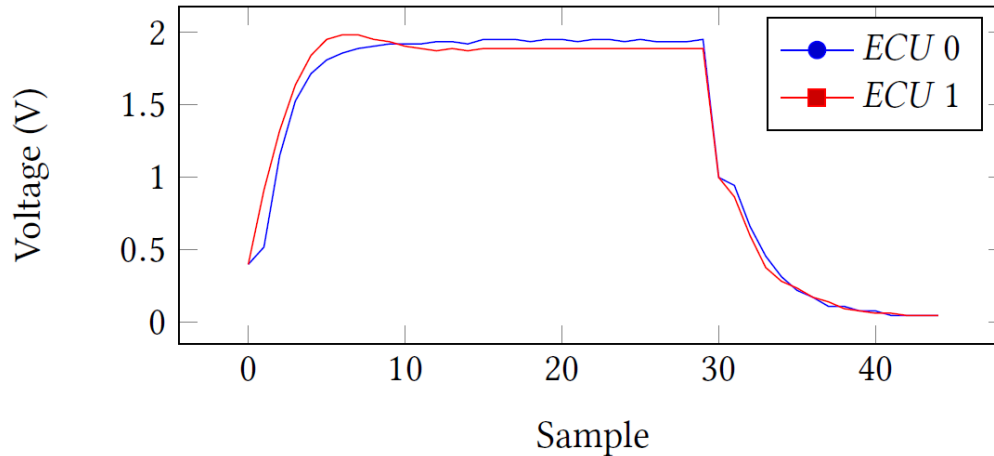
Sampling and Preprocessing

- ▶ Sampling differential signal (20 MS/s)
- ▶ Cluster symbols based on its signal shape



Feature Selection

- ▶ Statistical features (time, frequency) individual for each group
 - ▶ Mean, Standard Deviation, Variance, Skewness, ...



	Concatenated	Rising	Falling	High
ECU 0	1.286 V	1.623 V	0.289 V	1.947 V
ECU 1	1.285 V	1.691 V	0.275 V	1.890 V
Difference	0.001 V	0.068 V	0.014 V	0.057 V

Model Generation and Classification

▶ Logistic Regression

ECU 0	ECU 1	ECU2
95 %	3 %	2 %

▶ Supervised learning with 200 frames per ECU

▶ Initial training in safe environment

- ▶ Initiated by secure diagnostic access
- ▶ Key between ECUs and Scission assigned

▶ Performance Monitoring (aging, corrosion, ...)

- ▶ Probabilities of each ECU
- ▶ Online adaption of the classifiers
- ▶ MAC supported adaption/learning
 - AUTOSAR Secure Onboard Communication (SecOC)

Intrusion Detection

- ▶ Sender identification based on the highest probability

ECU 0	ECU 1	ECU2
95 %	3 %	2 %
2 %	98 %	0 %
49.9 %	50.1 %	0 %

- ▶ In-vehicle communication is static

- ▶ Each identifier is used by only one ECU
- ▶ Alarm if an identifier is used by a invalid ECU

- ▶ False positives

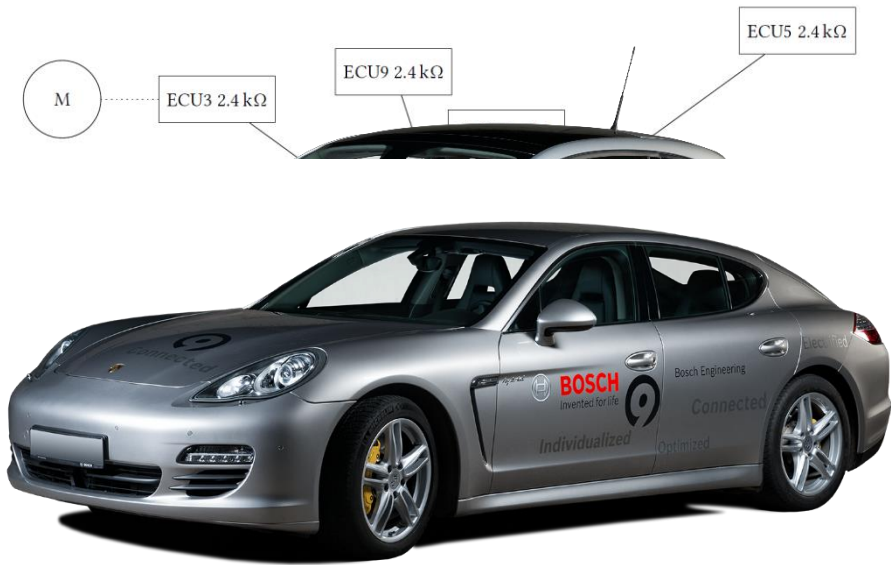
- ▶ Due to interferences (start of a strong consumer)
- ▶ Alarm if probability of invalid ECU exceeds threshold t_{\max} (e.g. 70 %)
- ▶ Leads to a higher false negative rate

Evaluation

	ECUs	Frames	Avg. accuracy	Min. accuracy
Prototype	10	56,560	99.9 %	99.58 %
Fiat	6+2	25,979	99.6 %	98.56 %
Porsche	6+2	6,389	99.88 %	99.58 %

► 99.85% Identification rate → FP after 666 frames → threshold t_{max}

		Predicted	
		No attack	Attack
Prototype	No attack	100 %	0 %
	Attack	1.5 %	98.5 %
Fiat	No attack	100 %	0 %
	Attack	0 %	100 %
Porsche	No attack	100 %	0 %
	Attack	3.18 %	96.82 %



Conclusion

- ▶ Sender identification based on physical properties of CAN signals
- ▶ Reduction in the necessary hardware requirements
- ▶ Evaluated on series production vehicles
 - ▶ High identification rate
 - ▶ No false positives
- ▶ Scission can improve the security of modern vehicles
 - ▶ IDS extension
 - ▶ Additional security functionality for gateways
 - ▶ Standalone system
- ▶ Outlook
 - ▶ Further reduction of hardware/performance requirements
 - ▶ Implementation on an embedded platform

THANK YOU!



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Stability

- ▶ Characteristics remain unchanged over several months [33]

- ▶ Fiat under changing conditions
 1. Measurement (includes training)
 - Engine off | 25°C (77°F) | 3369 frames | 100% identification
 2. Measurement
 - Driving 30 min. | 32°C (89.6°F) | 6672 frames | 100% identification
 3. Measurement (3 hours of cooling at 23°C (73.4°F))
 - Driving 20 min. | 36°C (96.6°F) | 4863 frames | 100% identification

- ▶ Biggest change in the voltage level between 0.012V and 0.026V

[33] P. S. Murvay and B. Groza. 2014. Source Identification Using Signal Characteristics in Controller Area Networks. IEEE Signal Processing Letters 21.

Reaction on intrusion

- ▶ Warn the driver

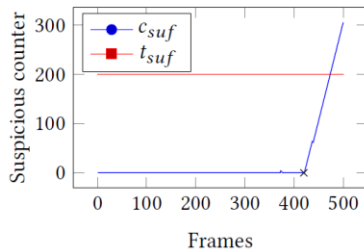
- ▶ Log the attack

- ▶ Prevent the attack
 - ▶ Invalidation of the CRC
 - ▶ Error Frame

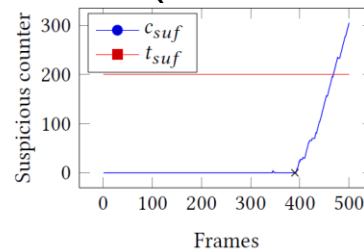
- ▶ Send the detected attack to Cloud-IDS
 1. Analyze the attack
 2. Update the in-vehicle Signature-based IDS
 3. Find the vulnerability
 4. Update the vulnerable ECU

Additional / Unknown ECU

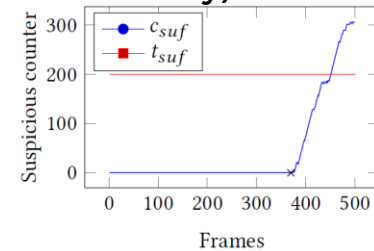
- ▶ Lower threshold t_{\min} (e.g. 30 %)
- ▶ Counter for each ECU
 - ▶ Increment if an unexpected ECUs probability > 30% but < 70%
 - ▶ Decrement if expected ECU > 30%
- ▶ Additional ECU (connected to the bus after training)
 - ▶ Counter of several ECUs will rise (no frames are necessary)



(a) Prototype

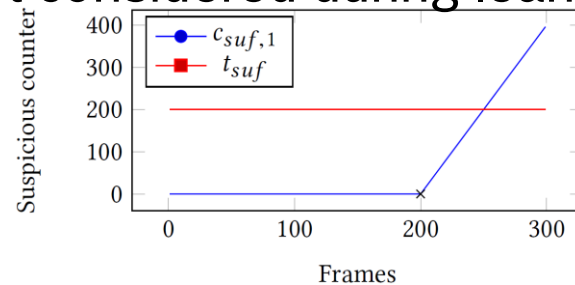


(b) Fiat 500



(c) Porsche Panamera

- ▶ Unknown ECU (connected but not considered during learning)
 - ▶ Detection like normal attack or
 - ▶ Counter of the faked ECU will rise



Scission-aware Attacker

- ▶ Influencing all ECUs (draining battery)
 - ▶ Quick and significantly → System maybe inactive during model adaption
 - ▶ Slow → System adapts model continuously
- ▶ Influencing its own signal (heating up / cooling down) to impersonate another ECU
 - ▶ No information about its own or the signal of the other ECU
 - ▶ Several signal characteristics must be similar
 - ▶ Precise adaption must be possible